

Contributions to $\kappa - \mu$ models in wireless systems

by

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Declaration

I, Priyanka Nagar, declare that the mini-dissertation, which I hereby submit for the degree MSc (Mathematical Statistics) at the University of Pretoria, is my own work and has not previously been submitted by me for a degree at this or any other tertiary institution.

SIGNATURE:.....

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Abstract

The $\kappa - \mu$ distribution is a popular model for the fading observed due to clusters of obstacles between a transmitter and a receiver in wireless communication channels. This model includes various classical fading distributions as special cases. The basic $\kappa - \mu$ distribution is used to model the signal strength in communication channels under the assumption that the contribution to the signal strength arising from each cluster follows a normal distribution. This study extends the model by relaxing the assumption of normality, instead these contributions are assumed to follow an elliptical law. The t-distribution plays an important role in this dissertation in that it is presented as an alternative to the normal distribution.

Three extended $\kappa - \mu$ models are presented in this study, in each case the model is extended by generalizing the underlying distributional assumption to that of an elliptical distribution. The models include a univariate model, a bivariate model (which is useful when one wants model the joint behaviour of communication channels) and a composite model (which is used to model shadowing as well). As a performance measure for the degradation in communication channels resulting from fading and shadowing, the outage probability is investigated for special cases of the extended models.

The proposed extensions to the $\kappa - \mu$ model result in more flexible distributions that can be used to model wireless communication channels.

Keywords: $\kappa - \mu$ model, elliptical class, fading, outage probability, shadowing, t-distribution.

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Contents

CONTENTS	v
1 INTRODUCTION	1
1.1 Background and motivation	1
1.2 Elements for study	3
1.2.1 Elliptical Class	3
1.2.2 Composite model	4
1.2.3 Concepts in communication systems	4
1.2.3.1 Fading and Shadowing	4
1.2.3.2 Metrics	6
1.3 Outline	7
2 SPECIAL FUNCTIONS AND NOTATION	8
2.1 Abbreviations and notation	8
2.2 Special functions and theory	9
3 UNIVARIATE MODELS	19
3.1 Introduction	19
3.2 The $\kappa - \mu$ distribution	20
3.2.1 Description	20
3.2.2 Derivation	21
3.2.3 Characteristics	25
3.2.3.1 Cumulative distribution function	25
3.2.3.2 Moments	26
3.2.3.3 Amount of fading	28
3.2.4 Special cases	28
3.2.4.1 The $\kappa - \mu$ distribution for $\kappa \rightarrow 0$ (Nakagami - m)	28
3.3 The $\kappa - \mu$ type distribution	29
3.3.1 Description	29

3.3.2	Derivation	30
3.3.3	Characteristics	36
3.3.3.1	Cumulative distribution function	36
3.3.3.2	Moments	40
3.3.3.3	Amount of fading	43
3.4	Performance measures	43
4	BIVARIATE MODELS	46
4.1	Introduction	46
4.2	The bivariate $\kappa - \mu$ distribution	46
4.2.1	Description	46
4.2.2	Derivation	47
4.2.3	Characteristics	53
4.2.3.1	Joint cumulative distribution function	53
4.2.3.2	Moments	58
4.2.4	Outage Probability	60
4.3	The bivariate $\kappa - \mu$ type distribution	61
4.3.1	Description	61
4.3.2	Derivation	62
4.3.3	Characteristics	69
4.3.3.1	Joint cumulative distribution function	69
4.3.3.2	Moments	75
4.3.4	Outage Probability	78
4.3.4.1	Special cases	78
4.4	Weight approximation function	82
4.5	Performance Measures	83
5	COMPOSITE MODELS	86
5.1	Introduction	86
5.2	The $\kappa - \mu$ composite distribution	87
5.2.1	Description	87
5.2.2	Derivation	87
5.2.3	Characteristics	89
5.2.3.1	Cumulative distribution function	89
5.2.3.2	Moments	90
5.2.3.3	Amount of fading	91
5.3	The $\kappa - \mu$ type composite model	91
5.3.1	Description	91

5.3.2	Derivation	92
5.3.3	Characteristics	94
5.3.3.1	Cumulative distribution function	94
5.3.3.2	Moments	96
5.4	Performance Measures	98
6	SUMMARY AND FUTURE WORK	101
	BIBLIOGRAPHY	103

Chapter 1

Introduction

1.1 Background and motivation

With the increasing popularity of wireless technologies, wireless communication systems are abundant. The vast networks of wireless channels are important tools for a fast and efficient means of data transmission. Hence, a careful study of the characteristics of these channels is required in an attempt to ensure that communication can be maintained with reliability. The signal transmission and the degradation of the signal as they pass through these channels are of interest to practitioners.

Consider the case where a channel is transmitted from a single transmitter to a single receiver. For engineering purposes, it is very important to be able to accurately model the strength of the channel. A transmission of this kind is typically subject to various disturbances. For example, there might be several large buildings, or even several clusters of buildings, between the transmitter and the receiver. On the one hand, these obstacles will impede the channels transmission by partially blocking the most direct route between the transmitter and the receiver. On the other hand, some of the obstacles might enhance the magnitude by reflecting the channel in the direction of the receiver. However, the resulting channels will have a reduced strength compared to the original channel because of the imperfections introduced by the reflection. These interferences are jointly referred to as fading and shadowing.

There is a wide variety of models available in literature to describe the statistical fluctuations in channels in fading or shadowing, fading models. The reader is referred to Simon and Alouini, (2005) [29], Shankar (2011) [28, page 247], Paris (2014) [23] and Moreno-Pozas et al (2016) [22].

Various models, with varying degrees of complexity, have been proposed in order to model the strength in a wireless channel. An appropriate model for this should be able to

1. INTRODUCTION

1.1. Background and motivation

allow for fading and shadowing. The basic model considered in this dissertation is known as the $\kappa - \mu$ model. In this model and its extensions, fading and shadowing are taken into account by considering the contribution of each cluster of obstacles to the received channel strength separately. It is assumed that a large number of reflections occur within each cluster and that each of these reflections add a random element to the resulting contribution of this cluster. The contributions of the various clusters are then aggregated in order to arrive at the strength of the channel received.

Under the standard $\kappa - \mu$ model the contribution associated with each cluster of obstacles is assumed to have an underlying model. A question that arises naturally is how sensitive the results of analyses on wireless channel models are to various departures from normality. This study will generalize the $\kappa - \mu$ model used in the communication system field by replacing the assumption of normality in the individual contributions by the assumption that these contributions follow an elliptical law. Ollila et al (2011)[21] pointed out that a more general assumption than the normal may not be far from reality. The main contribution of this study lies in relaxing the assumption of normality to that of an elliptical distribution. Different members of the elliptical class may be explored as possible candidates. Yacoub (2007) [35] and Stein (1987) [32] raised the inadequacy of the tails of some distributions. Therefore, the introduction of the t-distribution as an alternative to normality will play a central role in this study. Against this backdrop the study of various distributions within the elliptical class is of practical as well as theoretical interest.

The $\kappa - \mu$ model originated due to the fundamental differences in the underlying assumptions of this model when compared to classical fading models. The classical models assume that the received channel can be represented as a vector sum of scattered components coming from individual obstacles, while under the $\kappa - \mu$ model the assumption is that the received channel power arrives from obstacles that are scattering clusters. The underlying assumption of this model remains the normality of the individual contributions. In this study the new univariate $\kappa - \mu$ *type* model (that arise from the elliptical assumption) is proposed as an alternative in the communications environment to model and study the performance of the channel transmission through wireless channels.

In communication engineering fading distributions that are associated with multi-antenna wireless communication systems operating over *correlated* branches are also of interest. If the antennas are sufficiently separated, it is reasonable to assume that channels received by different antennas are independent. However, this assumption is unsuitable for systems with closely spaced antennas, such as mobile phones. Thus, in practical scenarios, the channels may present some sort of dependence. This is the motivation why *bivariate*

distributions have been extensively explored in the context of wireless systems; see for example Lopez-Martinez et al (2013) [15]. Ermolova and Tirkkonen (2015) [10], Reig et al (2014) [25], Mendes and Yacoub (2007) [18], de Souza et al (2012) [9] and de Souza and Yacoub (2008) [8]. For that reason, the *bivariate* $\kappa - \mu$ model will be investigated and extended to the elliptical case.

Another model of interest is the *shadowed fading* model (composite model) which simultaneously accounts for fading and shadowing in wireless systems. Shadowing can be incorporated in multipath fading models in various ways. Different approaches and models have been explored in the context of wireless systems to model random fluctuations in the received channel caused by short term fading and shadowing; see Cotton (2015) [7], Lopez-Fernandez et al (2017) [14], Lopez-Martinez et al (2016) [16], Moreno-Pozas et al (2016) [22], Paris (2014) [23], Reig et al (2014) [25], Shankar (2005) [27], Sofotasios and Freear (2013) [31], Sofotasios and Freear (2015) [30], Vural et al (2015) [34] and Yoo and Cotton (2015) [36]. In this study the focus will also be on the $\kappa - \mu$ *shadowed fading* model.

Each model is characterized in terms of quantitative measures such as the amount of fading, and the outage probability. Therefore these metrics will receive attention for special cases of the extended models. The new results in this study give a more general approach with additional flexibility since the assumption of normality is relaxed. The author trusts that the extensions obtained by assuming elliptical distributions will enrich the models within the communication systems and stimulate research.

1.2 Elements for study

Below we consider various important concepts that will be used throughout this dissertation.

1.2.1 Elliptical Class

A random variable is said to belong to the elliptical class if its probability density function (pdf) is a function of a quadratic form. This class includes the normal, Student's t, Cauchy, Pearson VII, Bessel and many other distributions. The elliptical class is useful as it includes distributions with heavier tails than the normal.

Description 1 [2]

A random variable X has an elliptical distribution with parameters μ and σ^2 , $E(\mu, \sigma^2)$,

if its pdf is of the form

$$f_X(x) = h \left[\frac{(x - \mu)^2}{2\sigma^2} \right]$$

for some non-negative function $h[\cdot]$, say density generator. The expected value is $E(X) = \mu$ and the $var(X) = k\sigma^2$, where k is a constant given by $k = -2 \left[\frac{d}{dy} \psi(y) \right]_{y=0} = -2\psi'(0)$, where $\psi(\cdot)$ is defined as some characteristic generator function. Chu (1973) [6] showed that the pdf of the elliptical distribution can be expressed as integrals of a set of normal pdfs. There is a scalar function, $W(t)$, defined on $0 < t < \infty$ such that

$$f_X(x) = \int_0^\infty W(t) f_{N(\mu, t^{-1}\sigma^2)}(x) dt$$

where $N(\mu, t^{-1}\sigma^2)$ is the normal pdf with mean μ and variance $t^{-1}\sigma^2$. This analytical framework provides a computationally convenient form and will be demonstrated in the subsequent chapters.

1.2.2 Composite model

Description 2

A random variable X with pdf $f(x)$ is said to have a compound or composite pdf if it has the general form

$$f(x) = \int f(x|\theta) h(\theta) d\theta$$

where $f(x|\theta)$ is a conditional pdf depending on the parameter θ , itself subject to variation described by the pdf $h(\theta)$, the compounding pdf. In literature relevant to the communication systems the compound model is referred to as a composite model.

1.2.3 Concepts in communication systems

1.2.3.1 Fading and Shadowing

As mentioned before, in wireless communications, signals transmitted do not often reach the receiver directly. The signal reaches the receiver after undergoing scattering, diffraction, reflection, etc. from the buildings, trees, and other structures in the channel between the transmitter and the receiver. As a result, there are multiple paths available for the signal to reach the receiver and the received signals have in-phase added (see Figure 1.1). The envelope (or amplitude/magnitude) of the signals from these paths are considered to be random variables, thus, the received power (square of the magnitude) is also considered random.

1. INTRODUCTION
1.2. Elements for study

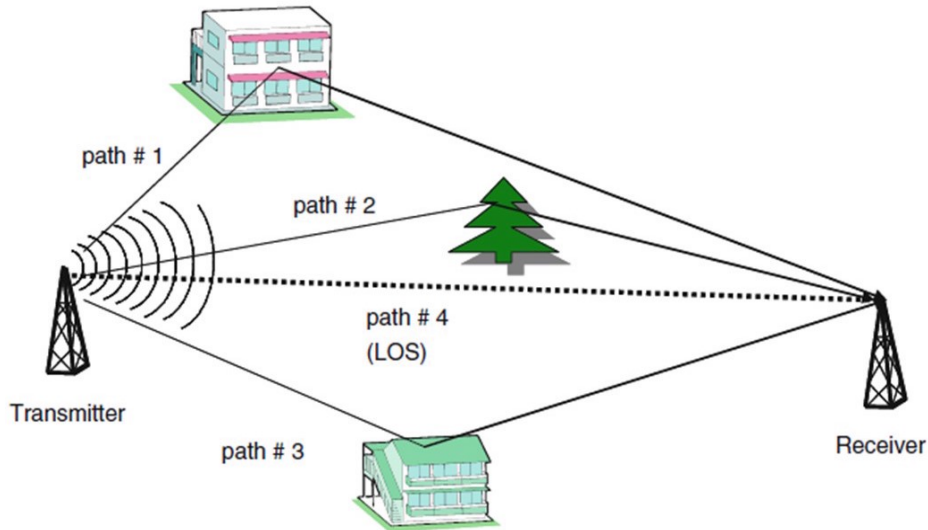


Figure 1.1: Multipath transmission of signals [28]

The random variation of the power is classified as “fading” in wireless systems. If the variation has a short period, it is known as short-term fading. Signals encountering multiple scattering or variation for long periods is known as long-term fading or “shadowing”.

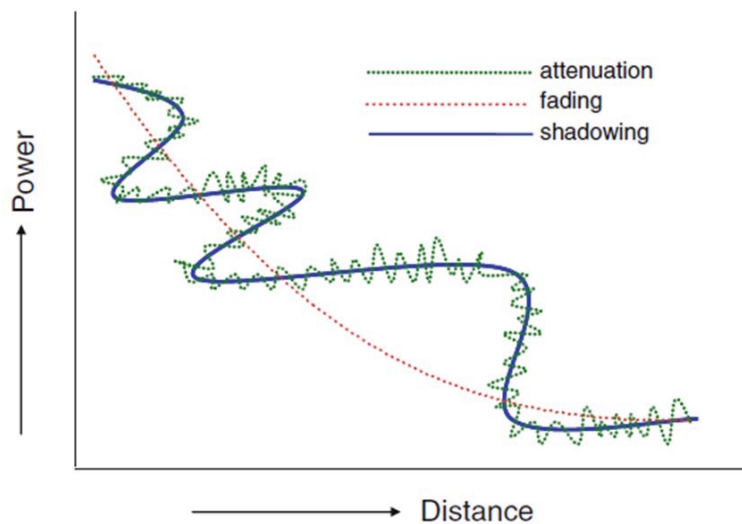


Figure 1.2: Fading and shadowing in wireless channels [28]

Signal degradation results from both the individual and/or simultaneous presence of fading and shadowing (see Figure 1.2). Several models have been used to describe fading in wireless systems, these include the simple models such as Rayleigh, Rician, Nakagami, gamma and Weibull (Simon and Alouini (2005) [29] and Shankar (2011) [28]), and complex models such as the $\kappa - \mu$ and $\eta - \mu$ (Yacoub (2007) [35]) and $\alpha - \eta - \mu$ models (Badarneh and Aloqlah (2016)[5]). The lognormal and normal models are most commonly used to describe shadowing. Composite models are used to account for the simultaneous presence of fading and shadowing.

1.2.3.2 Metrics

Comparison of these models is based on the quantitative metrics of the statistical characteristics of the models which include amongst others the amount of fading and outage probability.

Amount of fading

The amount of fading (AF) is a metric used to measure the severity of the fading in the wireless system. The AF is defined in terms of the first and second moments of the power of the fading signal, say Ω . The AF is expressed as (Simon and Alouini, 2005.[29, equation (2.5), page 18])

$$AF = \frac{var(\Omega)}{E(\Omega)^2} = \frac{E(\Omega^2) - [E(\Omega)]^2}{[E(\Omega)]^2} = \frac{E(\Omega^2)}{[E(\Omega)]^2} - 1 \quad (1.1)$$

where $E(\cdot)$ denotes the expected value and $var(\cdot)$ the variance. The AF for some fading models is given by Shankar (2011) [28, page 247].

Outage

The outage probability is a metric used to quantify the performance of the wireless communication systems in different channels. It is defined as the probability that the received signal falls below a certain threshold or equivalently, the probability that the signal-to-noise ratio (SNR) or power falls below a certain specified threshold. Whenever the signal power goes below the set threshold the channel goes into outage. The outage probability can be expressed as (Simon and Alouini, 2005.[29, equation (2.5), page 18])

$$P_{out} = \int_0^{Z_T} f(\omega) d\omega = F_\omega(Z_T) \quad (1.2)$$

where Z_T is the threshold SNR and W can denote either the instantaneous SNR, $W = \frac{R^2 E_s}{N_0}$, or the normalized power, $\Omega = R^2$, where R denotes the envelope (E_s is the energy per symbol and N_0 is the noise power, see Simon and Alouini, 2005.[29, equation (2.5),

page 18]). The pdf of the SNR is $f(\omega)$ and $F_\omega(Z_T)$ is the cdf of the SNR evaluated at $\omega = Z_T$.

1.3 Outline

- In Chapter 2, a collection of some fundamental mathematical results are given for use in later sections. Notation and abbreviations are also defined.
- In Chapter 3, firstly, the univariate $\kappa-\mu$ distribution with underlying normal model is reviewed, and thereafter, the univariate $\kappa-\mu$ type distribution with underlying elliptical model is derived. Closed form expressions for the envelope pdf, power pdfs, power cdfs and moments are derived.
- Chapter 4 expands the univariate cases in chapter 3 to the bivariate environment specifically to the elliptical case. Closed form expressions for the envelope pdf, envelope cdf, joint moments and outage probability are derived.
- Chapter 5, considers the composite model where the univariate $\kappa-\mu$ fading channel is subject to shadowing. An extended model is developed emanating from the elliptical assumption. Closed form expressions for the power pdf, power cdf and moments are derived.

These models originated from a underlying physical model. The focus of this study is to revisit the existing models and extend it from an elliptical viewpoint. Note that in the case of the elliptical assumption, the distributions are referred to as *type*. Some performance analysis is illustrated to justify the elliptical extension and the value added is specifically demonstrated by assuming the t-distribution

Chapter 2

Special functions and notation

A collection of some fundamental mathematical results are given in this chapter. These results will be used in later chapters.

2.1 Abbreviations and notation

AF	Amount of fading
cdf	Cumulative distribution function
jcdf	Joint cumulative distribution function
jpgf	Joint probability density function
pdf	Probability density function
SNR	Signal-to-Noise Ratio
$A > 0$	A is a positive definite matrix
$A^{\frac{1}{2}}$	Unique positive definite square root of A
$\det(A)$	Determinant of the square matrix A
$ A $	Norm of vector A
A^{-1}	Inverse of a square matrix A
A^T	Transpose of matrix A
$tr(A)$	Trace of the square matrix A
$J(X \rightarrow f(X))$	The Jacobian of the matrix transformation f
R^+	Set of positive real numbers
N	Set of natural numbers
Re	Real part of a number
R^p	Set of real vectors of size p
$C_k^v(t)$	Gegenbauer function

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

$\Gamma(\cdot)$	Gamma function
$\gamma(\cdot)$	Incomplete gamma function
$B(\cdot)$	Beta function
$(\alpha)_k$	Pochhammer coefficient
${}_rF_s(\cdot)$	Hypergeometric series with r upper parameters and s lower parameters
$I_\nu(\cdot)$	Modified Bessel function of the first kind and order ν
$K_\nu(\cdot)$	Modified Bessel function of the second kind and order ν
$N(\mu, \sigma^2)$	Normal distribution with mean μ and variance σ^2
$\chi^2(\nu)$	Chi-squared distribution with ν degrees of freedom
$\chi_\lambda^2(\nu)$	Non-central chi-squared distribution with ν degrees of freedom and noncentrality parameter λ
$E(\mu, \sigma^2)$	Elliptical distribution with parameters μ and σ^2

2.2 Special functions and theory

Result 1 [1, Abramowitz, M. and Stegun, I.A.,1964, equation 6.1.1]

The gamma function, denoted $\Gamma(\alpha)$, is defined as

$$\Gamma(\alpha) = \int_0^{\infty} t^{\alpha-1} \exp(-t) dt, \quad (2.1)$$

where $\text{Re}(\alpha) > 0$.

Let $\alpha \in \mathbb{R}^+$ then

$$\Gamma(\alpha + 1) = \alpha \Gamma(\alpha) \quad (2.2)$$

and if $\alpha \in \mathbb{N}$, then

$$\Gamma(\alpha) = (\alpha - 1)! \quad (2.3)$$

Result 2 [1, Abramowitz, M. and Stegun, I.A.,1964, equation 6.2.1]

The beta function, denoted $B(\alpha, \beta)$, is defined as

$$B(\alpha, \beta) = \int_0^1 t^{\alpha-1} (1-t)^{\beta-1} dt = \frac{\Gamma(\alpha) \Gamma(\beta)}{\Gamma(\alpha + \beta)}, \quad (2.4)$$

where $\text{Re}(\alpha) > 0$ and $\text{Re}(\beta) > 0$.

Result 3 [1, Abramowitz, M. and Stegun, I.A.,1964, equation 6.5.2]

The incomplete gamma function, denoted $\gamma(\alpha, x)$, is defined as

$$\gamma(\alpha, x) = \int_0^x t^{\alpha-1} \exp(-t) dt, \quad (2.5)$$

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

where $\text{Re}(\alpha) > 0$.

Result 4 [13, Gradshteyn, I. S. and Ryzhik, I. M., 2007, equation 8.354.1]

An infinite series approximation for the incomplete gamma function is expressed as

$$\gamma(\alpha, x) = \sum_{k=0}^{\infty} \frac{(-1)^k x^{\alpha+k}}{k! (\alpha+k)}. \quad (2.6)$$

Result 5 [1, Abramowitz, M. and Stegun, I.A., 1964, equation 4.2.1]

The series expansion for the exponential function is defined as

$$\exp(z) = \sum_{k=0}^{\infty} \frac{(z)^k}{k!}. \quad (2.7)$$

Result 6 [1, Abramowitz, M. and Stegun, I.A., 1964, equation 6.1.22]

The Pochhammer coefficient is defined as

$$(\alpha)_k = \alpha(\alpha+1)\dots(\alpha+k-1) = \frac{\Gamma(\alpha+k)}{\Gamma(\alpha)}, \quad (2.8)$$

where $k = 1, 2, \dots$, $(\alpha)_0 = 1$, $\alpha \neq 0$, $\text{Re}(\alpha) > 0$, $\text{Re}(\alpha+j) > 0$ and $\Gamma(\cdot)$ is the gamma function.

Result 7 [13, Gradshteyn, I. S. and Ryzhik, I. M., 2007, equation 9.14.1]

The hypergeometric series with r upper parameters and s lower parameters is defined for $|x| < 1$ as

$${}_rF_s(\alpha_1, \dots, \alpha_r; \beta_1, \dots, \beta_s; x) = \sum_{k=0}^{\infty} \frac{(\alpha_1)_k \dots (\alpha_r)_k x^k}{(\beta_1)_k \dots (\beta_s)_k k!}, \quad (2.9)$$

where $(\alpha)_k$ is the Pochhammer symbol (2.8).

Result 8 [13, Gradshteyn, I. S. and Ryzhik, I. M., 2007]

The confluent hypergeometric function is defined for $|x| < 1$ as

$${}_1F_1(\alpha; \beta; x) = \sum_{k=0}^{\infty} \frac{(\alpha)_k x^k}{(\beta)_k k!}. \quad (2.10)$$

Result 9 [13, Gradshteyn, I. S. and Ryzhik, I. M., 2007]

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

The Gaussian hypergeometric function is defined as

$${}_2F_1(\alpha_1, \alpha_2; \beta; x) = \sum_{k=0}^{\infty} \frac{(\alpha_1)_k (\alpha_2)_k x^k}{(\beta)_k k!}, \quad (2.11)$$

where $|x| < 1$.

Note that

$${}_2F_1(\alpha_1, \alpha_2, \beta; x) = (1-x)^{-\alpha_1} {}_2F_1\left(\alpha_1, \beta - \alpha_2, \beta; \frac{x}{x-1}\right). \quad (2.12)$$

Result 10 [1, Abramowitz, M. and Stegun, I.A., 1964, equation 9.6.10]

The modified Bessel function of the first kind and order v can be expressed as an infinite series

$$I_v(x) = \sum_{k=0}^{\infty} \left(\frac{x}{2}\right)^v \frac{\left(\frac{1}{4}x^2\right)^k}{k! \Gamma(v+k+1)}. \quad (2.13)$$

Result 11 [1, Abramowitz, M. and Stegun, I.A., 1964, equation 9.6.7]

The modified Bessel function of the first kind and order v can be approximated as

$$I_v(x) \simeq \frac{\left(\frac{x}{2}\right)^v}{\Gamma(v+1)}. \quad (2.14)$$

Result 12 [13, Gradshteyn, I. S. and Ryzhik, I. M., 2007, equation 8.351.2]

The incomplete gamma function can be expressed as follows

$$\gamma(\alpha, \beta) = \frac{\beta^\alpha}{\alpha} \exp(-\beta) {}_1F_1(1; 1 + \alpha; \beta) \quad (2.15)$$

where ${}_1F_1(\cdot)$ is defined in (2.10).

Result 13 [19, Miller, K., 1964, equation 2.2.23]

The generalized Neumann addition formula

$$\begin{aligned} & |a\xi - b\Omega|^{-v} I_v(|a\xi - b\Omega|) \\ &= 2^v \Gamma(v) \sum_{k=0}^{\infty} (-1)^k (v+k) \frac{I_{v+k}(a|\xi|)}{a^v |\xi|^v} \frac{I_{v+k}(b|\Omega|)}{b^v |\Omega|^v} C_k^v(\cos \phi) \end{aligned} \quad (2.16)$$

where ϕ is the angle between ξ and Ω and a and b are real, $I_v(\cdot)$ is the Bessel function of the first kind (2.13) and $C_k^v(t)$ is the Gegenbauer function. The Gegenbauer function is defined by [13, Gradshteyn, I. S. and Ryzhik, I. M., 2007, equation 8.930] as follows: The polynomials $C_k^v(t)$ of degree k are the coefficients of the α^k in the power-series expansion of the function

$$(1 - 2t\alpha + \alpha^2)^{-v} = \sum_{k=0}^{\infty} C_k^v(t) \alpha^k$$

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

Result 14 [19, Miller, K., 1964, equation 2.2.25]

The Gegenbauer's generalization of the Poisson formula

$$\begin{aligned} & \int_0^\pi \exp(\omega_1 \cos \phi) C_k^v(\cos \phi) \sin^{2v} \phi d\phi \\ &= \frac{2^v \Gamma(v + \frac{1}{2}) \Gamma(\frac{1}{2}) \Gamma(2v + 2)}{b^v n! \Gamma(2v)} I_{v+n}(b) \end{aligned} \quad (2.17)$$

where b is real and $I_v(\cdot)$ is the modified Bessel function of the first kind (2.13).

Result 15 [19, Miller, K., 1964, equation 1.5.19]

Evaluation of the expression in terms of generalized spherical co-ordinates results in

$$\begin{aligned} & R_1^{2\mu-1} \int_0^{2\pi} d\theta \prod_{k=2}^{2\mu-2} \int_0^\pi \sin^{2\mu-1-k} \phi_k d\phi_k \\ &= \frac{2\pi^{\frac{1}{2}(2\mu-1)} R_1^{2\mu-1}}{\Gamma(\frac{1}{2}(2\mu-1))}. \end{aligned} \quad (2.18)$$

Result 16 [1, Abramowitz, M. and Stegun, I.A., 1964, equation 6.1.1]

The following integral can be evaluated as

$$\int_0^\infty x^{\alpha-1} \exp(-\beta x) dx = \frac{\Gamma(\alpha)}{\beta^\alpha}, \quad (2.19)$$

where $\text{Re}(\alpha) > 0$ and $\text{Re}(\beta) > 0$.

Result 17 [13, Gradshteyn, I. S. and Ryzhik, I. M., 2007, equation 3.381.8]

The integral can be evaluated as

$$\int_0^u x^m \exp(-\beta x^n) dx = \frac{\gamma(v, \beta u^n)}{n\beta^v}, \quad v = \frac{m+1}{n}, \quad (2.20)$$

where $\text{Re}(v) > 0$, $\text{Re}(\beta) > 0$, $\text{Re}(n) > 0$ and $u > 0$.

Result 18 [13, Gradshteyn, I. S. and Ryzhik, I. M., 2007, equation 6.455.2]

The following integral can be evaluated as

$$\int_0^\infty x^{\mu-1} \exp(-\beta x) \gamma(v, \alpha x) dx = \frac{\alpha^v \Gamma(\mu+v)}{v(\alpha+\beta)^{\mu+v}} {}_2F_1\left(1; \mu+v; v+1; \frac{\alpha}{\alpha+\beta}\right), \quad (2.21)$$

where $\text{Re}(\alpha+\beta) > 0$, $\text{Re}(\beta) > 0$, $\text{Re}(v+\mu) > 0$ and ${}_2F_1(\cdot)$ is defined in (2.11).

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

Result 19 [24, Prudnikov, A.P., Brychkov, Y.A, et. al, 1988, equation 2.15.5.4]

The following integral can be evaluated as

$$\begin{aligned} & \int_0^{\infty} x^{\alpha-1} \exp(-\beta x^2) I_v(\gamma x) dx \\ &= 2^{-v-1} \gamma^v \beta^{-\frac{(\alpha+v)}{2}} \Gamma \left[\frac{\alpha+v}{2} \right] {}_1F_1 \left(\frac{\alpha+v}{2}; v+1; \frac{\gamma^2}{4\beta} \right), \end{aligned} \quad (2.22)$$

where $\text{Re}(\beta) > 0$, $\text{Re}(\alpha+v) > 0$, $|\arg(\gamma)| < \pi$, $\Gamma \left[\begin{smallmatrix} a, b \\ a+b \end{smallmatrix} \right] = \frac{\Gamma(a)\Gamma(b)}{\Gamma(a+b)} = B(a, b)$ and ${}_1F_1(\cdot)$ is defined in (2.10). (Note that $\Gamma \left[\begin{smallmatrix} a \\ a+b \end{smallmatrix} \right] = \frac{\Gamma(a)}{\Gamma(a+b)}$.)

Result 20 [13, Gradshteyn, I. S. and Ryzhik, I. M. , 2007, equation 7.621.4]

The following integral can be evaluated as

$$\int_0^{\infty} x^{\beta-1} \exp(-sx) {}_1F_1(\alpha; \lambda; kx) dx = \Gamma(\beta) s^{-\beta} {}_2F_1(\alpha, \beta; \lambda; ks^{-1}), \quad (2.23)$$

where $|s| > |k|$, $\text{Re}(\beta) > 0$ and ${}_1F_1(\cdot)$ is defined in (2.10).

Result 21 [13, Gradshteyn, I. S. and Ryzhik, I. M. , 2007, equation 3.194.2]

The following integral can be evaluated as

$$\int_0^u \frac{x^{\alpha-1}}{(1+\beta x)^v} dx = \frac{u^\alpha}{\alpha} {}_2F_1(v, \alpha; 1+\alpha; -\beta u), \quad (2.24)$$

where $\text{Re}(\alpha) > 0$ and $|\arg(1+\beta u)| < \pi$.

Result 22 [1, Abramowitz, M. and Stegun, I.A., 1964, equation 29.1.1]

The Laplace transform is defined as

$$f(s) = L\{F(t)\} = \int_0^{\infty} \exp(-st) F(t) dt, \quad (2.25)$$

where $F(t)$ is a real valued function and s a complex random variable. The function $F(t)$ is called the original function and $f(s)$ the image function.

Result 23 [1, Abramowitz, M. and Stegun, I.A., 1964, equation 29.3.81]

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

Given the image function

$$f(s) = \frac{1}{s^\mu} \exp\left\{\frac{k}{s}\right\}, \quad [\mu > 0],$$

the inverse Laplace transform of the original function is

$$F(t) = \left(\frac{t}{k}\right)^{\frac{\mu-1}{2}} I_{\mu-1}\left(2\sqrt{kt}\right), \quad (2.26)$$

where $I_\nu(\cdot)$ is the modified Bessel function of the first kind (2.13).

Result 24 [19, Miller, K., 1964, equation 2.1.1]

Let $X \in R^p$ follow the multivariate normal distribution with mean parameter $\mu \in R^p$ and covariance matrix $\Sigma > 0$, then the pdf is given as

$$f_X(x) = \frac{1}{(2\pi)^{\frac{p}{2}} \det(\Sigma)^{\frac{1}{2}}} \exp\left\{-\frac{1}{2}(x - \mu)^T \Sigma^{-1}(x - \mu)\right\}. \quad (2.27)$$

Result 25 [6, Chu, K, 1973]

If X is an elliptical random variable with parameters μ and σ^2 and pdf $f(X)$, then there is a scalar function $W(t)$ defined on $0 < t < \infty$ such that

$$f_X(x) = \int_0^\infty W(t) f_{N(\mu, t^{-1}\sigma^2)}(x) dt, \quad (2.28)$$

where $f_{N(\mu, t^{-1}\sigma^2)}(x)$ is the normal pdf with mean μ and variance $t^{-1}\sigma^2$.

Result 26 [6, Chu, K, 1973]

Let $X \in R^p$ follow the multivariate elliptical distribution with mean parameter $\mu \in R^p$ and covariance matrix $\Sigma > 0$. Then for a scalar function $W(t)$ defined on $0 < t < \infty$ it follows that

$$f_X(x) = \int_0^\infty \frac{1}{(2\pi)^{\frac{p}{2}} \det(t^{-1}\Sigma)^{\frac{1}{2}}} \exp\left\{-\frac{1}{2}(x - \mu)^T (t^{-1}\Sigma)^{-1}(x - \mu)\right\} W(t) dt, \quad (2.29)$$

where $f_{N(\mu, t^{-1}\Sigma)}(x)$ is the multivariate normal pdf with expected value $\mu \in R^p$ and covariance matrix $t^{-1}\Sigma$.

Result 27 [28, Shankar, P.M., 2011, equation 2.83]

A random variable X is said to have a non-central chi-squared distribution with k degrees of freedom and non-centrality parameter λ , denoted as $X \sim \chi_\lambda^2(k)$, if X has pdf

$$f_X(x) = \frac{1}{2} \exp\left\{-\frac{x + \lambda}{2}\right\} \left(\frac{x}{\lambda}\right)^{\frac{k-2}{4}} I_{\frac{k-2}{2}}\left(\sqrt{\lambda x}\right), \quad x > 0, \quad (2.30)$$

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

where $k > 0$ and $I_v(\cdot)$ is the modified Bessel function of the first kind (2.13). The expected value is given by

$$E(X) = k + \lambda. \quad (2.31)$$

Result 28 [28, Shankar, P.M., 2011, equation 2.72]

A random variable X is said to have the Nakagami- m distribution if X has pdf

$$f_X(x) = \frac{2m^m x^{2m-1}}{\Omega^m \Gamma(m)} \exp\left\{-\frac{mx^2}{\Omega}\right\}, \quad x > 0, \quad (2.32)$$

where $m \geq \frac{1}{2}$ with m the Nakagami-parameter and $\Omega > 0$.

Result 29 [28, Shankar, P.M., 2011, equation 2.94]

A random variable X is said to have the Rayleigh distribution if X has pdf

$$f_X(x) = \frac{x}{\beta} \exp\left\{-\frac{x^2}{2\beta}\right\}, \quad x > 0, \quad (2.33)$$

where $\beta > 0$.

Result 30 [28, Shankar, P.M., 2011, equation 4.185]

A random variable X is said to have the Rice distribution if X has pdf

$$f_X(x) = \frac{1}{2\sigma^2} \exp\left\{-\frac{x+d^2}{2\sigma^2}\right\} I_0\left(\frac{d\sqrt{x}}{\sigma^2}\right), \quad x > 0, \quad (2.34)$$

where $I_0(\cdot)$ is the modified Bessel function of the first kind and order zero.

Result 31 [6, Chu, K, 1973]

With reference to (2.28): The weight function for the normal distribution is defined as the dirac delta function,

$$W(t) = \delta(t-1). \quad (2.35)$$

Result 32 [6, Chu, K, 1973]

With reference to (2.28): The weight function for the t-distribution with v degrees of freedom is defined

$$W(t) = \frac{v\left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right)}{2\Gamma\left(\frac{v}{2}\right)}. \quad (2.36)$$

Result 33 [2, Arashi, M. and Nadarajah, S., 2016]

Based on the assumption that the weighting function admits the expansion, a weight approximation function is given as

$$W(t) = \sum_{k=1}^{\infty} a_k t^k. \quad (2.37)$$

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

Theorem 2.1 [19, Miller, K., 1964] Let X_n be normally distributed with mean vector A_n and positive covariance matrix M_n . Then $|X_n|$ has a one-dimensional density function. If $M_n = \psi_0 I_n$ where ψ_0 is a positive constant, then $r = |X_n|$ is called the Rayleigh variate. Then the density function of $r = |X_n|$ is

$$g(r) = \frac{a}{\psi_0} \left(\frac{r}{a}\right)^{\frac{1}{2}n} e^{-(r^2+a^2)/2\psi_0} I_{\frac{1}{2}(n-2)}\left(\frac{ra}{\psi_0}\right), \quad r \geq 0 \quad (2.38)$$

where $a = |A_n|$ and I_v is the modified Bessel function of the first kind and order v .

Proof. The density function of X_n is

$$\begin{aligned} f_{X_n}(x_n) &= \frac{1}{(2\pi\psi_0)^{\frac{n}{2}}} \exp\left\{-\frac{1}{2\psi_0} (X_n - A_n)'(X_n - A_n)\right\} \\ &= \frac{1}{(2\pi\psi_0)^{\frac{n}{2}}} \exp\left\{-\frac{1}{2\psi_0} [X_n'X_n - X_n'A_n - A_n'X_n + A_n'A_n]\right\} \\ &= \frac{1}{(2\pi\psi_0)^{\frac{n}{2}}} \exp\left\{-\frac{1}{2\psi_0} [|X_n|^2 + |A_n|^2 - 2X_n'A_n]\right\} \\ &= \frac{1}{(2\pi\psi_0)^{\frac{n}{2}}} \exp\left\{-\frac{1}{2\psi_0} [|X_n|^2 + |A_n|^2]\right\} \exp\left\{-\frac{1}{2\psi_0} [-2X_n'A_n]\right\} \\ &= \frac{1}{(2\pi\psi_0)^{\frac{n}{2}}} \exp\left\{-\frac{1}{2\psi_0} [r^2 + a^2]\right\} \exp\left\{\frac{X_n'A_n}{\psi_0}\right\}, \end{aligned} \quad (2.39)$$

where $r = |X_n|$ and $a = |A_n|$.

Introduce a complete orthonormal set of vectors $E_n^{(1)}, E_n^{(2)}, \dots, E_n^{(n)}$, with $E_n^{(1)}$ in the direction A_n . Then A_n is the polar axis. If we make the generalized spherical coordinate change of variable

$$\begin{aligned} x_j &= r \left(\prod_{k=1}^{j-1} \sin \phi_k \right) \cos \phi_j, \quad 1 \leq j \leq n-2, \\ x_{n-1} &= r \left(\prod_{k=1}^{n-2} \sin \phi_k \right) \cos \theta, \\ x_n &= r \left(\prod_{k=1}^{n-2} \sin \phi_k \right) \sin \theta, \end{aligned}$$

where $0 \leq \phi_k \leq \pi$, $1 \leq k \leq n-2$, $0 \leq \theta \leq 2\pi$ and $0 \leq r \leq \infty$, we obtain

$$g(r) = \int_{|X_n|=r} f_n(X_n) d\sigma, \quad r \geq 0 \quad (2.40)$$

as the density function of r . Substituting (2.39) in (2.40) yields

$$g(r) = \frac{1}{(2\pi\psi_0)^{\frac{n}{2}}} \exp\left\{-\frac{1}{2\psi_0} [r^2 + a^2]\right\} \int_{|X_n|=r} \exp\left\{\frac{X_n'A_n}{\psi_0}\right\} d\sigma. \quad (2.41)$$

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

The Jacobian of the transformation is

$$J_n(r, \theta, \phi_1, \phi_2, \dots, \phi_{n-2}) = r^{n-1} \prod_{k=1}^{n-2} \sin^{n-1-k} \phi_k$$

and since A_n is the polar axis, $X_n' A_n = |X_n| |A_n| \cos \phi_1 = ra \cos \phi_1$. Thus, from (2.41), we may write the integral as an iterated integral:

$$\begin{aligned} \int_{|X_n|=r} \exp\left\{\frac{X_n' A_n}{\psi_0}\right\} d\sigma &= \int_{|X_n|=r} \exp\left\{\frac{ra \cos \phi_1}{\psi_0}\right\} d\sigma \\ &= r^{n-1} \int_0^{2\pi} d\theta \int_0^\pi \exp\left\{\frac{ra \cos \phi_1}{\psi_0}\right\} \sin^{n-2} \phi_1 d\phi_1 \prod_{k=1}^{n-2} \int_0^\pi \sin^{n-1-k} \phi_k d\phi_k. \end{aligned} \quad (2.42)$$

We recall the identity

$$I_\nu(z) = \frac{1}{\Gamma(\frac{1}{2})\Gamma(\nu + \frac{1}{2})} \left(\frac{1}{2}z\right)^\nu \int_0^\pi \exp\{\pm z \cos \xi\} \sin^{2\nu} \xi d\xi, \quad (2.43)$$

for $\text{Re}(\nu) > -\frac{1}{2}$, where I_ν is the modified Bessel function of the first kind and order ν . Using (2.43), the second integral in (2.42) becomes

$$\int_0^\pi \exp\left\{\frac{ra \cos \phi_1}{\psi_0}\right\} \sin^{n-2} \phi_1 d\phi_1 = \Gamma\left(\frac{1}{2}\right)\Gamma\left(\frac{1}{2}(n-1)\right) \left(\frac{2\psi_0}{ra}\right)^{\frac{1}{2}(n-2)} I_{\frac{1}{2}(n-2)}\left(\frac{ra}{\psi_0}\right).$$

The product of integrals in (2.42) is as follows, since

$$\int_0^\pi \sin^{n-1-k} \phi_k d\phi_k = B\left(\frac{1}{2}(n-k), \frac{1}{2}\right) = \frac{\Gamma\left(\frac{1}{2}(n-k)\right)\Gamma\left(\frac{1}{2}\right)}{\Gamma\left(\frac{1}{2}(n-k+1)\right)},$$

where $B(u, w)$ is the Beta function and $\Gamma(u)$ is the Gamma function. Therefore, we conclude that

$$\prod_{k=1}^{n-2} \int_0^\pi \sin^{n-1-k} \phi_k d\phi_k = \frac{\Gamma^{n-2}\left(\frac{1}{2}\right)}{\Gamma\left(\frac{1}{2}n\right)}.$$

Hence, (2.42) reduces to

$$\begin{aligned} \int_{|X_n|=r} \exp\left\{\frac{X_n' A_n}{\psi_0}\right\} d\sigma &= r^{n-1} (2\pi) \Gamma\left(\frac{1}{2}\right) \Gamma\left(\frac{1}{2}(n-1)\right) \left(\frac{2\psi_0}{ra}\right)^{\frac{1}{2}(n-2)} \\ &\quad \times I_{\frac{1}{2}(n-2)}\left(\frac{ra}{\psi_0}\right) \frac{\Gamma^{n-2}\left(\frac{1}{2}\right)}{\Gamma\left(\frac{1}{2}n\right)} \end{aligned}$$

2. SPECIAL FUNCTIONS AND NOTATION

2.2. Special functions and theory

$$\begin{aligned}
&= r^{n-1}(2\pi)\Gamma\left(\frac{1}{2}\right)\Gamma\left(\frac{1}{2}(n-1)\right)\left(\frac{2\psi_0}{ra}\right)^{-1}\frac{\Gamma^{n-2}\left(\frac{1}{2}\right)}{\Gamma\left(\frac{1}{2}n\right)} \\
&\quad \times \left(\frac{2\psi_0}{ra}\right)^{\frac{1}{2}n} I_{\frac{1}{2}(n-2)}\left(\frac{ra}{\psi_0}\right) \\
&= \frac{a}{\psi_0}\left(\frac{2\pi r\psi_0}{a}\right)^{\frac{1}{2}n} I_{\frac{1}{2}(n-2)}\left(\frac{ra}{\psi_0}\right). \tag{2.44}
\end{aligned}$$

Substituting (2.44) in (2.41) we obtain the density function,

$$\begin{aligned}
g(r) &= \frac{1}{(2\pi\psi_0)^{\frac{n}{2}}}\exp\left\{-\frac{1}{2\psi_0}[r^2+a^2]\right\}\int_{|X_n|=r}\exp\left\{\frac{X_n'A_n}{\psi_0}\right\}d\sigma \\
&= \frac{1}{(2\pi\psi_0)^{\frac{n}{2}}}\exp\left\{-\frac{1}{2\psi_0}[r^2+a^2]\right\}\frac{a}{\psi_0}\left(\frac{2\pi r\psi_0}{a}\right)^{\frac{1}{2}n} I_{\frac{1}{2}(n-2)}\left(\frac{ra}{\psi_0}\right) \\
&= \frac{a}{\psi_0}\left(\frac{r}{a}\right)^{\frac{1}{2}n} e^{-(r^2+a^2)/2\psi_0} I_{\frac{1}{2}(n-2)}\left(\frac{ra}{\psi_0}\right).
\end{aligned}$$

■

Result 34 Let X_n follow an elliptical distribution with mean vector A_n and positive covariance matrix M_n . Then $|X_n|$ has a one-dimensional density function. If $M_n = \psi_0 I_n$ where ψ_0 is a positive constant, then $r = |X_n|$ is called the Rayleigh variate. Then using (2.28) and (2.38) the density function of $r = |X_n|$ is

$$\begin{aligned}
g(r) &= \int_0^\infty W(t) g_{N(A_n, t^{-1}\psi_0)}(r|t) dt, \quad r \geq 0 \\
&= \int_0^\infty W(t) \frac{ta}{\psi_0} \left(\frac{r}{a}\right)^{\frac{1}{2}n} e^{-t(r^2+a^2)/2\psi_0} I_{\frac{1}{2}(n-2)}\left(\frac{tra}{\psi_0}\right) dt, \quad r \geq 0 \tag{2.45}
\end{aligned}$$

where $a = |A_n|$ and I_v is the modified Bessel function of the first kind and order v . (Proof similar to theorem 2.1).

Chapter 3

Univariate Models

3.1 Introduction

Yacoub (2007) [35] pointed out that the well-known fading distributions have been derived assuming a homogenous, diffuse scattering field, that emanated from randomly distributed point scatters (see Figure 1.1). But this assumption is an approximation and a cluster based scattering model (see Figure 3.1) may be more relevant. The $\kappa - \mu$ model proposed by Yacoub (2007) [35] is a general physical fading model where the signal is characterised in terms of measurable physical parameters as described below in Section 3.2. The pdf of the power at the receiver and statistical properties of the $\kappa - \mu$ model will receive attention in Sections 3.2 and 3.3. The $\kappa - \mu$ type model that emanates from the elliptical assumption will be derived in Section 3.3; followed by section 3.4 where the performance measures under different members of the elliptical model will be illustrated.

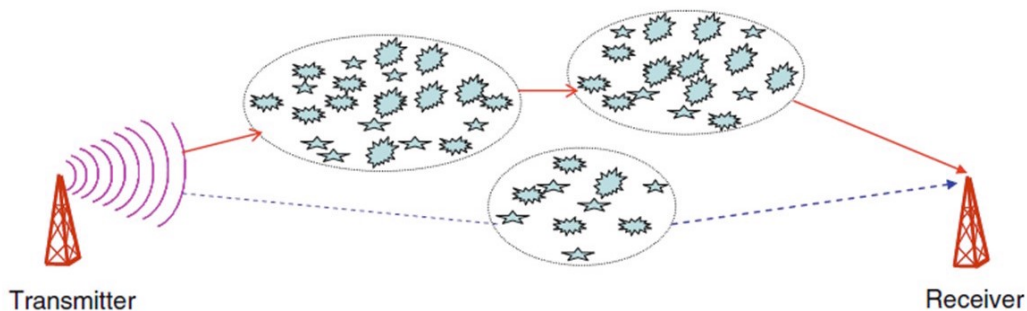


Figure 3.1: Clusters of multipath scattering [28]

3.2 The $\kappa - \mu$ distribution

3.2.1 Description

Let X_i and Y_i be mutually independent normal processes with $E(X_i) = E(Y_i) = 0$ and $\text{var}(X_i) = \text{var}(Y_i) = \sigma^2$. Then the envelope, R , of the physical model for the $\kappa - \mu$ distribution can be written in terms of the in-phase and quadrature components of the fading signal as

$$W = R^2 = \sum_{i=1}^n (X_i + p_i)^2 + \sum_{i=1}^n (Y_i + q_i)^2, \quad (3.1)$$

where p_i and q_i are the mean values of the in-phase and quadrature components of the multipath waves of cluster i and n is the number of clusters of the multipath. Since $X_i \sim N(0, \sigma^2)$ and $Y_i \sim N(0, \sigma^2)$, we can write $(X_i + p_i) \sim N(p_i, \sigma^2)$ and $X_i^* = \left(\frac{X_i + p_i}{\sigma}\right) \sim N\left(\frac{p_i}{\sigma}, 1\right)$. Thus $(X_i^*)^2$ follows a $\chi_{\frac{p_i^2}{\sigma^2}}^2(1)$ with non-centrality parameter $\frac{p_i^2}{\sigma^2}$ (see (2.30)) and similarly we can express $Y_i^* = \left(\frac{Y_i + q_i}{\sigma}\right)$ where $(Y_i^*)^2$ follows a $\chi_{\frac{q_i^2}{\sigma^2}}^2(1)$ with non-centrality parameter $\frac{q_i^2}{\sigma^2}$ (see (2.30)).

Let $\frac{R_i^2}{\sigma^2} = \frac{W_i}{\sigma^2} = (X_i^*)^2 + (Y_i^*)^2$ then $\frac{R_i^2}{\sigma^2}$ follows a $\chi_{\lambda}^2(2)$ with non-centrality parameter $\lambda = \frac{d_i^2}{\sigma^2} = \frac{p_i^2 + q_i^2}{\sigma^2}$.

For $R^2 = \sum_{i=1}^n R_i^2$, the power of the fading signal is $W = \sum_{i=1}^n W_i$, and we are interested in the pdf of the envelope R and the power W respectively.

The following is a schematic outline of the mathematical algorithm that will be followed to obtain the necessary distributions.

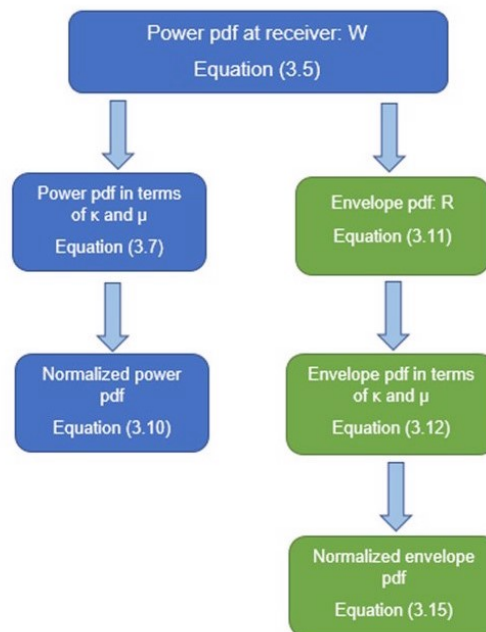


Figure 3.2

3.2.2 Derivation

Since the distribution of $\left(\frac{W_i}{\sigma^2}\right)$ is a $\chi^2_{\frac{d_i^2}{\sigma^2}}$ (2) distribution with non-centrality parameter $\frac{d_i^2}{\sigma^2}$, by using (2.30) the pdf of W_i is then expressed as

$$f_{W_i}(w) = \frac{1}{2\sigma^2} \exp\left(-\frac{(w + d_i^2)}{2\sigma^2}\right) I_0\left(\frac{d_i\sqrt{w}}{\sigma^2}\right), \quad (3.2)$$

where $I_0(\cdot)$ is the modified Bessel function of the first kind and order zero (2.13).

Remark 3.1 Note that equation (3.2) is the pdf of the Rice distribution (see (2.34))

By using (2.13) to express the Bessel function in (3.2) as an infinite series we obtain the pdf as

$$\begin{aligned} f_{W_i}(w) &= \frac{1}{2\sigma^2} \exp\left(-\frac{(w + d_i^2)}{2\sigma^2}\right) \sum_{k=0}^{\infty} \frac{\left(\frac{d_i^2 w}{4\sigma^4}\right)^k}{k! \Gamma(k+1)} \\ &= \frac{1}{2\sigma^2} \exp\left(-\frac{d_i^2}{2\sigma^2}\right) \sum_{k=0}^{\infty} \frac{\left(\frac{d_i^2}{4\sigma^4}\right)^k}{(k!)^2} w^k \exp\left(-\frac{w}{2\sigma^2}\right). \end{aligned} \quad (3.3)$$

The Laplace transform (2.25) of (3.3) is given by

$$\begin{aligned} f(s) &= L\{f_{W_i}(w)\} = \int_0^{\infty} f_{W_i}(w) \exp(-sw) dw \\ &= \int_0^{\infty} \frac{1}{2\sigma^2} \exp\left(-\frac{d_i^2}{2\sigma^2}\right) \sum_{k=0}^{\infty} \frac{\left(\frac{d_i^2}{4\sigma^4}\right)^k}{(k!)^2} w^k \exp\left(-\frac{w}{2\sigma^2}\right) \exp(-sw) dw \\ &= \frac{1}{2\sigma^2} \exp\left(-\frac{d_i^2}{2\sigma^2}\right) \sum_{k=0}^{\infty} \frac{\left(\frac{d_i^2}{4\sigma^4}\right)^k}{(k!)^2} \int_0^{\infty} w^k \exp\left\{-w\left(\frac{1}{2\sigma^2} + s\right)\right\} dw. \end{aligned}$$

Using (2.19), we obtain that

$$\begin{aligned} f(s) &= \frac{1}{2\sigma^2} \exp\left\{-\frac{d_i^2}{2\sigma^2}\right\} \sum_{k=0}^{\infty} \frac{\left(\frac{d_i^2}{4\sigma^4}\right)^k}{(k!)^2} \\ &\quad \times \frac{\Gamma(k+1)}{\left(\frac{1}{2\sigma^2} + s\right)^{k+1}}, \end{aligned}$$

and from (2.3), follows that

3. UNIVARIATE MODELS

3.2. The $\kappa - \mu$ distribution

$$\begin{aligned}
 f(s) &= \frac{1}{2\sigma^2} \exp\left\{-\frac{d_i^2}{2\sigma^2}\right\} \sum_{k=0}^{\infty} \frac{d_i^{2k}}{(4\sigma^4)^k (k!)^2} \frac{k!}{\left(\frac{1}{2\sigma^2} + s\right)^{k+1}} \\
 &= \frac{1}{2\sigma^2 \left(\frac{1}{2\sigma^2} + s\right)} \exp\left\{-\frac{d_i^2}{2\sigma^2}\right\} \sum_{k=0}^{\infty} \frac{1}{k!} \left(\frac{d_i^2}{4\sigma^4 \left(\frac{1}{2\sigma^2} + s\right)}\right)^k \\
 &= \frac{1}{(1 + 2\sigma^2 s)} \exp\left\{-\frac{d_i^2}{2\sigma^2}\right\} \sum_{k=0}^{\infty} \frac{1}{k!} \left(\frac{d_i^2}{2\sigma^2 + 4\sigma^4 s}\right)^k.
 \end{aligned}$$

Using the series expansion formula (2.7), we have that

$$\begin{aligned}
 f(s) &= \frac{1}{(1 + 2\sigma^2 s)} \exp\left(-\frac{d_i^2}{2\sigma^2}\right) \exp\left(\frac{d_i^2}{2\sigma^2 + 4\sigma^4 s}\right) \\
 &= \frac{1}{(1 + 2\sigma^2 s)} \exp\left(-\frac{d_i^2}{2\sigma^2} + \frac{d_i^2}{2\sigma^2 + 4\sigma^4 s}\right) \\
 &= \frac{1}{(1 + 2\sigma^2 s)} \exp\left\{-\frac{sd_i^2}{(1 + 2\sigma^2 s)}\right\}.
 \end{aligned}$$

Since the W_i , $i = 1, 2, \dots, n$ are independent random variables it follows that the Laplace transformation, $L\{f_W(w)\}$, of $f_W(w)$ is found to be

$$\begin{aligned}
 L\{f_W(w)\} &= \prod_{i=1}^n L\{f_{W_i}(t)\} \\
 &= \prod_{i=1}^n \frac{1}{(1 + 2\sigma^2 s)} \exp\left\{-\frac{sd_i^2}{(1 + 2\sigma^2 s)}\right\} \\
 &= \frac{1}{(1 + 2\sigma^2 s)^n} \exp\left\{-\sum_{i=1}^n \frac{sd_i^2}{(1 + 2\sigma^2 s)}\right\} \\
 &= \frac{1}{(1 + 2\sigma^2 s)^n} \exp\left\{-\frac{sd^2}{(1 + 2\sigma^2 s)}\right\}, \tag{3.4}
 \end{aligned}$$

where $d^2 = \sum_{i=1}^n d_i^2$. Hence, using (2.26) the pdf of the power is obtained from equation (3.4) as

$$f_W(w) = \frac{1}{2\sigma^2} \left(\frac{w}{d^2}\right)^{\frac{n-1}{2}} \exp\left\{-\frac{(w + d^2)}{2\sigma^2}\right\} I_{n-1}\left(\frac{d\sqrt{w}}{\sigma^2}\right), \quad w > 0, \tag{3.5}$$

where $I_v(\cdot)$ is the modified Bessel function of the first kind and order v (2.13). Since $\frac{W}{\sigma^2}$ follows a $\chi_{\frac{d^2}{\sigma^2}}^2(n)$ distribution with non-centrality parameter $\frac{d^2}{\sigma^2}$, the expected value of W is given by (2.31) as

$$E(W) = 2n\sigma^2 + d^2. \tag{3.6}$$

3. UNIVARIATE MODELS

3.2. The $\kappa - \mu$ distribution

Remark 3.2 Let $U = \frac{W}{\sigma^2}$ then $W = \sigma^2 U$ and $\frac{dW}{dU} = \sigma^2$. The pdf of U is given by

$$f_U(u) = \frac{1}{2} \exp\left(-\frac{u + \frac{d^2}{\sigma^2}}{2}\right) \left(\frac{\sigma^2 u}{d^2}\right)^{\frac{n-1}{2}} I_{n-1}\left(\frac{d\sqrt{u}}{\sigma}\right).$$

This is the pdf of $\chi_{\frac{d^2}{\sigma^2}}^2(2n)$ with non-centrality parameter $\frac{d^2}{\sigma^2}$. Then the $E(U) = 2n + \frac{d^2}{\sigma^2}$ thus, $E(W) = 2n\sigma^2 + d^2$.

Re-parameterization of the pdf of the power (3.5) in terms of κ and μ . Let $\mu = n$ and define $\kappa = \frac{d^2}{2\mu\sigma^2}$, then $\frac{d}{\sqrt{2\sigma^2}} = \sqrt{\kappa\mu}$, and the pdf of the power becomes

$$\begin{aligned} f_W(w) &= \frac{1}{2\sigma^2} \left(\frac{w}{2\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \exp\left\{-\frac{(w + 2\sigma^2\kappa\mu)}{2\sigma^2}\right\} I_{\mu-1}\left(\frac{\sqrt{2\sigma^2\kappa\mu}\sqrt{w}}{\sigma^2}\right) \\ &= \frac{1}{2\sigma^2} \left(\frac{w}{2\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \exp\left\{-\frac{w}{2\sigma^2} - \kappa\mu\right\} I_{\mu-1}\left(\frac{\sqrt{2\sigma^2\kappa\mu}\sqrt{w}}{\sigma^2}\right), \quad w > 0. \end{aligned} \tag{3.7}$$

From (3.7) the $\kappa - \mu$ pdf of the power can be written as, with \bar{w} given by (see equation (3.6))

$$\bar{w} = E(W) = 2\sigma^2\mu(1 + \kappa), \tag{3.8}$$

$$\begin{aligned} \bar{w} f_W(w) &= \frac{\bar{w}}{2\sigma^2} \left(\frac{\bar{w} \frac{w}{\bar{w}}}{2\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \exp\left\{-\frac{\bar{w} \frac{w}{\bar{w}}}{2\sigma^2} - \kappa\mu\right\} I_{\mu-1}\left(\frac{\sqrt{2\sigma^2\kappa\mu}\sqrt{\bar{w} \frac{w}{\bar{w}}}}{\sigma^2}\right) \\ &= \frac{2\sigma^2\mu(1 + \kappa)}{2\sigma^2} \left(\frac{(2\sigma^2\mu(1 + \kappa)) \frac{w}{\bar{w}}}{2\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \exp\left\{-\frac{(2\sigma^2\mu(1 + \kappa)) \frac{w}{\bar{w}}}{2\sigma^2} - \kappa\mu\right\} \\ &\quad \times I_{\mu-1}\left(\frac{\sqrt{2\sigma^2\kappa\mu}\sqrt{(2\sigma^2\mu(1 + \kappa)) \frac{w}{\bar{w}}}}{\sigma^2}\right) \\ &= \frac{\mu(1 + \kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \left(\frac{w}{\bar{w}}\right)^{\frac{\mu-1}{2}} \exp\left\{-\frac{(\mu(1 + \kappa)) w}{\bar{w}} - \kappa\mu\right\} \\ &\quad \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1 + \kappa)} \frac{w}{\bar{w}}\right). \end{aligned} \tag{3.9}$$

For the normalized power $\Omega = \frac{W}{\bar{w}}$ (see Simon and Alouini, 2005 [29]), the pdf of Ω can be obtained from (3.9) as

$$\begin{aligned} f_\Omega(\omega) &= \frac{\mu(1 + \kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (\omega)^{\frac{\mu-1}{2}} \exp\{-\mu(1 + \kappa)\omega - \kappa\mu\} \\ &\quad \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1 + \kappa)}\omega\right). \end{aligned} \tag{3.10}$$

3. UNIVARIATE MODELS

3.2. The $\kappa - \mu$ distribution

The pdf of the envelope can be obtained from (3.5). Since $W = R^2$ we can express the pdf of the envelope as

$$\begin{aligned} f_R(r) &= \frac{2r}{2\sigma^2} \left(\frac{r^2}{d^2}\right)^{\frac{n-1}{2}} \exp\left\{-\frac{(r^2+d^2)}{2\sigma^2}\right\} I_{n-1}\left(\frac{dr}{\sigma^2}\right) \\ &= \frac{r^n}{\sigma^2 d^{n-1}} \exp\left\{-\frac{(r^2+d^2)}{2\sigma^2}\right\} I_{n-1}\left(\frac{d}{\sigma^2}r\right). \end{aligned} \quad (3.11)$$

Using (3.7), the pdf of the envelope can be expressed in terms of κ and μ as

$$\begin{aligned} f_R(r) &= \frac{2r}{2\sigma^2} \left(\frac{r^2}{2\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \exp\left\{-\frac{r^2}{2\sigma^2} - \kappa\mu\right\} I_{\mu-1}\left(\frac{\sqrt{2\sigma^2\kappa\mu}\sqrt{r^2}}{\sigma^2}\right) \\ &= \frac{1}{\sigma^2} \left(\frac{1}{2\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} r^\mu \exp\left\{-\frac{r^2}{2\sigma^2} - \kappa\mu\right\} I_{\mu-1}\left(\frac{r\sqrt{2\sigma^2\kappa\mu}}{\sigma^2}\right). \end{aligned} \quad (3.12)$$

Let

$$\hat{r} = \sqrt{E(R^2)} = \sqrt{E(W)} = \sqrt{2\sigma^2\mu(1+\kappa)}, \quad (3.13)$$

then from (3.12) it follows that

$$\begin{aligned} \hat{r}f_R(r) &= \frac{\hat{r}}{\sigma^2} \left(\frac{1}{2\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \left(\frac{r}{\hat{r}}\right)^\mu \exp\left\{-\frac{\left(\frac{r}{\hat{r}}\right)^2}{2\sigma^2} - \kappa\mu\right\} I_{\mu-1}\left(\frac{\left(\frac{r}{\hat{r}}\right)\sqrt{2\sigma^2\kappa\mu}}{\sigma^2}\right) \\ &= \frac{\left(\sqrt{2\sigma^2\mu(1+\kappa)}\right)^{\mu+1}}{\sigma^2} \left(\frac{1}{2\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \left(\frac{r}{\hat{r}}\right)^\mu \exp\left\{-\frac{\left(\sqrt{2\sigma^2\mu(1+\kappa)}\right)^2\left(\frac{r}{\hat{r}}\right)^2}{2\sigma^2} - \kappa\mu\right\} \\ &\quad \times I_{\mu-1}\left(\frac{\left(\frac{r}{\hat{r}}\right)\left(\sqrt{2\sigma^2\mu(1+\kappa)}\right)\sqrt{2\sigma^2\kappa\mu}}{\sigma^2}\right) \\ &= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \left(\frac{r}{\hat{r}}\right)^\mu \exp\left\{-\mu(1+\kappa)\left(\frac{r}{\hat{r}}\right)^2 - \kappa\mu\right\} \\ &\quad \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)}\left(\frac{r}{\hat{r}}\right)\right). \end{aligned} \quad (3.14)$$

For the normalized envelope $P = \frac{R}{\hat{r}}$, the pdf can be obtained from (3.14) and is given as

$$f_P(\rho) = \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \rho^\mu \exp\{-\mu\kappa - \mu(1+\kappa)\rho^2\} I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)}\rho\right), \quad (3.15)$$

where $\rho > 0$.

Remark 3.3 $\kappa = \frac{d^2}{2n\sigma^2}$ is defined as the ratio between the total power of the dominant components and the total power of the scattered waves (see Yacoub (2007) [35]).

3. UNIVARIATE MODELS

3.2. The $\kappa - \mu$ distribution

Remark 3.4 Let $\mu = \frac{1}{\text{var}(\rho^2)} \frac{1+2\kappa}{(1+\kappa)^2}$ with μ being the real extension of n . Thus, μ can take on non-integer values which may account for (a) non-zero correlation among the clusters of multipath components, (b) non-zero correlation between in-phase and quadrature components within each cluster, (c) the non-normal nature of the in-phase and quadrature components of each cluster of the fading signal. (See Yacoub (2007) [35]).

Remark 3.5 Equations (3.7), (3.9), (3.10), (3.14) and (3.15) correspond to equations (7), (10), (2), (11) and (1) of Yacoub (2007) [35].

Remark 3.6 Note that equation (3.10) can be approximated using the mixture gamma model which facilitates complicated or intractable performance analysis functions (see Atapattu (2011) [4]).

3.2.3 Characteristics

Statistical characteristics for the $\kappa - \mu$ fading model is now discussed in this Section.

3.2.3.1 Cumulative distribution function

Consider the $\kappa - \mu$ pdf for the normalized power (3.10), Ω . Using (2.13), the pdf can be written as

$$f_{\Omega}(\omega) = \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (\omega)^{\frac{\mu-1}{2}} \exp\{-\mu(1+\kappa)\omega - \kappa\mu\} \sum_{l=0}^{\infty} \left(\frac{2\mu\sqrt{\kappa(1+\kappa)}\omega}{2} \right)^{\mu-1} \times \frac{\left(\frac{1}{4} \left(2\mu\sqrt{\kappa(1+\kappa)}\omega \right)^2 \right)^l}{l! \Gamma(\mu+l)}.$$

The cdf of Ω with Z_T the specified threshold (see (1.2)), is

$$\begin{aligned} F(Z_T) &= \int_0^{Z_T} f_{\Omega}(\omega) d\omega \\ &= \int_0^{Z_T} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (\omega)^{\frac{\mu-1}{2}} \exp\{-\mu(1+\kappa)\omega - \kappa\mu\} \sum_{l=0}^{\infty} \left(\frac{2\mu\sqrt{\kappa(1+\kappa)}\omega}{2} \right)^{\mu-1} \\ &\quad \times \frac{\left(\frac{1}{4} \left(2\mu\sqrt{\kappa(1+\kappa)}\omega \right)^2 \right)^l}{l! \Gamma(\mu+l)} d\omega \end{aligned}$$

3. UNIVARIATE MODELS

3.2. The $\kappa - \mu$ distribution

$$\begin{aligned}
F(Z_T) &= \sum_{l=0}^{\infty} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times \int_0^{Z_T} \omega^{\mu-1} \exp\{-\mu(1+\kappa)\omega\} (\mu^2\kappa(1+\kappa)\omega)^l d\omega \\
&= \sum_{l=0}^{\infty} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} \mu^{2l}\kappa^l(1+\kappa)^l}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times \int_0^{Z_T} \omega^{\mu+l-1} \exp\{-\mu(1+\kappa)\omega\} d\omega. \tag{3.16}
\end{aligned}$$

Using (2.20), the cdf can be written as

$$\begin{aligned}
F(Z_T) &= \sum_{l=0}^{\infty} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} \mu^{2l}\kappa^l(1+\kappa)^l}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times [\mu(1+\kappa)]^{-(\mu+l-1)-1} \gamma(\mu+l-1+1, \mu(1+\kappa)Z_T) \\
&= \sum_{l=0}^{\infty} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} \mu^{2l}\kappa^l(1+\kappa)^l}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times [\mu(1+\kappa)]^{-\mu-l} \gamma(\mu+l, \mu(1+\kappa)Z_T) \\
&= \sum_{l=0}^{\infty} \frac{(\mu\kappa)^l \gamma(\mu+l, \mu(1+\kappa)Z_T)}{l! \Gamma(\mu+l) \exp(\mu\kappa)}, \tag{3.17}
\end{aligned}$$

for $\mu > 0$, index l where $\gamma(\cdot)$ is the incomplete gamma function (2.5).

Remark 3.7 *The outage probability can be calculated from (3.17) for a specified threshold Z_T .*

3.2.3.2 Moments

From (3.15) the j^{th} moment of the normalized envelope, P , is

$$\begin{aligned}
E(P^j) &= \int_0^{\infty} \rho^j f_P(\rho) d\rho \\
&= \int_0^{\infty} \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \rho^{\mu+j} \exp\{-\mu\kappa - \mu(1+\kappa)\rho^2\} I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)}\rho\right) d\rho.
\end{aligned}$$

3. UNIVARIATE MODELS

3.2. The $\kappa - \mu$ distribution

Using (2.22), the j^{th} moment is

$$\begin{aligned}
 E(P^j) &= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} 2^{-\mu+1-1} \left(2\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} (\mu(1+\kappa))^{-\frac{(\mu+j+1+\mu-1)}{2}} \Gamma\left[\frac{\mu+j+1+\mu-1}{2}\right] \\
 &\quad \times {}_1F_1\left(\frac{\mu+j+1+\mu-1}{2}; \mu-1+1; \frac{\left(2\mu\sqrt{\kappa(1+\kappa)}\right)^2}{4\mu(1+\kappa)}\right) \\
 &= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} 2^{-\mu} (2\mu)^{\mu-1} (\kappa(1+\kappa))^{\frac{\mu-1}{2}} (\mu(1+\kappa))^{-\mu-\frac{j}{2}} \frac{\Gamma\left(\mu+\frac{j}{2}\right)}{\Gamma(\mu)} \\
 &\quad \times {}_1F_1\left(\mu+\frac{j}{2}; \mu; \frac{4\mu^2\kappa(1+\kappa)}{4\mu(1+\kappa)}\right) \\
 &= \frac{\Gamma\left(\mu+\frac{j}{2}\right)}{[\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa)} {}_1F_1\left(\mu+\frac{j}{2}; \mu; \mu\kappa\right), \tag{3.18}
 \end{aligned}$$

where $\Gamma(\cdot)$ is the gamma function (see (2.1)) and ${}_1F_1(\cdot)$ is the confluent hypergeometric series (see (2.10)).

From (3.10) the j^{th} moment of the normalized power, Ω , is

$$\begin{aligned}
 E(\Omega^j) &= \int_0^\infty \omega^j f_\Omega(\omega) d\omega \\
 &= \int_0^\infty \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (\omega)^{\frac{\mu-1}{2}+j} \exp\{-\mu(1+\kappa)\omega - \kappa\mu\} \\
 &\quad \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)}\omega\right) d\omega.
 \end{aligned}$$

Let $\alpha = \sqrt{\omega}$, then

$$\begin{aligned}
 E(\Omega^j) &= \int_0^\infty \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (\alpha)^{\mu+1+2j-1} \exp\{-\mu(1+\kappa)\alpha^2 - \kappa\mu\} \\
 &\quad \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)}\alpha\right) d\alpha.
 \end{aligned}$$

Using (2.22), we obtain the j^{th} moment as

$$\begin{aligned}
 E(\Omega^j) &= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} 2^{-\mu+1-1} \left(2\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} (\mu(1+\kappa))^{-\frac{(\mu+1+2j+\mu-1)}{2}} \Gamma\left[\frac{\mu+1+2j+\mu-1}{2}\right] \\
 &\quad \times {}_1F_1\left(\frac{\mu+1+2j+\mu-1}{2}; \mu-1+1; \frac{\left(2\mu\sqrt{\kappa(1+\kappa)}\right)^2}{4\mu(1+\kappa)}\right)
 \end{aligned}$$

3. UNIVARIATE MODELS

3.2. The $\kappa - \mu$ distribution

$$\begin{aligned}
 E(\Omega^j) &= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} 2^{-\mu} (2\mu)^{\mu-1} (\kappa(1+\kappa))^{\frac{\mu-1}{2}} (\mu(1+\kappa))^{-\mu-j} \frac{\Gamma(\mu+j)}{\Gamma(\mu)} \\
 &\quad \times {}_1F_1\left(\frac{\mu+1+2j+\mu-1}{2}; \mu-1+1; \frac{4\mu^2\kappa(1+\kappa)}{4\mu(1+\kappa)}\right) \\
 &= \frac{\Gamma(\mu+j)}{\Gamma(\mu) \exp(\mu\kappa) [\mu(1+\kappa)]^j} {}_1F_1(\mu+j; \mu; \mu\kappa). \tag{3.19}
 \end{aligned}$$

3.2.3.3 Amount of fading

The amount of fading for the $\kappa - \mu$ model is given as (see (1.1))

$$\begin{aligned}
 AF &= \frac{E(\Omega^2)}{E(\Omega)^2} - 1 \\
 &= \frac{\left[\frac{\Gamma(\mu+2)}{\Gamma(\mu) \exp\{\mu\kappa\} [\mu(1+\kappa)]^2} {}_1F_1(\mu+2; \mu; \mu\kappa) \right]}{\left[\frac{\Gamma(\mu+1)}{\Gamma(\mu) \exp\{\mu\kappa\} [\mu(1+\kappa)]} {}_1F_1(\mu+1; \mu; \mu\kappa) \right]^2} - 1 \\
 &= \left(\frac{\Gamma(\mu+2) (2\Gamma(\mu) \exp\{\mu\kappa\} [\mu(1+\kappa)])^2} {\Gamma(\mu) \exp\{\mu\kappa\} [\mu(1+\kappa)]^2 (\Gamma(\mu+1))^2} \frac{{}_1F_1(\mu+2; \mu; \mu\kappa)}{({}_1F_1(\mu+1; \mu; \mu\kappa))^2} \right) - 1 \\
 &= \left(\frac{\Gamma(\mu+2) \Gamma(\mu) \exp\{\mu\kappa\}}{(\Gamma(\mu+1))^2} \frac{{}_1F_1(\mu+2; \mu; \mu\kappa)}{({}_1F_1(\mu+1; \mu; \mu\kappa))^2} \right) - 1 \\
 &= \left(\frac{(\mu+1) \exp\{\mu\kappa\}}{\mu} \frac{{}_1F_1(\mu+2; \mu; \mu\kappa)}{({}_1F_1(\mu+1; \mu; \mu\kappa))^2} \right) - 1. \tag{3.20}
 \end{aligned}$$

3.2.4 Special cases

3.2.4.1 The $\kappa - \mu$ distribution for $\kappa \rightarrow 0$ (Nakagami - m)

From (3.15) and (2.14) it follows

$$\begin{aligned}
 f_P(\rho) &= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}} \exp\{\mu\kappa\}} \rho^\mu \exp\{-\mu(1+\kappa)\rho^2\} \frac{\left(\mu\sqrt{\kappa(1+\kappa)}\rho\right)^{\mu-1}}{\Gamma(\mu)} \\
 &= \frac{2\mu^\mu(1+\kappa)^\mu}{\Gamma(\mu) \exp\{\mu\kappa\}} \rho^{2\mu-1} \exp\{-\mu(1+\kappa)\rho^2\}. \tag{3.21}
 \end{aligned}$$

As $\kappa \rightarrow 0$, the pdf (3.21) tends to

$$f_P(\rho) = \frac{2\mu^\mu}{\Gamma(\mu)} \rho^{2\mu-1} \exp\{-\mu\rho^2\}, \quad \rho > 0,$$

which has the form of a Nakagami- m distribution (see (2.32) with $\mu = m$ and $\Omega = 1$).

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

Table 1 gives other special cases of the $\kappa - \mu$ model

Distribution	Parameters
Rice	$\mu = 1$
Rayleigh	$\mu = 1, \kappa = 0$
One-sided Gaussian	$\mu = 0.5, \kappa \rightarrow 0$
Table 1: Special cases	

3.3 The $\kappa - \mu$ type distribution

3.3.1 Description

Let X_i and Y_i be mutually independent elliptical distributed with $E(X_i) = E(Y_i) = 0$ and $var(X_i) = var(Y_i) = -2\sigma^2\Psi'(0)$. Then the envelope, R , of the physical model for the $\kappa - \mu$ distribution can be written in terms of the in-phase and quadrature components of the fading signal as

$$W = R^2 = \sum_{i=1}^n (X_i + p_i)^2 + \sum_{i=1}^n (Y_i + q_i)^2, \quad (3.22)$$

where p_i and q_i are the mean values of the in-phase and quadrature components of the multipath waves of cluster i and n is the number of clusters of multipath.

Since $X_i \sim E(0, \sigma^2)$ and $Y_i \sim E(0, \sigma^2)$, therefore from (2.28), $X_i|t \sim N(0, a_t)$ and $Y_i|t \sim N(0, a_t)$ where $a_t = t^{-1}\sigma^2$. Hence, $(X_i + p_i)|t \sim N(p_i, a_t)$ and $X_i^*|t = \left(\frac{X_i + p_i}{\sqrt{a_t}}|t\right) \sim N\left(\frac{p_i}{\sqrt{a_t}}, 1\right)$. Thus $(X_i^*)^2|t$ follows a $\chi_{\frac{p_i^2}{a_t}}^2(1)$ with non-centrality parameter $\frac{p_i^2}{a_t}$ and similarly we can express $Y_i^*|t = \left(\frac{Y_i + q_i}{\sqrt{a_t}}|t\right) \sim N\left(\frac{q_i}{\sqrt{a_t}}, 1\right)$ and $(Y_i^*)^2|t$ follows a $\chi_{\frac{q_i^2}{a_t}}^2(1)$ with non-centrality parameter $\frac{q_i^2}{a_t}$.

Let $\frac{R_i^2|t}{a_t} = \frac{W_i|t}{a_t} = (X_i^*)^2|t + (Y_i^*)^2|t$ then $\frac{R_i^2|t}{a_t}$ follows a $\chi_{\frac{d_i^2}{a_t}}^2(2)$ with non-centrality parameter $\frac{d_i^2}{a_t} = \frac{p_i^2 + q_i^2}{a_t}$.

For $R^2|t = \sum_{i=1}^n R_i^2|t$, the power of the fading signal is $W|t = \sum_{i=1}^n W_i|t$, and we are interested in the pdf of the envelope R and the power W respectively.

The following is a schematic outline of the mathematical algorithm that will be followed to obtain the necessary distributions.

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

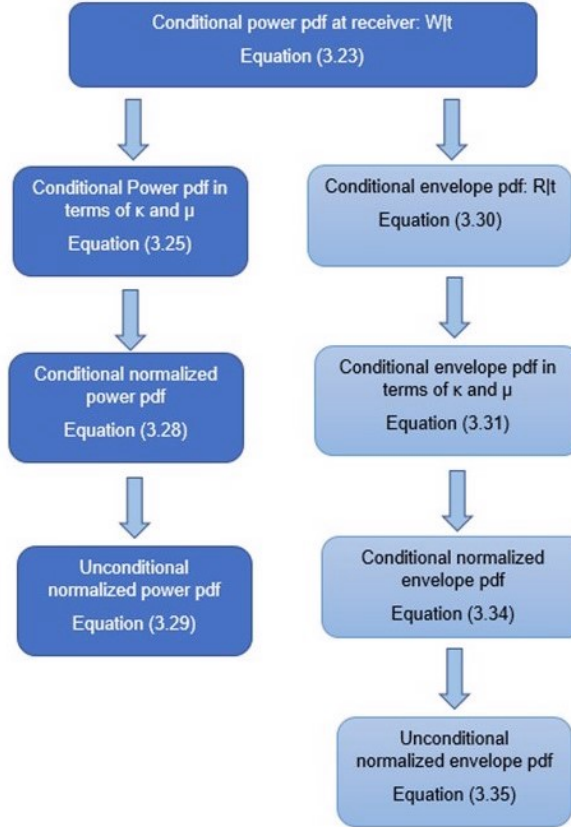


Figure 3.3

3.3.2 Derivation

The pdf of $\left(\frac{W_i|t}{a_t}\right)$ is a $\chi_{\frac{d_i^2}{a_t}}^2$ (2) distribution with non-centrality parameter $\frac{d_i^2}{a_t}$. (Note for the elliptical case for simplicity, $f_{W_i|t}(w_i|t) \equiv f(w_i|t)$). The conditional pdf of W_i is then

$$f(w_i|t) = \frac{1}{2a_t} \exp\left\{-\frac{(w_i + d_i^2)}{2a_t}\right\} I_0\left(\frac{d_i\sqrt{w_i}}{a_t}\right).$$

Using the Laplace transform method as in Section 3.2, the conditional pdf of the power, W , is

$$f(w|t) = \frac{1}{2a_t} \left(\frac{w}{d^2}\right)^{\frac{n-1}{2}} \exp\left\{-\frac{(w + d^2)}{2a_t}\right\} I_{n-1}\left(\frac{d\sqrt{w}}{a_t}\right), \quad w > 0. \quad (3.23)$$

Since $\frac{W|t}{a_t}$ follows a $\chi_{\frac{d^2}{a_t}}^2$ (n) distribution with non-centrality parameter $\frac{d^2}{a_t}$, the expected value of $W|t$ is given by (2.31) as

$$E(W|t) = 2na_t + d^2. \quad (3.24)$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

Re-parameterization of the conditional pdf of the power in terms of κ and μ and let $\mu = n$ with $\kappa = \frac{d^2}{2\mu t^{-1}\sigma^2}$, then $\frac{d}{\sqrt{2t^{-1}\sigma^2}} = \sqrt{\kappa\mu}$. The conditional pdf of the power becomes

$$\begin{aligned} f(w|t) &= \frac{1}{2t^{-1}\sigma^2} \left(\frac{w}{2t^{-1}\sigma^2\kappa\mu} \right)^{\frac{\mu-1}{2}} \exp \left\{ -\frac{(w + 2t^{-1}\sigma^2\kappa\mu)}{2t^{-1}\sigma^2} \right\} I_{\mu-1} \left(\frac{\sqrt{2t^{-1}\sigma^2\kappa\mu}\sqrt{w}}{t^{-1}\sigma^2} \right) \\ &= \frac{1}{2t^{-1}\sigma^2} \left(\frac{w}{2t^{-1}\sigma^2\kappa\mu} \right)^{\frac{\mu-1}{2}} \exp \left\{ -\frac{w}{2t^{-1}\sigma^2} - \kappa\mu \right\} I_{\mu-1} \left(\frac{\sqrt{2t^{-1}\sigma^2\kappa\mu}\sqrt{w}}{t^{-1}\sigma^2} \right) \end{aligned} \quad (3.25)$$

From (3.24) and (3.25) the conditional pdf of the power can be written as follows, with (see equation (3.24))

$$\bar{w}_t \equiv E(W|t) = 2a_t\mu(1 + \kappa) = 2t^{-1}\sigma^2\mu(1 + \kappa) = t^{-1}\omega, \quad (3.26)$$

$$\begin{aligned} \bar{w}_t f(w|t) &= \frac{\bar{w}_t}{2t^{-1}\sigma^2} \left(\frac{\bar{w}_t \frac{w}{\bar{w}_t}}{2t^{-1}\sigma^2\kappa\mu} \right)^{\frac{\mu-1}{2}} \exp \left\{ -\frac{\bar{w}_t \frac{w}{\bar{w}_t}}{2t^{-1}\sigma^2} - \kappa\mu \right\} I_{\mu-1} \left(\frac{\sqrt{2t^{-1}\sigma^2\kappa\mu}\sqrt{\bar{w}_t \frac{w}{\bar{w}_t}}}{t^{-1}\sigma^2} \right) \\ &= \frac{2t^{-1}\sigma^2\mu(1 + \kappa)}{2t^{-1}\sigma^2} \left(\frac{(2t^{-1}\sigma^2\mu(1 + \kappa)) \frac{w}{\bar{w}_t}}{2t^{-1}\sigma^2\kappa\mu} \right)^{\frac{\mu-1}{2}} \exp \left\{ -\frac{(2t^{-1}\sigma^2\mu(1 + \kappa)) \frac{w}{\bar{w}_t}}{2t^{-1}\sigma^2} - \kappa\mu \right\} \\ &\quad \times I_{\mu-1} \left(\frac{\sqrt{2t^{-1}\sigma^2\kappa\mu}\sqrt{(2t^{-1}\sigma^2\mu(1 + \kappa)) \frac{w}{\bar{w}_t}}}{t^{-1}\sigma^2} \right) \\ &= \frac{\mu(1 + \kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \left(\frac{w}{\bar{w}_t} \right)^{\frac{\mu-1}{2}} \exp \left\{ -\frac{(\mu(1 + \kappa))w}{\bar{w}_t} - \kappa\mu \right\} \\ &\quad \times I_{\mu-1} \left(2\mu\sqrt{\kappa(1 + \kappa)} \frac{w}{\bar{w}_t} \right). \end{aligned} \quad (3.27)$$

For the pdf of the normalized power $\Omega_t \equiv \Omega|t = \frac{W|t}{\bar{w}_t} = \frac{\omega_t}{\bar{w}_t}$, the pdf can be obtained from (3.27) as

$$\begin{aligned} f_{\Omega_t}(\omega_t) &= \frac{\mu(1 + \kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (\omega_t)^{\frac{\mu-1}{2}} \exp \{ -\mu(1 + \kappa)\omega_t - \kappa\mu \} \\ &\quad \times I_{\mu-1} \left(2\mu\sqrt{\kappa(1 + \kappa)}\omega_t \right). \end{aligned} \quad (3.28)$$

Hence the unconditional pdf of the normalized power, $\Omega = \frac{W}{t^{-1}\bar{w}} = t\omega$ (see (3.26)), is given by

$$f_{\Omega}(\omega) = \int_0^{\infty} W(t) f_{\Omega_t}(\omega_t) dt$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

$$f_{\Omega}(\omega) = \int_0^{\infty} W(t) \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (t\omega)^{\frac{\mu-1}{2}} \exp\{-\mu(1+\kappa)t\omega - \kappa\mu\} \\ \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)t\omega}\right) dt. \quad (3.29)$$

The conditional pdf of the envelope can be obtained from (3.23). Since $W = R^2$, we can express the conditional pdf of the envelope as

$$f(r|t) = \frac{1}{2a_t} \left(\frac{r^2}{d^2}\right)^{\frac{n-1}{2}} \exp\left\{-\frac{(r^2+d^2)}{2a_t}\right\} I_{n-1}\left(\frac{dr}{a_t}\right). \quad (3.30)$$

Using (3.25), the conditional pdf of the envelope can be expressed in terms of κ and μ , as follows

$$f(r|t) = \frac{2r}{2t^{-1}\sigma^2} \left(\frac{r^2}{2t^{-1}\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \exp\left\{-\frac{r^2}{2t^{-1}\sigma^2} - \kappa\mu\right\} I_{\mu-1}\left(\frac{\sqrt{2t^{-1}\sigma^2\kappa\mu}\sqrt{r^2}}{t^{-1}\sigma^2}\right) \\ = \frac{1}{t^{-1}\sigma^2} \left(\frac{1}{2t^{-1}\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} r^{\mu} \exp\left\{-\frac{r^2}{2t^{-1}\sigma^2} - \kappa\mu\right\} \\ \times I_{\mu-1}\left(\frac{r\sqrt{2t\kappa\mu}}{\sigma}\right). \quad (3.31)$$

The conditional $\kappa - \mu$ pdf of the envelope is then obtained from (3.31), with (see (3.13) and (3.26))

$$\hat{r}_t \equiv \sqrt{E(R^2|t)} = \sqrt{E(W|t)} = \left(\sqrt{2t^{-1}\sigma^2\mu(1+\kappa)}\right), \quad (3.32)$$

from equation (3.26), as follows

$$\hat{r}_t f(r|t) = \frac{\hat{r}_t}{t^{-1}\sigma^2} \left(\frac{1}{2t^{-1}\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \left(\frac{r}{\hat{r}_t}\right)^{\mu} \exp\left\{-\frac{\left(\frac{r}{\hat{r}_t}\right)^2}{2t^{-1}\sigma^2} - \kappa\mu\right\} \\ \times I_{\mu-1}\left(\frac{\left(\frac{r}{\hat{r}_t}\right)\sqrt{2t\kappa\mu}}{\sigma}\right)$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

$$\begin{aligned}
&= \frac{\left(\sqrt{2t^{-1}\sigma^2\mu(1+\kappa)}\right)^{\mu+1}}{t^{-1}\sigma^2} \left(\frac{1}{2t^{-1}\sigma^2\kappa\mu}\right)^{\frac{\mu-1}{2}} \left(\frac{r}{\hat{r}_t}\right)^\mu \\
&\times \exp \left\{ -\frac{\left(\sqrt{2t^{-1}\sigma^2\mu(1+\kappa)}\right)^2 \left(\frac{r}{\hat{r}_t}\right)^2}{2t^{-1}\sigma^2} - \kappa\mu \right\} \\
&\times I_{\mu-1} \left(\frac{\left(\frac{r}{\hat{r}_t}\right) \left(\sqrt{2t^{-1}\sigma^2\mu(1+\kappa)}\right) \sqrt{2t\kappa\mu}}{\sigma} \right) \\
&= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \left(\frac{r}{\hat{r}_t}\right)^\mu \exp \left\{ -\mu(1+\kappa) \left(\frac{r}{\hat{r}_t}\right)^2 - \kappa\mu \right\} \\
&\times I_{\mu-1} \left(2\mu\sqrt{\kappa(1+\kappa)} \left(\frac{r}{\hat{r}_t}\right) \right). \tag{3.33}
\end{aligned}$$

From (3.33) the pdf for the normalized envelope $P_t = P|t = \frac{R|t}{\hat{r}_t}$, is

$$f_{P_t}(\rho_t) = \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \rho_t^\mu \exp \left\{ -\mu\kappa - \mu(1+\kappa)\rho_t^2 \right\} I_{\mu-1} \left(2\mu\sqrt{\kappa(1+\kappa)}\rho_t \right). \tag{3.34}$$

As before the unconditional pdf of the normalized envelope is given by

$$\begin{aligned}
f_P(\rho) &= \int_0^\infty W(t) \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (t\rho)^\mu \exp \left\{ -\mu\kappa - \mu(1+\kappa)(t\rho)^2 \right\} \\
&\times I_{\mu-1} \left(2\mu\sqrt{\kappa(1+\kappa)}t\rho \right) dt, \tag{3.35}
\end{aligned}$$

where $\rho > 0$.

Remark 3.8 For the normal distribution the weight function is (2.35). The unconditional pdf of the normalized envelope follows from (3.35) as

$$\begin{aligned}
f_P(\rho) &= \int_0^\infty \delta(t-1) \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (t\rho)^\mu \exp \left\{ -\mu\kappa - \mu(1+\kappa)(t\rho)^2 \right\} \\
&\times I_{\mu-1} \left(2\mu\sqrt{\kappa(1+\kappa)}t\rho \right) dt. \\
&= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \rho^\mu \exp \left\{ -\mu\kappa - \mu(1+\kappa)\rho^2 \right\} \\
&\times I_{\mu-1} \left(2\mu\sqrt{\kappa(1+\kappa)}\rho \right). \tag{3.36}
\end{aligned}$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

The unconditional pdf of the normalized power follows from (3.29)

$$\begin{aligned}
 f_{\Omega}(\omega) &= \int_0^{\infty} \delta(t-1) \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (t\omega)^{\frac{\mu-1}{2}} \exp\{-\mu(1+\kappa)t\omega - \kappa\mu\} \\
 &\quad \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)t\omega}\right) dt. \\
 &= \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} \omega^{\frac{\mu-1}{2}} \exp\{-\mu(1+\kappa)\omega - \kappa\mu\} \\
 &\quad \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)\omega}\right) \\
 &= (3.10)
 \end{aligned} \tag{3.37}$$

Remark 3.9 For the t -distribution the weight function is (2.36). The unconditional pdf of the normalized envelope follows from (3.35) as

$$\begin{aligned}
 f_P(\rho) &= \int_0^{\infty} \frac{v\left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right) 2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{2\Gamma\left(\frac{v}{2}\right) \kappa^{\frac{\mu-1}{2}}} (t\rho)^{\mu} \\
 &\quad \times \exp\{-\mu\kappa - \mu(1+\kappa)(t\rho)^2\} I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)t\rho}\right) dt \\
 &= \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} \rho^{\mu} \left(\frac{v}{2}\right)^{\frac{v}{2}} \frac{1}{2\Gamma\left(\frac{v}{2}\right)} \int_0^{\infty} t^{\mu+\frac{v}{2}-1} \\
 &\quad \times \exp\left\{-\mu(1+\kappa)t^2\rho^2 - \frac{vt}{2}\right\} I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)t\rho}\right) dt \\
 &= \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\frac{v}{2}\right)^{\frac{v}{2}} \rho^{\mu}}{\kappa^{\frac{\mu-1}{2}} \Gamma\left(\frac{v}{2}\right) \exp(\mu\kappa)} \sum_{z=0}^{\infty} \frac{(-1)^z}{z!} \left(\frac{v}{2}\right)^z \int_0^{\infty} t^{\mu+\frac{v}{2}+z-1} \\
 &\quad \times \exp\{-\mu(1+\kappa)t^2\rho^2\} I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)t\rho}\right) dt.
 \end{aligned}$$

Using (2.22), we get

$$\begin{aligned}
 f_P(\rho) &= \sum_{z=0}^{\infty} \frac{(-1)^z \mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\frac{v}{2}\right)^{\frac{v}{2}+z} \rho^{\mu}}{z! \kappa^{\frac{\mu-1}{2}} \Gamma\left(\frac{v}{2}\right) \exp(\mu\kappa)} 2^{-\mu+1-1} \left[2\mu\sqrt{\kappa(1+\kappa)}\rho\right]^{\mu-1} \\
 &\quad \times \left[\mu(1+\kappa)\rho^2\right]^{-\frac{1}{2}(\mu+\frac{v}{2}+z+\mu-1)} \Gamma\left[\frac{\frac{1}{2}(\mu+\frac{v}{2}+z+\mu-1)}{\mu-1+1}\right] \\
 &\quad \times {}_1F_1\left(\frac{1}{2}\left(\mu+\frac{v}{2}+z+\mu-1\right); \mu-1+1; \frac{4\mu^2\kappa(1+\kappa)\rho^2}{4\mu(1+\kappa)\rho^2}\right)
 \end{aligned}$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

$$\begin{aligned}
 f_P(\rho) &= \sum_{z=0}^{\infty} \frac{(-1)^z \kappa^{\frac{\mu}{2}-\frac{1}{2}} (1+\kappa)^{-\frac{1}{2}(\frac{v}{2}+z-1)} \mu^{\frac{v}{2}} v^z \rho^{-\frac{v}{2}-z}}{2^{\frac{v}{2}+z+1} z! \Gamma\left(\frac{v}{2}\right) \exp(\mu\kappa)} \\
 &\quad \times \Gamma\left[\mu + \frac{v}{4} + \frac{z}{2} - \frac{1}{2}\right] {}_1F_1\left(\mu + \frac{v}{4} + \frac{z}{2} - \frac{1}{2}; \mu; \mu\kappa\right), \quad (3.38)
 \end{aligned}$$

where $\rho > 0$.

The unconditional pdf of the normalized power follows from (3.29) as

$$\begin{aligned}
 f_{\Omega}(\omega) &= \int_0^{\infty} \frac{v \left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right) \mu (1+\kappa)^{\frac{\mu+1}{2}}}{2\Gamma\left(\frac{v}{2}\right) \kappa^{\frac{\mu-1}{2}}} (t\omega)^{\frac{\mu-1}{2}} \\
 &\quad \times \exp\{-\mu(1+\kappa)t\omega - \kappa\mu\} I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)t\omega}\right) dt \\
 &= \int_0^{\infty} \frac{\mu (1+\kappa)^{\frac{\mu+1}{2}} v^{\frac{v}{2}} \omega^{\frac{\mu-1}{2}}}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} t^{\frac{v}{2}-1+\frac{\mu}{2}-\frac{1}{2}} \\
 &\quad \times \exp\left\{-\mu(1+\kappa)t\omega - \frac{vt}{2}\right\} I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)t\omega}\right) dt.
 \end{aligned}$$

Let $q = \sqrt{t}$ then $dt = 2qdq$. The pdf can be written as

$$\begin{aligned}
 f_{\Omega}(\omega) &= \int_0^{\infty} \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}} v^{\frac{v}{2}} \omega^{\frac{\mu-1}{2}}}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} q^{v+\mu-2} \\
 &\quad \times \exp\left\{-q^2\left[\mu(1+\kappa)\omega - \frac{v}{2}\right]\right\} I_{\mu-1}\left(q2\mu\sqrt{\kappa(1+\kappa)\omega}\right) dq.
 \end{aligned}$$

Using (2.22), we get

$$\begin{aligned}
 f_{\Omega}(\omega) &= \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} v^{\frac{v}{2}} \omega^{\frac{\mu-1}{2}}}{2^{\frac{v}{2}-1} \Gamma\left(\frac{v}{2}\right) \kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} 2^{-\mu+1-1} \left[2\mu\sqrt{\kappa(1+\kappa)\omega}\right]^{\mu-1} \\
 &\quad \times \left[\mu(1+\kappa)\omega - \frac{v}{2}\right]^{-\frac{(v+\mu-1+\mu-1)}{2}} \Gamma\left[\frac{(v+\mu-1+\mu-1)}{2}\right] \\
 &\quad \times {}_1F_1\left(\frac{(v+\mu-1+\mu-1)}{2}; \mu-1+1; \frac{\left[2\mu\sqrt{\kappa(1+\kappa)\omega}\right]^2}{4\left[\mu(1+\kappa)\omega - \frac{v}{2}\right]}\right)
 \end{aligned}$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

$$\begin{aligned}
 &= \frac{\mu^\mu (1 + \kappa)^\mu v^{\frac{v}{2}} \omega^{\mu-1}}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \exp(\mu\kappa)} \left[\mu(1 + \kappa)\omega - \frac{v}{2} \right]^{-\left(\frac{v}{2} + \mu - 1\right)} \Gamma\left[\begin{matrix} \frac{v}{2} + \mu - 1 \\ \mu \end{matrix} \right] \\
 &\times {}_1F_1\left(\frac{v}{2} + \mu - 1; \mu; \frac{\mu^2 \kappa (1 + \kappa) \omega}{\left[\mu(1 + \kappa)\omega - \frac{v}{2} \right]} \right), \quad \omega > 0.
 \end{aligned} \tag{3.39}$$

In Table 2 below, particular cases of the pdf's are focused on, since they form part of the investigation in section 3.4.

Weight function $W(t)$	normalized pdf
normal weight (2.35)	Envelope equation (3.36) Power equation (3.37)
t-distribution weight (2.36)	Envelope equation (3.38) Power equation (3.39)

Table 2: Particular cases of (3.29) and (3.35)

3.3.3 Characteristics

Subsequently the statistical characteristics for the $\kappa - \mu$ type fading model is derived.

3.3.3.1 Cumulative distribution function

Consider the $\kappa - \mu$ type pdf for normalized power, Ω (3.29), and using (2.13) it follows that

$$\begin{aligned}
 f_\Omega(\omega) &= \int_0^\infty W(t) \frac{\mu(1 + \kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (t\omega)^{\frac{\mu-1}{2}} \exp\{-\mu(1 + \kappa)t\omega - \kappa\mu\} \sum_{l=0}^\infty \left(\frac{2\mu\sqrt{\kappa(1 + \kappa)t\omega}}{2} \right)^{\mu-1} \\
 &\times \frac{\left(\frac{1}{4} \left(2\mu\sqrt{\kappa(1 + \kappa)t\omega} \right)^2 \right)^l}{l! \Gamma(\mu + l)} dt.
 \end{aligned}$$

The cdf of Ω , is given by

$$\begin{aligned}
 F(Z_T) &= \int_0^{Z_T} f_\Omega(\omega) d\omega \\
 &= \int_0^{Z_T} \int_0^\infty W(t) \frac{\mu(1 + \kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (t\omega)^{\frac{\mu-1}{2}} \exp\{-\mu(1 + \kappa)t\omega - \kappa\mu\} \sum_{l=0}^\infty \left(\frac{2\mu\sqrt{\kappa(1 + \kappa)t\omega}}{2} \right)^{\mu-1} \\
 &\times \frac{\left(\frac{1}{4} \left(2\mu\sqrt{\kappa(1 + \kappa)t\omega} \right)^2 \right)^l}{l! \Gamma(\mu + l)} dt d\omega
 \end{aligned}$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

$$\begin{aligned}
&= \int_0^{Z_T} \int_0^\infty \sum_{l=0}^\infty \frac{W(t) \mu (1 + \kappa)^{\frac{\mu+1}{2}} (t)^{\frac{\mu-1}{2}} \left(\mu \sqrt{t\kappa(1+\kappa)} \right)^{\mu-1}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times \omega^{\mu-1} \exp\{-\mu(1+\kappa)t\omega\} (\mu^2 \kappa (1 + \kappa) t\omega)^l dt d\omega \\
&= \int_0^{Z_T} \int_0^\infty \sum_{l=0}^\infty \frac{W(t) \mu (1 + \kappa)^{\frac{\mu+1}{2}} \left(\mu \sqrt{\kappa(1+\kappa)} \right)^{\mu-1} \mu^{2l} \kappa^l (1 + \kappa)^l (t)^{\mu+l-1}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times \omega^{\mu+l-1} \exp\{-\mu(1+\kappa)t\omega\} dt d\omega, \tag{3.40}
\end{aligned}$$

where Z_T is the specified threshold (see 1.2).

Subsequently special cases (normal distribution and t-distribution) for the outage probability will be derived.

Remark 3.10 For the normal distribution the weight function is (2.35). The outage probability follows from (3.40) as

$$\begin{aligned}
F(Z_T) &= \int_0^{Z_T} \int_0^\infty \sum_{l=0}^\infty \frac{\delta(t-1) \mu (1 + \kappa)^{\frac{\mu+1}{2}} \left(\mu \sqrt{\kappa(1+\kappa)} \right)^{\mu-1} \mu^{2l} \kappa^l (1 + \kappa)^l (t)^{\mu+l-1}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times \omega^{\mu+l-1} \exp\{-\mu(1+\kappa)t\omega\} dt d\omega.
\end{aligned}$$

Consider the result,

$$\begin{aligned}
F(Z_T, t) &= \int_0^\infty \delta(t) F(Z_T, t+1) dt \\
&= F(Z_T, 1).
\end{aligned}$$

Let $q = t - 1$ then $t = q + 1$, the outage probability is

$$\begin{aligned}
F(Z_T, t) &= F(Z_T, q+1) \\
&= \int_0^{Z_T} \int_0^\infty \sum_{l=0}^\infty \frac{\delta(q) \mu (1 + \kappa)^{\frac{\mu+1}{2}} \left(\mu \sqrt{\kappa(1+\kappa)} \right)^{\mu-1} \mu^{2l} \kappa^l (1 + \kappa)^l (q+1)^{\mu+l-1}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times \omega^{\mu+l-1} \exp\{-\mu(1+\kappa)t\omega\} dt d\omega \\
&= F(Z_T, 1).
\end{aligned}$$

thus,

$$\begin{aligned}
F(Z_T) &= \int_0^{Z_T} \sum_{l=0}^\infty \frac{\mu (1 + \kappa)^{\frac{\mu+1}{2}} \left(\mu \sqrt{\kappa(1+\kappa)} \right)^{\mu-1} \mu^{2l} \kappa^l (1 + \kappa)^l}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times \omega^{\mu+l-1} \exp\{-\mu(1+\kappa)t\omega\} d\omega.
\end{aligned}$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

Using (2.20) the outage is

$$\begin{aligned}
F(Z_T) &= \sum_{l=0}^{\infty} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} \mu^{2l} \kappa^l (1+\kappa)^l}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times [\mu(1+\kappa)]^{-(\mu+l-1)-1} \gamma(\mu+l-1+1, \mu(1+\kappa) Z_T) \\
&= \sum_{l=0}^{\infty} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} \mu^{2l} \kappa^l (1+\kappa)^l}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times [\mu(1+\kappa)]^{-\mu-l} \gamma(\mu+l, \mu(1+\kappa) Z_T) \\
&= \sum_{l=0}^{\infty} \frac{(\mu\kappa)^l \gamma(\mu+l, \mu(1+\kappa) Z_T)}{l! \Gamma(\mu+l) \exp(\mu\kappa)} \\
&= (3.17).
\end{aligned} \tag{3.41}$$

Remark 3.11 For the t -distribution the weight function is (2.36). The outage probability follows from (3.40) as

$$\begin{aligned}
F(Z_T) &= \int_0^{Z_T} \int_0^{\infty} \sum_{l=0}^{\infty} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} \mu^{2l} \kappa^l (1+\kappa)^l (t)^{\mu+l-1}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times \frac{v \left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left\{-\frac{vt}{2}\right\}}{2\Gamma\left(\frac{v}{2}\right)} \omega^{\mu+l-1} \exp\{-\mu(1+\kappa)t\omega\} dt d\omega \\
&= \int_0^{Z_T} \int_0^{\infty} \sum_{l=0}^{\infty} \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} \left(\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} \mu^{2l} \kappa^l (1+\kappa)^l v \left(\frac{v}{2}\right)^{\frac{v}{2}-1}}{2\Gamma\left(\frac{v}{2}\right) \kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa) l! \Gamma(\mu+l)} \\
&\quad \times (t)^{\mu+l+\frac{v}{2}-2} \exp\left\{-\frac{vt}{2}\right\} \omega^{\mu+l-1} \exp\{-\mu(1+\kappa)t\omega\} dt d\omega \\
&= \sum_{l=0}^{\infty} \frac{\mu^{\mu+2l} \kappa^l (1+\kappa)^{\mu+l} v \left(\frac{v}{2}\right)^{\frac{v}{2}-1}}{2\Gamma\left(\frac{v}{2}\right) \exp(\mu\kappa) l! \Gamma(\mu+l)} \int_0^{Z_T} \omega^{\mu+l-1} \\
&\quad \times \int_0^{\infty} (t)^{\mu+l+\frac{v}{2}-2} \exp\left\{-t\left[\mu(1+\kappa)\omega + \frac{v}{2}\right]\right\} dt d\omega.
\end{aligned}$$

Using (2.19),

$$\begin{aligned}
F(Z_T) &= \sum_{l=0}^{\infty} \frac{\mu^{\mu+2l} \kappa^l (1+\kappa)^{\mu+l} v \left(\frac{v}{2}\right)^{\frac{v}{2}-1}}{2\Gamma\left(\frac{v}{2}\right) \exp(\mu\kappa) l! \Gamma(\mu+l)} \int_0^{Z_T} \omega^{\mu+l-1} \\
&\quad \times \frac{\Gamma\left(\mu+l+\frac{v}{2}-1\right)}{\left[\mu(1+\kappa)\omega + \frac{v}{2}\right]^{\mu+l+\frac{v}{2}-1}}
\end{aligned}$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

$$\begin{aligned}
&= \sum_{l=0}^{\infty} \frac{\mu^{\mu+2l} \kappa^l (1+\kappa)^{\mu+l} v^{\frac{v}{2}} \Gamma\left(\mu+l+\frac{v}{2}-1\right)}{2^{\frac{v}{2}} l! \Gamma(\mu+l) \Gamma\left(\frac{v}{2}\right) \exp(\mu\kappa)} \\
&\quad \times \int_0^{Z_T} \frac{\omega^{\mu+l-1}}{\left[\mu(1+\kappa)\omega + \frac{v}{2}\right]^{\mu+l+\frac{v}{2}-1}} d\omega \\
&= \sum_{l=0}^{\infty} \frac{\mu^{\mu+2l} \kappa^l (1+\kappa)^{\mu+l} v^{\frac{v}{2}}}{2^{\frac{v}{2}} l! B\left(\frac{v}{2}, \mu+l\right) (\mu+l+\frac{v}{2}-1) \exp(\mu\kappa)} \\
&\quad \times \int_0^{Z_T} \frac{\omega^{\mu+l-1}}{\left[\frac{v}{2} \left\{1 + \frac{2}{v} \mu(1+\kappa)\omega\right\}\right]^{\mu+l+\frac{v}{2}-1}} d\omega \\
&= \sum_{l=0}^{\infty} \frac{\mu^{\mu+2l} \kappa^l (1+\kappa)^{\mu+l} v^{\frac{v}{2}}}{2^{\frac{v}{2}} l! B\left(\frac{v}{2}, \mu+l\right) (\mu+l+\frac{v}{2}-1) \exp(\mu\kappa)} \\
&\quad \times \left(\frac{v}{2}\right)^{-(\mu+l+\frac{v}{2}-1)} \int_0^{Z_T} \frac{\omega^{\mu+l-1}}{\left[1 + \frac{2}{v} \mu(1+\kappa)\omega\right]^{\mu+l+\frac{v}{2}-1}} d\omega.
\end{aligned}$$

Using (2.24) and (2.12), follows that

$$\begin{aligned}
F(Z_T) &= \sum_{l=0}^{\infty} \frac{\mu^{\mu+2l} \kappa^l (1+\kappa)^{\mu+l} v^{\frac{v}{2}} \left(\frac{v}{2}\right)^{-(\mu+l+\frac{v}{2}-1)}}{2^{\frac{v}{2}} l! B\left(\frac{v}{2}, \mu+l\right) (\mu+l+\frac{v}{2}-1) \exp(\mu\kappa)} \\
&\quad \times \frac{(Z_T)^{\mu+l}}{(\mu+l)} {}_2F_1\left(\mu+l+\frac{v}{2}-1, \mu+l; \mu+l+1; -\mu\left(\frac{2}{v}\right)(1+\kappa)Z_T\right) \\
&= \sum_{l=0}^{\infty} \frac{\mu^{\mu+2l} \kappa^l (1+\kappa)^{\mu+l} v^{-\mu-l+1} (Z_T)^{\mu+l}}{2^{-(\mu+l-1)} l! B\left(\frac{v}{2}, \mu+l\right) (\mu+l+\frac{v}{2}-1) (\mu+l) \exp(\mu\kappa)} \\
&\quad \times {}_2F_1\left(\mu+l+\frac{v}{2}-1, \mu+l; \mu+l+1; -\frac{2\mu(1+\kappa)Z_T}{v}\right) \\
&= \sum_{l=0}^{\infty} \frac{\mu^{\mu+2l} \kappa^l (1+\kappa)^{\mu+l} v^{-\mu-l+1} (Z_T)^{\mu+l}}{2^{-(\mu+l-1)} l! B\left(\frac{v}{2}, \mu+l\right) (\mu+l+\frac{v}{2}-1) (\mu+l) \exp(\mu\kappa)} \\
&\quad \times \left(1 - \frac{2\mu(1+\kappa)Z_T}{v}\right)^{-(\mu+l+\frac{v}{2}-1)} \\
&\quad \times {}_2F_1\left(\mu+l+\frac{v}{2}-1, 1; \mu+l+1; -\frac{2\mu(1+\kappa)Z_T}{2\mu(1+\kappa)Z_T+v}\right), \tag{3.42}
\end{aligned}$$

where $(\mu+l) > 0$ and ${}_2F_1(\cdot)$ is the Gaussian hypergeometric function (2.11).

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

3.3.3.2 Moments

From (3.35) the j^{th} moment of the normalized envelope, P , is

$$\begin{aligned}
 E(P^j) &= \int_0^{\infty} \rho^j f_P(\rho) d\rho \\
 &= \int_0^{\infty} \int_0^{\infty} \rho^j W(t) \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (t\rho)^\mu \exp\{-\mu\kappa - \mu(1+\kappa)(t\rho)^2\} \\
 &\quad \times I_{\mu-1}(2\mu\sqrt{\kappa(1+\kappa)}t\rho) dt d\rho \\
 &= \int_0^{\infty} \int_0^{\infty} W(t) \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}} t^\mu}{\kappa^{\frac{\mu-1}{2}}} \rho^{\mu+j} \exp\{-\mu\kappa - \mu(1+\kappa)t^2\rho^2\} \\
 &\quad \times I_{\mu-1}(2\mu\sqrt{\kappa(1+\kappa)}t\rho) dt d\rho.
 \end{aligned}$$

Using (2.22) the j^{th} moment is

$$\begin{aligned}
 E(P^j) &= \int_0^{\infty} W(t) \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}} t^\mu}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} 2^{-\mu+1-1} \left(2t\mu\sqrt{\kappa(1+\kappa)}\right)^{\mu-1} (\mu(1+\kappa)t^2)^{-\frac{(\mu+j+1+\mu-1)}{2}} \\
 &\quad \times \Gamma\left[\frac{\mu+j+1+\mu-1}{2}\right] {}_1F_1\left(\frac{\mu+j+1+\mu-1}{2}; \mu-1+1; \frac{(2t\mu\sqrt{\kappa(1+\kappa)})^2}{4\mu(1+\kappa)t^2}\right) dt \\
 &= \int_0^{\infty} W(t) \frac{\Gamma(\mu + \frac{j}{2})}{t^{j+1} [\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa)} {}_1F_1\left(\mu + \frac{j}{2}; \mu; \mu\kappa\right) dt. \tag{3.43}
 \end{aligned}$$

From (3.29) the j^{th} moment of the normalized power, Ω , is

$$\begin{aligned}
 E(\Omega^j) &= \int_0^{\infty} \omega^j f_\Omega(\omega) d\omega \\
 &= \int_0^{\infty} \int_0^{\infty} \omega^j W(t) \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (t\omega)^{\frac{\mu-1}{2}} \exp\{-\mu(1+\kappa)t\omega - \kappa\mu\} \\
 &\quad \times I_{\mu-1}(2\mu\sqrt{\kappa(1+\kappa)}t\omega) dt d\omega \\
 &= \int_0^{\infty} \int_0^{\infty} W(t) \frac{\mu(1+\kappa)^{\frac{\mu+1}{2}} (t)^{\frac{\mu-1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (\omega)^{\frac{\mu-1}{2}+j} \exp\{-\mu(1+\kappa)t\omega - \kappa\mu\} \\
 &\quad \times I_{\mu-1}(2\mu\sqrt{\kappa(1+\kappa)}t\omega) dt d\omega.
 \end{aligned}$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

Let $\alpha = \sqrt{\omega}$, then

$$E(\Omega^j) = \int_0^\infty \int_0^\infty W(t) \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}} (t)^{\frac{\mu-1}{2}}}{\kappa^{\frac{\mu-1}{2}}} (\alpha)^{\mu+1+2j-1} \exp\{-\mu(1+\kappa)t\alpha^2 - \kappa\mu\} \\ \times I_{\mu-1}\left(2\mu\sqrt{\kappa(1+\kappa)}t\alpha\right) d\alpha dt.$$

Using (2.22), we obtain the j^{th} moment as

$$E(\Omega^j) = \int_0^\infty W(t) \frac{2\mu(1+\kappa)^{\frac{\mu+1}{2}} (t)^{\frac{\mu-1}{2}}}{\kappa^{\frac{\mu-1}{2}} \exp(\mu\kappa)} 2^{-\mu+1-1} \left(2\mu\sqrt{t\kappa(1+\kappa)}\right)^{\mu-1} (\mu(1+\kappa)t)^{-\frac{(\mu+1+2j+\mu-1)}{2}} \\ \times \Gamma\left[\frac{\mu+1+2j+\mu-1}{2}\right] {}_1F_1\left(\frac{\mu+1+2j+\mu-1}{2}; \mu-1+1; \frac{\left(2\mu\sqrt{t\kappa(1+\kappa)}\right)^2}{4\mu(1+\kappa)t}\right) dt \\ E(\Omega^j) = \int_0^\infty W(t) \frac{\Gamma(\mu+j)}{t^{j+1}\Gamma(\mu)\exp(\mu\kappa)[\mu(1+\kappa)]^j} {}_1F_1(\mu+j; \mu; \mu\kappa) dt. \quad (3.44)$$

Remark 3.12 Substituting the weight functions (2.35) and (2.36) into (3.43) and (3.44) respectively. The following expressions are derived:

(i) For the normal distribution, the j^{th} moment of the normalized envelope, P , is

$$E(P^j) = \int_0^\infty \delta(t-1) \frac{\Gamma(\mu + \frac{j}{2})}{t^{j+1} [\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa)} {}_1F_1\left(\mu + \frac{j}{2}; \mu; \mu\kappa\right) dt.$$

Consider the result,

$$F(Z_T, t) = \int_0^\infty \delta(t) F(Z_T, t+1) dt \\ = F(Z_T, 1). \quad (3.45)$$

Let $q = t - 1$ then $t = q + 1$, the j^{th} moment is

$$E(P^j) = \int_0^\infty \delta(q) \frac{\Gamma(\mu + \frac{j}{2})}{(q+1)^{j+1} [\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa)} {}_1F_1\left(\mu + \frac{j}{2}; \mu; \mu\kappa\right) dq \\ = \frac{\Gamma(\mu + \frac{j}{2})}{[\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa)} {}_1F_1\left(\mu + \frac{j}{2}; \mu; \mu\kappa\right). \\ \equiv (3.18)$$

3. UNIVARIATE MODELS

3.3. The $\kappa - \mu$ type distribution

Using (3.44), the j^{th} moment of the normalized power, Ω , is

$$\begin{aligned}
 E(\Omega^j) &= \int_0^{\infty} \delta(t-1) \frac{\Gamma(\mu+j)}{t^{j+1} \Gamma(\mu) \exp(\mu\kappa) [\mu(1+\kappa)]^j} {}_1F_1(\mu+j; \mu; \mu\kappa) dt \\
 &= \frac{\Gamma(\mu+j)}{\Gamma(\mu) \exp(\mu\kappa) [\mu(1+\kappa)]^j} {}_1F_1(\mu+j; \mu; \mu\kappa). \\
 &\equiv (3.19)
 \end{aligned}$$

(ii) For the t -distribution, the j^{th} moment of the normalized envelope, P , is

$$\begin{aligned}
 E(P^j) &= \int_0^{\infty} \frac{v \left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right)}{2\Gamma\left(\frac{v}{2}\right)} \frac{\Gamma\left(\mu + \frac{j}{2}\right)}{t^{j+1} [\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa)} \\
 &\quad \times {}_1F_1\left(\mu + \frac{j}{2}; \mu; \mu\kappa\right) dt \\
 &= \frac{v \left(\frac{v}{2}\right)^{\frac{v}{2}-1} \Gamma\left(\mu + \frac{j}{2}\right) {}_1F_1\left(\mu + \frac{j}{2}; \mu; \mu\kappa\right)}{2\Gamma\left(\frac{v}{2}\right) [\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa)} \\
 &\quad \times \int_0^{\infty} t^{\frac{v}{2}-j-2} \exp\left(-\frac{vt}{2}\right) dt.
 \end{aligned}$$

Using (2.19), the j^{th} moment simplifies to

$$\begin{aligned}
 E(P^j) &= \frac{v \left(\frac{v}{2}\right)^{\frac{v}{2}-1} \Gamma\left(\mu + \frac{j}{2}\right) {}_1F_1\left(\mu + \frac{j}{2}; \mu; \mu\kappa\right) \Gamma\left(\frac{v}{2} - j - 1\right)}{2\Gamma\left(\frac{v}{2}\right) [\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa) \left(\frac{v}{2}\right)^{\frac{v}{2}-j-1}} \\
 &= \frac{2^{-j-1} v^{j+1} \Gamma\left(\mu + \frac{j}{2}\right) \Gamma\left(\frac{v}{2} - j - 1\right) {}_1F_1\left(\mu + \frac{j}{2}; \mu; \mu\kappa\right)}{\Gamma\left(\frac{v}{2}\right) [\mu(1+\kappa)]^{\frac{j}{2}} \Gamma(\mu) \exp(\mu\kappa)}. \tag{3.46}
 \end{aligned}$$

The j^{th} moment of the normalized power, Ω , is

$$\begin{aligned}
 E(\Omega^j) &= \int_0^{\infty} \frac{v \left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right)}{2\Gamma\left(\frac{v}{2}\right)} \frac{\Gamma(\mu+j)}{t^{j+1} \Gamma(\mu) \exp(\mu\kappa) [\mu(1+\kappa)]^j} \\
 &\quad \times {}_1F_1(\mu+j; \mu; \mu\kappa) dt \\
 &= \frac{v \left(\frac{v}{2}\right)^{\frac{v}{2}-1} \Gamma(\mu+j) {}_1F_1(\mu+j; \mu; \mu\kappa)}{2\Gamma\left(\frac{v}{2}\right) \Gamma(\mu) \exp\{\mu\kappa\} [\mu(1+\kappa)]^j} \\
 &\quad \times \int_0^{\infty} t^{\frac{v}{2}-j-2} \exp\left(-\frac{vt}{2}\right) dt.
 \end{aligned}$$

3. UNIVARIATE MODELS

3.4. Performance measures

Using (2.19), the j^{th} moment simplifies to

$$\begin{aligned}
 E(\Omega^j) &= \frac{v \left(\frac{v}{2}\right)^{\frac{v}{2}-1} \Gamma(\mu+j) {}_1F_1(\mu+j; \mu; \mu\kappa) \Gamma\left(\frac{v}{2}-j-1\right)}{2\Gamma\left(\frac{v}{2}\right) \Gamma(\mu) \exp\{\mu\kappa\} [\mu(1+\kappa)]^j \left(\frac{v}{2}\right)^{\frac{v}{2}-j-1}} \\
 &= \frac{2^{j+1} v^{j+1} \Gamma(\mu+j) \Gamma\left(\frac{v}{2}-j-1\right) {}_1F_1(\mu+j; \mu; \mu\kappa)}{\Gamma\left(\frac{v}{2}\right) \Gamma(\mu) \exp(\mu\kappa) [\mu(1+\kappa)]^j}.
 \end{aligned}$$

3.3.3.3 Amount of fading

The amount of fading for the $\kappa - \mu$ type model for the weight function (2.36) is given as (see (1.1))

$$\begin{aligned}
 AF &= \frac{E(\Omega^2)}{E(\Omega)^2} - 1 \\
 &= \frac{2^3 v^3 \Gamma(\mu+2) \Gamma\left(\frac{v}{2}-3\right) {}_1F_1(\mu+2; \mu; \mu\kappa)}{\Gamma\left(\frac{v}{2}\right) \Gamma(\mu) \exp(\mu\kappa) [\mu(1+\kappa)]^2} - 1 \\
 &= \frac{\left[\frac{2^2 v^2 \Gamma(\mu+1) \Gamma\left(\frac{v}{2}-2\right) {}_1F_1(\mu+1; \mu; \mu\kappa)}{\Gamma\left(\frac{v}{2}\right) \Gamma(\mu) \exp(\mu\kappa) [\mu(1+\kappa)]} \right]^2 - 1}{\left[\frac{2^2 v^2 \Gamma(\mu+1) \Gamma\left(\frac{v}{2}-2\right) {}_1F_1(\mu+1; \mu; \mu\kappa)}{\Gamma\left(\frac{v}{2}\right) \Gamma(\mu) \exp(\mu\kappa) [\mu(1+\kappa)]} \right]^2} - 1 \\
 &= \frac{(\mu+1) \left(\frac{v}{2}-2\right) \exp(\mu\kappa) {}_1F_1(\mu+2; \mu; \mu\kappa)}{2v\mu \left(\frac{v}{2}-1\right) \left(\frac{v}{2}\right) [{}_1F_1(\mu+1; \mu; \mu\kappa)]^2} - 1. \tag{3.47}
 \end{aligned}$$

3.4 Performance measures

In this chapter the $\kappa - \mu$ model is revisited and the $\kappa - \mu$ type model is derived. This section has graphical displays of the $\kappa - \mu$ type model and corresponding performance metrics. The plots of the $\kappa - \mu$ type model for the envelope (3.35) and power (3.29) will be illustrated for the weight functions (2.35) and (2.36) respectively. Similarly the outage probability as well as the amount of fading are plotted.

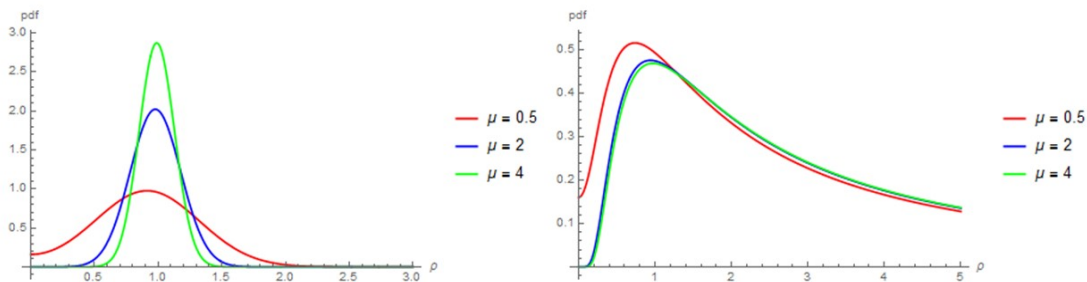


Figure 3.4.1 shows the pdf of the envelope if $\kappa = 5$ for the normal case (left) and t-distribution, where $v = 3$ degrees of freedom was considered, (right).

3. UNIVARIATE MODELS

3.4. Performance measures

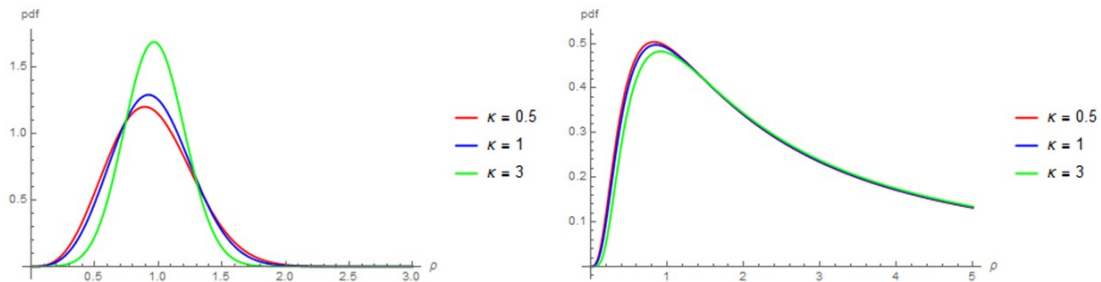


Figure 3.4.2 shows the pdf of the envelope if $\mu = 2$ for the normal case (left) and t-distribution, where $v = 3$ degrees of freedom was considered, (right).

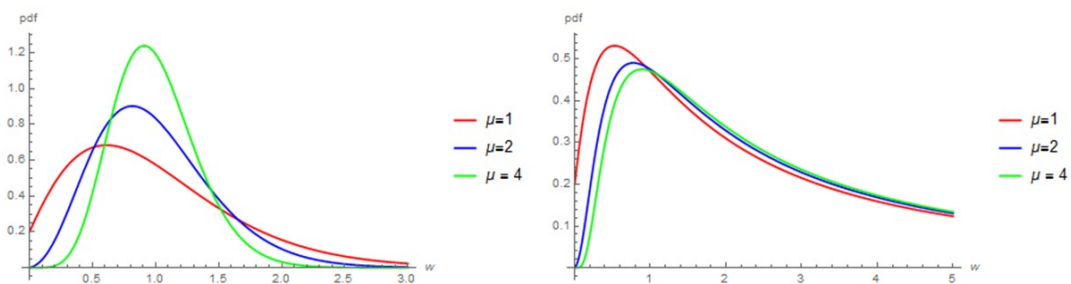


Figure 3.4.3 shows the pdf of the power if $\kappa = 3$ for the normal case (left) and t-distribution, where $v = 3$ degrees of freedom was considered, (right).

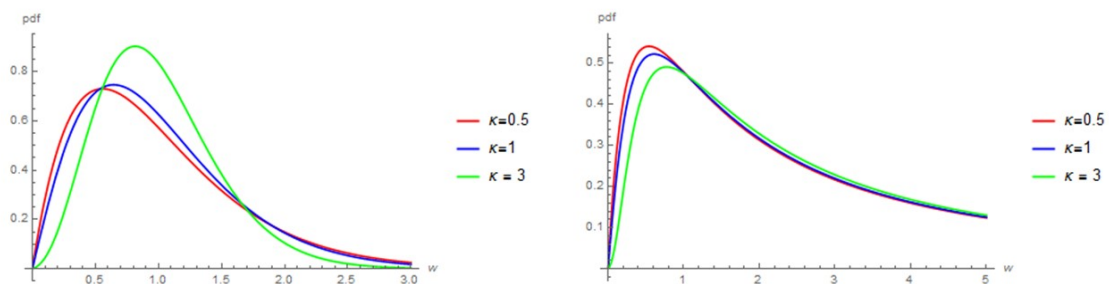


Figure 3.4.4 shows the pdf of the power if $\mu = 2$ for the normal case (left) and t-distribution, where $v = 3$ degrees of freedom was considered, (right).

The outage probability (3.40) is plotted for the weight function (2.35) and (2.36) in Figure 3.4.5 for $\kappa = 5$ and in Figure 3.4.6 for $\mu = 2$. (Note that for the t-distribution, $v = 3$ degrees of freedom was considered.). The assumption of the t-distribution as the underlying model is more appropriate for larger specified threshold values since the t-distribution outperforms the normal distribution only after a certain threshold.

3. UNIVARIATE MODELS

3.4. Performance measures

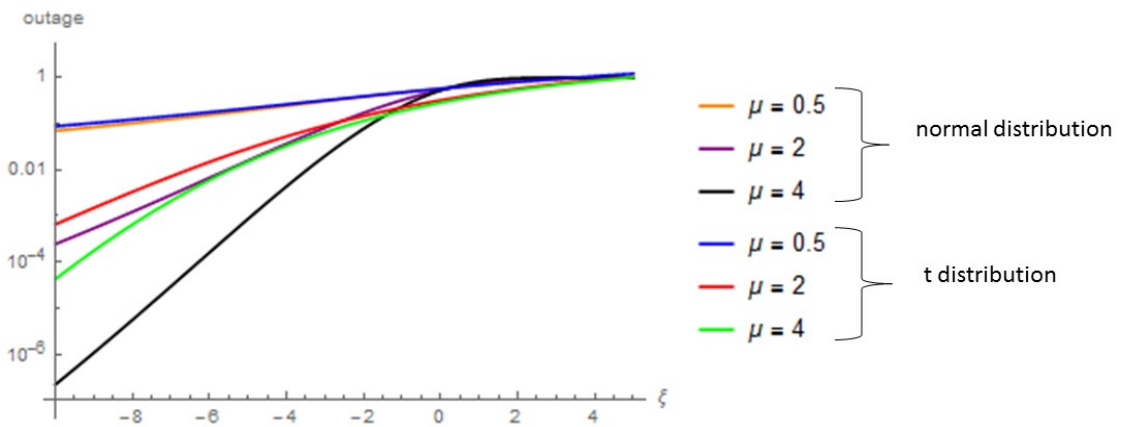


Figure 3.4.5

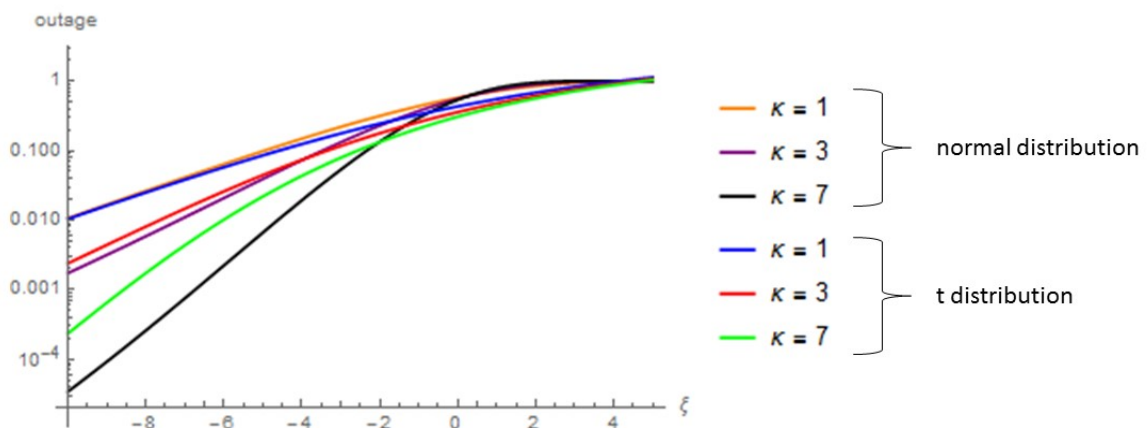


Figure 3.4.6

The amount of fading in the $\kappa - \mu$ channel is shown in Figure 3.4.7 for the normal ((3.20), on left) and t-distribution, where $v = 3$ degrees of freedom was considered, ((3.47), on right) cases respectively. The amount of fading for the t-distribution is lower than the normal distribution which indicates that t-distribution more suitable.

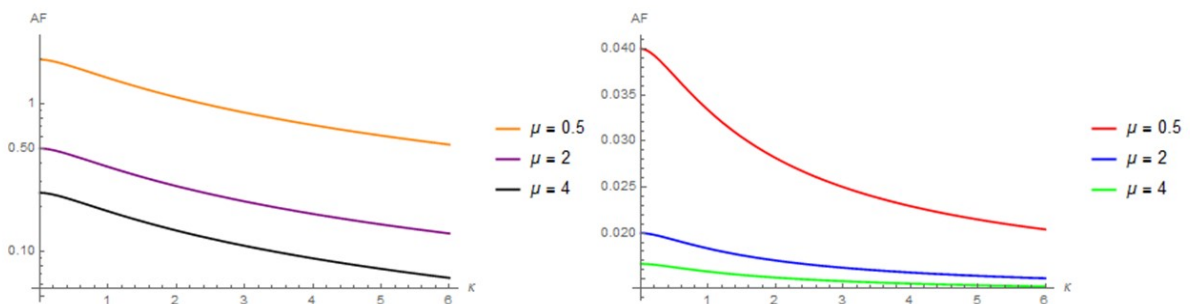


Figure 3.4.7

Chapter 4

Bivariate Models

4.1 Introduction

Villavicencio, et al.,(2016) [33] presented the bivariate $\kappa - \mu$ fading model used to assess the performance of a dual-branch selection combining (SC) scheme by using the framework developed by [19] for a general multidimensional normal distribution. According to Miller (1964) [19] each random process is considered to be normally distributed in $n = \mu$ (integer) dimensions. The approach of Miller (1964) [19] is followed by Villavicencio, et al.,(2016) [33] and generalized to accommodate a real extension of μ and outline in Section 4.2. In Section 4.2 the statistical properties of the bivariate $\kappa - \mu$ model will be given. The bivariate $\kappa - \mu$ type model that emanates from the elliptical assumption will then be derived in Section 4.3; followed by Section 4.5 where the performance measures under the different members of the elliptical model will be illustrated.

4.2 The bivariate $\kappa - \mu$ distribution

The aim of this section is to study a bivariate $\kappa - \mu$ distribution with as the underlying model, the normal distribution. Exact formulae for the jpdf, jcdf and joint arbitrary moments will be given. The outage probability is the primary metric for the analysis of diversity schemes in wireless communications systems and will also receive attention.

4.2.1 Description

Let R_1 and R_2 be two $\kappa - \mu$ envelopes then,

$$R_1^2 = \sum_{i=1}^{2\mu} X_i^2$$

$$R_2^2 = \sum_{i=1}^{2\mu} Y_i^2,$$

where X_i and Y_i are mutually independent normal processes with mean equal to a and variances σ_1^2 and σ_2^2 respectively. [Note: $2\mu =$ number of clusters of multipath].

4. BIVARIATE MODELS

4.2. The bivariate $\kappa - \mu$ distribution

If $\{X, Y, A\}$ be 2μ - dimensional vectors where $X = [X_1 \ X_2 \ \dots \ X_{2\mu}]^T$, $Y = [Y_1 \ Y_2 \ \dots \ Y_{2\mu}]^T$ and $A = [a \ a \ \dots \ a]^T$. Let $V_i = [X_i \ Y_i]^T$ where X_i and Y_i are correlated if and only if $i = j$, $i = 1, 2, \dots, 2\mu$. Note that A is the vector of the dominant component of the cluster. The interest is the jpdf of the envelope R_1 and R_2 , as well as, the normalized envelopes, P_1 and P_2 with $P_i = \frac{R_i}{\sqrt{\hat{r}_i}}$ and \hat{r}_i as defined in (3.13).

4.2.2 Derivation

The covariance matrix, $\Sigma > 0$, is then

$$\Sigma = \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix} = \begin{bmatrix} \sigma_1^2 & \delta\sigma_1\sigma_2 \\ \delta\sigma_1\sigma_2 & \sigma_2^2 \end{bmatrix},$$

where δ is the correlation coefficient of the normal component,

$$\begin{aligned} \delta &= \frac{\Sigma_{12}}{\sqrt{\Sigma_{11}\Sigma_{22}}} = \frac{\Sigma_{12}}{\sqrt{\sigma_1^2\sigma_2^2}} \\ \therefore \Sigma_{12} &= \delta\sqrt{\sigma_1^2\sigma_2^2} = \delta\sigma_1\sigma_2. \end{aligned}$$

The inverse of the covariance matrix is:

$$\begin{aligned} \Sigma^{-1} &= \begin{bmatrix} \sigma_1^2 & \delta\sigma_1\sigma_2 \\ \delta\sigma_1\sigma_2 & \sigma_2^2 \end{bmatrix}^{-1} \\ &= \frac{1}{\sigma_1^2\sigma_2^2 - \delta^2\sigma_1^2\sigma_2^2} \begin{bmatrix} \sigma_2^2 & -\delta\sigma_1\sigma_2 \\ -\delta\sigma_1\sigma_2 & \sigma_1^2 \end{bmatrix} \\ &= \frac{1}{\sigma_1^2\sigma_2^2(1 - \delta^2)} \begin{bmatrix} \sigma_2^2 & -\delta\sigma_1\sigma_2 \\ -\delta\sigma_1\sigma_2 & \sigma_1^2 \end{bmatrix} \\ &= \frac{1}{(1 - \delta^2)} \begin{bmatrix} \sigma_1^{-2} & -\delta\sigma_1^{-1}\sigma_2^{-1} \\ -\delta\sigma_1^{-1}\sigma_2^{-1} & \sigma_2^{-2} \end{bmatrix} \\ &= \begin{bmatrix} \Sigma_{11}^{-1} & \Sigma_{21}^{-1} \\ \Sigma_{21}^{-1} & \Sigma_{22}^{-1} \end{bmatrix}. \end{aligned} \tag{4.1}$$

The pdf of the multivariate normal distribution is given in (2.27). Thus, the jpdf of X and Y is as follows

$$\begin{aligned} f_{X,Y}(X, Y) &= \prod_{i=1}^{2\mu} \frac{1}{(2\pi) \det(\Sigma)^{\frac{1}{2}}} \exp\left(-\frac{1}{2} [(X_i - a)(Y_i - a)]^T \Sigma^{-1} \begin{bmatrix} (X_i - a) \\ (Y_i - a) \end{bmatrix}\right) \\ &= \prod_{i=1}^{2\mu} \frac{1}{(2\pi) \det(\Sigma)^{\frac{1}{2}}} \exp\left(-\frac{1}{2} [V_i - C_i]^T \Sigma^{-1} [V_i - C_i]\right) \\ &= \prod_{i=1}^{2\mu} f_i(V_i) \end{aligned}$$

4. BIVARIATE MODELS

4.2. The bivariate $\kappa - \mu$ distribution

$$f_{X,Y}(X, Y) = \frac{1}{(2\pi)^{2\mu} \det(\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} \sum_{i=1}^{2\mu} [V_i - C_i]^T \Sigma^{-1} [V_i - C_i]\right), \quad (4.2)$$

where $V_i = [X_i \ Y_i]^T$ and $C_i = [a \ a]^T$.

Expanding the quadratic form and replacing the V_i by X_i and Y_i it follows that

$$\begin{aligned} & \sum_{i=1}^{2\mu} [V_i - C_i]^T \Sigma^{-1} [V_i - C_i] \\ &= \sum_{i=1}^{2\mu} [(X_i - a) \ (Y_i - a)] \Sigma^{-1} \begin{bmatrix} (X_i - a) \\ (Y_i - a) \end{bmatrix}. \end{aligned} \quad (4.3)$$

Then from above and (4.1) it follows that

$$\begin{aligned} & \sum_{i=1}^{2\mu} [(X_i - a) \ (Y_i - a)] \begin{bmatrix} \Sigma_{11}^{-1} & \Sigma_{12}^{-1} \\ \Sigma_{21}^{-1} & \Sigma_{22}^{-1} \end{bmatrix} \begin{bmatrix} (X_i - a) \\ (Y_i - a) \end{bmatrix} \\ &= \sum_{i=1}^{2\mu} [(X_i - a) \Sigma_{11}^{-1} + (Y_i - a) \Sigma_{21}^{-1}] (X_i - a) + [(X_i - a) \Sigma_{21}^{-1} + (Y_i - a) \Sigma_{22}^{-1}] (Y_i - a) \\ &= \sum_{i=1}^{2\mu} (X_i - a) \Sigma_{11}^{-1} (X_i - a) + (Y_i - a) \Sigma_{21}^{-1} (X_i - a) + (X_i - a) \Sigma_{21}^{-1} (Y_i - a) \\ & \quad + (Y_i - a) \Sigma_{22}^{-1} (Y_i - a) \\ &= \sum_{i=1}^{2\mu} X_i \Sigma_{11}^{-1} X_i - X_i \Sigma_{11}^{-1} a - a \Sigma_{11}^{-1} X_i + a \Sigma_{11}^{-1} a + Y_i \Sigma_{21}^{-1} X_i - Y_i \Sigma_{21}^{-1} a - a \Sigma_{21}^{-1} X_i + \\ & \quad a \Sigma_{21}^{-1} a + X_i \Sigma_{21}^{-1} Y_i - X_i \Sigma_{21}^{-1} a - a \Sigma_{21}^{-1} X_i + a \Sigma_{21}^{-1} a + Y_i \Sigma_{22}^{-1} Y_i - Y_i \Sigma_{22}^{-1} a - a \Sigma_{22}^{-1} Y_i + a \Sigma_{22}^{-1} a \\ &= \sum_{i=1}^{2\mu} a (\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) a - a (\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) X_i - X_i (\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) a - \\ & \quad a (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) Y_i - Y_i (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) a + X_i \Sigma_{11}^{-1} X_i + X_i \Sigma_{21}^{-1} Y_i + Y_i \Sigma_{21}^{-1} X_i + Y_i \Sigma_{22}^{-1} Y_i. \end{aligned}$$

Since the inverse of the covariance matrix are constant, the above expression simplifies to

$$\begin{aligned} & \sum_{i=1}^{2\mu} a (\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) a - a (\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) X_i - X_i (\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) a - \\ & a (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) Y_i - Y_i (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) a + X_i \Sigma_{11}^{-1} X_i + X_i \Sigma_{21}^{-1} Y_i + Y_i \Sigma_{21}^{-1} X_i + Y_i \Sigma_{22}^{-1} Y_i \end{aligned}$$

4. BIVARIATE MODELS

 4.2. The bivariate $\kappa - \mu$ distribution

$$\begin{aligned}
 &= (\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} |X|^2 + \Sigma_{22}^{-1} |Y|^2 + \\
 &\quad \sum_{i=1}^{2\mu} [X_i \Sigma_{21}^{-1} Y_i + Y_i \Sigma_{21}^{-1} X_i - a (\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) X_i - X_i (\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) a \\
 &\quad - a (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) Y_i - Y_i (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) a] \\
 &= (\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} |X|^2 + \Sigma_{22}^{-1} |Y|^2 + 2\Sigma_{21}^{-1} XY - 2 (\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) AX \\
 &\quad - 2 (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) AY \tag{4.4}
 \end{aligned}$$

with $|A| = a = \sqrt{a_1^2 + a_2^2 + \dots + a_{2\mu}^2}$. Substituting (4.4) into (4.2), it follows that

$$\begin{aligned}
 &\frac{1}{(2\pi)^{2\mu} \det(\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} \sum_{i=1}^{2\mu} [V_i - C_i]^T \Sigma^{-1} [V_i - C_i]\right) \\
 &= \frac{1}{(2\pi)^{2\mu} \det(\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} |X|^2 + \Sigma_{22}^{-1} |Y|^2 + 2\Sigma_{21}^{-1} XY]\right) \\
 &\quad \exp\left(-\frac{1}{2} [-2 (\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) AX - 2 (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) AY]\right) \\
 &= \frac{1}{(2\pi)^{2\mu} |\det(\Sigma)|^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} |X|^2 + \Sigma_{22}^{-1} |Y|^2]\right) \\
 &\quad \exp[(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) AX + (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) AY - \Sigma_{21}^{-1} XY] \\
 &= \frac{1}{(2\pi)^{2\mu} \det(\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} |X|^2 + \Sigma_{22}^{-1} |Y|^2]\right) \\
 &\quad \exp[(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) AX + Y ((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} X)].
 \end{aligned}$$

Hence, the jpdf can be written as follows

$$\begin{aligned}
 f_{X,Y}(X, Y) &= \frac{1}{(2\pi)^{2\mu} \det(\Sigma)^\mu} \exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} |X|^2 + \Sigma_{22}^{-1} |Y|^2]\right) \\
 &\quad \times \exp[(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) AX + Y ((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} X)].
 \end{aligned}$$

In order to obtain the jpdf of the envelopes, an integration with respect to X and Y subject to the constraint $|X| = R_1$, and $|Y| = R_2$ is required (see Miller (1964) [19, page 33]). That is

$$\begin{aligned}
 f_{X,Y}(X, Y) &= \frac{\exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} |X|^2 + \Sigma_{22}^{-1} |Y|^2]\right)}{(2\pi)^{2\mu} \det(\Sigma)^\mu} \\
 &\quad \times \int_{|X|=R_1} \exp[(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) X^T A] ds_1 \\
 &\quad \times \int_{|Y|=R_2} \exp[Y^T ((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} X)] ds_2, \tag{4.5}
 \end{aligned}$$

4. BIVARIATE MODELS

 4.2. The bivariate $\kappa - \mu$ distribution

where ds_1 and ds_2 are the elements of the integration surface area (see theorem 2.1). The second integral in the (4.5) can be expressed as

$$\begin{aligned}
 & \int_{|Y|} \exp [Y^T ((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} X)] ds_2 \\
 &= (2\pi R_2)^{\frac{2\mu}{2}} |(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} R_1|^{1-\mu} \\
 & \quad \times I_{\frac{1}{2}(2\mu-2)}(R_2 |(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} R_1|), \tag{4.6}
 \end{aligned}$$

Using (2.16), the expression (4.6) can be written as

$$\begin{aligned}
 & (2\pi R_2)^{\frac{2\mu}{2}} |(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} R_1|^{1-\mu} I_{\frac{1}{2}(2\mu-2)}(R_2 |(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} R_1|) \\
 &= \frac{(2\pi R_2)^\mu}{(R_2)^{1-\mu}} (R_2 |(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} R_1|)^{1-\mu} I_{\frac{1}{2}(2\mu-2)}(R_2 |(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) A - \Sigma_{21}^{-1} R_1|) \\
 &= \frac{(2\pi R_2)^\mu}{(R_2)^{1-\mu}} 2^{\mu-1} \Gamma(\mu-1) \sum_{k=0}^{\infty} (-1)^k (\mu-1+k) \frac{I_{\mu-1+k}((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A| R_2)}{(R_2)^{\mu-1} (\Sigma_{21}^{-1} + \Sigma_{22}^{-1})^{\mu-1} |A|^{\mu-1}} \\
 & \quad \times \frac{I_{\mu-1+k}(\Sigma_{21}^{-1} R_1 R_2)}{(\Sigma_{21}^{-1})^{\mu-1} (R_1 R_2)^{\mu-1}} C_k^{\mu-1}(\cos \phi_1) \\
 &= \frac{2^{2\mu-1} (\pi R_2)^\mu}{[(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) (\Sigma_{21}^{-1}) R_2]^{\mu-1} |A|^{\mu-1} R_1^{\mu-1}} \Gamma(\mu-1) \sum_{k=0}^{\infty} (-1)^k (\mu-1+k) \\
 & \quad \times I_{\mu-1+k}((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A| R_2) I_{\mu-1+k}(\Sigma_{21}^{-1} R_1 R_2) C_k^{\mu-1}(\cos \phi_1),
 \end{aligned}$$

where ϕ_1 is the angle between X and A .

Thus, the jpdf of the envelopes is given as

$$\begin{aligned}
 & \frac{2^{2\mu-1} (\pi R_2)^\mu \Gamma(\mu-1) \exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} R_1^2 + \Sigma_{22}^{-1} R_2^2]\right)}{(2\pi)^{2\mu} \det(\Sigma)^\mu [(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) (\Sigma_{21}^{-1}) R_2]^{\mu-1} |A|^{\mu-1} R_1^{\mu-1}} \\
 & \quad \times \sum_{k=0}^{\infty} (-1)^k (\mu-1+k) I_{\mu-1+k}((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A| R_2) I_{\mu-1+k}(\Sigma_{21}^{-1} R_1 R_2) \\
 & \quad \times \int_{|X|} \exp [(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) X^T A] C_k^{\mu-1}(\cos \phi_1) ds_1. \tag{4.7}
 \end{aligned}$$

To evaluate the remaining integral in equation (4.7), express it in generalized spherical coordinates

$$\begin{aligned}
 & \int_{|X|} \exp [(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) X^T A] C_k^{\mu-1}(\cos \phi_1) ds_1 \\
 &= \int_{|X|} \exp [(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A \cos \phi_1] C_k^{\mu-1}(\cos \phi_1) ds_1
 \end{aligned}$$

4. BIVARIATE MODELS

4.2. The bivariate $\kappa - \mu$ distribution

$$\begin{aligned}
&= R_1^{2\mu-1} \int_0^{2\pi} d\theta \int_0^\pi \exp [(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A \cos \phi_1] C_k^{\mu-1}(\cos \phi_1) \sin^{2\mu-2} d\phi_1 \\
&\quad \times \prod_{k=2}^{2\mu-2} \int_0^\pi \sin^{2\mu-1-k} \phi_k d\phi_k.
\end{aligned} \tag{4.8}$$

Applying (2.17) then it follows that

$$\begin{aligned}
&\int_0^\pi \exp [(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A \cos \phi_1] C_k^{\mu-1}(\cos \phi_1) \sin^{2\mu-2} d\phi_1 \\
&= \frac{2^{\mu-1} \Gamma(\mu - 1 + \frac{1}{2}) \Gamma(\frac{1}{2}) \Gamma(2(\mu - 1) + k)}{((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A)^{\mu-1} k! \Gamma(2(\mu - 1))} \\
&\quad \times I_{\mu-1+k}((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A),
\end{aligned} \tag{4.9}$$

and using (2.18),

$$\begin{aligned}
&R_1^{2\mu-1} \int_0^{2\pi} d\theta \prod_{k=2}^{2\mu-2} \int_0^\pi \sin^{2\mu-1-k} \phi_k d\phi_k \\
&= \frac{2\pi^{\frac{1}{2}(2\mu-1)} R_1^{2\mu-1}}{\Gamma(\frac{1}{2}(2\mu-1))}.
\end{aligned} \tag{4.10}$$

Therefore from (4.8), (4.9) and (4.10) follows that

$$\begin{aligned}
&\int_{|X|} \exp [(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) X^T A] C_k^{\mu-1}(\cos \phi_1) ds_1 \\
&= \frac{2\pi^{\frac{1}{2}(2\mu-1)} R_1^{2\mu-1} 2^{\mu-1} \Gamma(\mu - 1 + \frac{1}{2}) \Gamma(\frac{1}{2}) \Gamma(2(\mu - 1) + k)}{\Gamma(\frac{1}{2}(2\mu-1)) ((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A)^{\mu-1} k! \Gamma(2(\mu - 1))} I_{\mu-1+k}((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A) \\
&= \frac{2^\mu \pi^{\mu-\frac{1}{2}} \Gamma(\frac{1}{2}) R_1^{2\mu-1}}{((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A)^{\mu-1}} \frac{\Gamma(2\mu - 2 + k)}{k! \Gamma(2\mu - 2)} I_{\mu-1+k}((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A) \\
&= \frac{2^\mu \pi^\mu R_1^{2\mu-1}}{((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A)^{\mu-1}} \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k}((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A),
\end{aligned} \tag{4.11}$$

where $\binom{2\mu + k - 3}{2\mu - 3}$ is the binomial coefficient for $k \geq 0$. Substituting (4.11) into (4.7)

4. BIVARIATE MODELS
4.2. The bivariate $\kappa - \mu$ distribution

the jpdf of the envelopes is

$$\begin{aligned}
f_{R_1, R_2}(R_1, R_2) &= \frac{2^{2\mu-1} R_2 \Gamma(\mu - 1) \exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} R_1^2 + \Sigma_{22}^{-1} R_2^2]\right)}{\pi^\mu \det(\Sigma)^\mu [(\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) (\Sigma_{21}^{-1})]^{\mu-1} |A|^{\mu-1} R_1^{\mu-1}} \\
&\times \sum_{k=0}^{\infty} (-1)^k (\mu - 1 + k) I_{\mu-1+k}((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A| R_2) I_{\mu-1+k}(\Sigma_{21}^{-1} R_1 R_2) \\
&\times \frac{2^\mu \pi^\mu R_1^{2\mu-1}}{((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A)^{\mu-1}} \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k}((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A) \\
&= \frac{2^{\mu-1} R_1 R_2 \Gamma(\mu - 1) \exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} R_1^2 + \Sigma_{22}^{-1} R_2^2]\right)}{\det(\Sigma)^\mu [(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) (\Sigma_{21}^{-1}) |A|^2]^{\mu-1}} \\
&\times \sum_{k=0}^{\infty} (-1)^k (\mu - 1 + k) \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k}((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A| R_2) \\
&\times I_{\mu-1+k}(\Sigma_{21}^{-1} R_1 R_2) I_{\mu-1+k}((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A). \tag{4.12}
\end{aligned}$$

This expression (4.12) can be written in terms of the coefficients of the inverse of the covariance matrix by using (4.1). Thus, the jpdf is

$$\begin{aligned}
f_{R_1, R_2}(R_1, R_2) &= \frac{2^{\mu-1} R_1 R_2 \Gamma(\mu - 1) \exp\left(-\frac{1}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} R_1^2 + \Sigma_{22}^{-1} R_2^2]\right)}{\det(\Sigma)^\mu [(\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) (\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) (\Sigma_{21}^{-1}) |A|^2]^{\mu-1}} \\
&\times \sum_{k=0}^{\infty} (-1)^k (\mu - 1 + k) \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k}((\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A| R_2) \\
&\times I_{\mu-1+k}(\Sigma_{21}^{-1} R_1 R_2) I_{\mu-1+k}((\Sigma_{11}^{-1} + \Sigma_{21}^{-1}) R_1 A) \\
&= \sum_{k=0}^{\infty} \frac{2^{\mu-1} R_1 R_2 \Gamma(\mu - 1) (-1)^k (\mu - 1 + k) \binom{2\mu + k - 3}{2\mu - 3}}{[\sigma_1^2 \sigma_2^2 (1 - \delta^2)]^\mu \left[\frac{|A| \delta (1 - \delta \frac{\sigma_2^2}{\sigma_1}) (1 - \delta \frac{\sigma_1}{\sigma_2})}{-(1 - \delta^2)^3 \sigma_1^3 \sigma_2^3} \right]^{\mu-1}} \\
&\times \exp \left\{ -\frac{1}{2} \left(\frac{1}{(1 - \delta^2)} \left[\frac{R_1^2}{\sigma_1^2} + \frac{R_2^2}{\sigma_2^2} + \left(\frac{\sigma_1^2 + \sigma_2^2 - 2\delta\sigma_1\sigma_2}{\sigma_1^2 \sigma_2^2} \right) |A|^2 \right] \right) \right\} \\
&\times I_{\mu-1+k} \left(\frac{|A| R_2 (1 - \delta \frac{\sigma_2^2}{\sigma_1})}{(1 - \delta^2) \sigma_2^2} \right) I_{\mu-1+k} \left(\frac{|A| R_1 (1 - \delta \frac{\sigma_1}{\sigma_2})}{(1 - \delta^2) \sigma_1^2} \right) I_{\mu-1+k} \left(\frac{R_1 R_2 \delta}{(1 - \delta^2) \sigma_1 \sigma_2} \right) \\
&= \sum_{k=0}^{\infty} \frac{R_1 R_2 \Gamma(\mu - 1)}{(\sigma_1^2 \sigma_2^2)^\mu (1 - \delta^2)^\mu} \left[\frac{-2(1 - \delta^2)^3 \sigma_1^3 \sigma_2^3}{|A|^2 \delta (1 - \delta \frac{\sigma_2^2}{\sigma_1}) (1 - \delta \frac{\sigma_1}{\sigma_2})} \right]^{\mu-1} \\
&\times \exp \left\{ -\frac{1}{2} \left(\frac{1}{(1 - \delta^2)} \left[\frac{R_1^2}{\sigma_1^2} + \frac{R_2^2}{\sigma_2^2} + \left(\frac{\sigma_1^2 + \sigma_2^2 - 2\delta\sigma_1\sigma_2}{\sigma_1^2 \sigma_2^2} \right) |A|^2 \right] \right) \right\} \\
&\times (-1)^k (\mu - 1 + k) \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k} \left(\frac{R_1 R_2 \delta}{(1 - \delta^2) \sigma_1 \sigma_2} \right) \\
&\times I_{\mu-1+k} \left(\frac{|A| R_2 (1 - \delta \frac{\sigma_2^2}{\sigma_1})}{(1 - \delta^2) \sigma_2^2} \right) I_{\mu-1+k} \left(\frac{|A| R_1 (1 - \delta \frac{\sigma_1}{\sigma_2})}{(1 - \delta^2) \sigma_1^2} \right). \tag{4.13}
\end{aligned}$$

4. BIVARIATE MODELS

4.2. The bivariate $\kappa - \mu$ distribution

Define $\kappa_i = \frac{|A_i|^2}{2\mu\sigma_i^2} = \frac{a_i^2}{\sigma_i^2}$, $i = 1, 2$. Then by normalizing the two $\kappa - \mu$ envelopes, R_1 and R_2 (see equation (3.13)), where $P_i = \frac{R_i}{\sqrt{\hat{\sigma}_i}}$. The Jacobian of the transformation from $(R_1, R_2) \longrightarrow (P_1, P_2)$ is

$$\begin{aligned} |J((R_1, R_2) \longrightarrow (P_1, P_2))| &= \begin{vmatrix} \sigma_1 \sqrt{2\mu(1 + \kappa_1)} & 0 \\ 0 & \sigma_2 \sqrt{2\mu(1 + \kappa_2)} \end{vmatrix} \\ &= \sigma_1 \sigma_2 2\mu \sqrt{(1 + \kappa_1)(1 + \kappa_2)}. \end{aligned}$$

The final expression for the normalized jpdf of the envelopes is

$$\begin{aligned} f_{P_1, P_2}(\rho_1, \rho_2) &= \sum_{k=0}^{\infty} \frac{\Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} 2\mu\rho_1\rho_2 \sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(\sigma_1\sigma_2) [\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \\ &\quad \times \exp \left\{ \frac{-2\mu}{2(1 - \delta^2)} [\rho_1^2(1 + \kappa_1) + \rho_2^2(1 + \kappa_2) + \kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}] \right\} \\ &\quad \times (-1)^k (\mu + k - 1) \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k} \left(\frac{2\delta\mu\rho_1\rho_2 \sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right) \\ &\quad \times I_{\mu-1+k} \left(\frac{2\mu\rho_2 \sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right) I_{\mu-1+k} \left(\frac{2\mu\rho_1 \sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right) \\ &= \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1)(1 + \kappa_2)}{[\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \\ &\quad \times \exp \left\{ \frac{-\mu}{(1 - \delta^2)} [\rho_1^2(1 + \kappa_1) + \rho_2^2(1 + \kappa_2) + \kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}] \right\} \\ &\quad \times \sum_{k=0}^{\infty} \frac{(\mu + k - 1) (2\mu - 2)_k}{k!} I_{\mu-1+k} \left(\frac{2\delta\mu\rho_1\rho_2 \sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right) \\ &\quad \times I_{\mu-1+k} \left(\frac{2\mu\rho_2 \sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right) \\ &\quad \times I_{\mu-1+k} \left(\frac{2\mu\rho_1 \sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right). \end{aligned} \tag{4.14}$$

4.2.3 Characteristics

4.2.3.1 Joint cumulative distribution function

The jpdf in (4.14) is represented as follows in order to get an expression for the jcdf, that will be used to obtain the outage probability. Expand the Bessel functions as infinite series by using (2.13) and then integrate term-wise, the jpdf (4.14) can be written as

4. BIVARIATE MODELS

 4.2. The bivariate $\kappa - \mu$ distribution

$$\begin{aligned}
 f_{P_1, P_2}(\rho_1, \rho_2) &= \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1) (1 + \kappa_2)}{[\delta\mu (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^\mu} \\
 &\times \exp \left\{ \frac{-2\mu}{2(1 - \delta^2)} [\rho_1^2 (1 + \kappa_1) + \rho_2^2 (1 + \kappa_2) + \kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1 \kappa_2}] \right\} \\
 &\times \sum_{k=0}^{\infty} \frac{(\mu + k - 1) (2\mu - 2)_k}{k!} \\
 &\times \sum_{z=0}^{\infty} \left(\frac{2\delta\mu \rho_1 \rho_2 \sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{2(1 - \delta^2)} \right)^{\mu-1+k} \frac{\left(\frac{1}{4} \left(\frac{2\delta\mu \rho_1 \rho_2 \sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right)^2 \right)^z}{z! \Gamma(\mu + k + z)} \\
 &\times \sum_{w=0}^{\infty} \left(\frac{2\mu \rho_2 \sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{2(1 - \delta^2)} \right)^{\mu-1+k} \frac{\left(\frac{1}{4} \left(\frac{2\mu \rho_2 \sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right)^2 \right)^w}{w! \Gamma(\mu + k + w)} \\
 &\times \sum_{l=0}^{\infty} \left(\frac{2\mu \rho_1 \sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{2(1 - \delta^2)} \right)^{\mu-1+k} \frac{\left(\frac{1}{4} \left(\frac{2\mu \rho_1 \sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right)^2 \right)^l}{l! \Gamma(\mu + k + l)} \\
 &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1) (1 + \kappa_2)}{[\delta\mu (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^\mu} \\
 &\times \exp \left\{ \frac{-\mu}{(1 - \delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1 \kappa_2}) \right\} \\
 &\times \frac{(\mu + k - 1) (2\mu - 2)_k}{k! z! w! l! \Gamma(\mu + k + l) \Gamma(\mu + k + w) \Gamma(\mu + k + z)} \\
 &\times \exp \left\{ \frac{-\mu \rho_1^2 (1 + \kappa_1)}{(1 - \delta^2)} \right\} \left(\frac{\rho_1^2 \mu (1 + \kappa_1)}{1 - \delta^2} \right)^{l+k+z+\mu-1} \left(\frac{1 - \delta^2}{\rho_1^2 \mu (1 + \kappa_1)} \right)^{l+k+z+\mu-1} \\
 &\times \exp \left\{ \frac{-\mu \rho_2^2 (1 + \kappa_2)}{(1 - \delta^2)} \right\} \left(\frac{\rho_2^2 \mu (1 + \kappa_2)}{1 - \delta^2} \right)^{w+k+z+\mu-1} \left(\frac{1 - \delta^2}{\rho_2^2 \mu (1 + \kappa_2)} \right)^{w+k+z+\mu-1} \\
 &\times \left(\frac{\mu \rho_1 \sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{\mu^2 \rho_1^2 (1 + \kappa_1) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1 - \delta^2)^2} \right)^l \\
 &\times \left(\frac{\mu \rho_2 \sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{\mu^2 \rho_2^2 (1 + \kappa_2) (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1 - \delta^2)^2} \right)^w \\
 &\times \left(\frac{\delta\mu \rho_1 \rho_2 \sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{\delta^2 \mu^2 \rho_1^2 \rho_2^2 (1 + \kappa_1) (1 + \kappa_2)}{(1 - \delta^2)^2} \right)^z. \quad (4.15)
 \end{aligned}$$

4. BIVARIATE MODELS

4.2. The bivariate $\kappa - \mu$ distribution

Thus from (4.15), the jcdf

$$\begin{aligned}
F_{P_1, P_2}(p_1, p_2) &= \int_0^{p_1} \int_0^{p_2} f_{P_1, P_2}(\rho_1, \rho_2) d\rho_1 d\rho_2 \\
&= \int_0^{q_1} \int_0^{q_2} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu-1) (1-\delta^2)^{2\mu-3} (1+\kappa_1)(1+\kappa_2)}{[\delta\mu(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})]^{\mu-1}} \\
&\quad \times \exp\left\{\frac{-\mu}{(1-\delta^2)}(\kappa_1+\kappa_2-2\delta\sqrt{\kappa_1\kappa_2})\right\} \\
&\quad \times \frac{(\mu+k-1)(2\mu-2)_k}{k!z!w!!\Gamma(\mu+k+l)\Gamma(\mu+k+w)\Gamma(\mu+k+z)} \\
&\quad \times \exp\left\{\frac{-\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)}\right\} \left(\frac{\rho_1^2\mu(1+\kappa_1)}{1-\delta^2}\right)^{l+k+z+\mu-1} \left(\frac{1-\delta^2}{\rho_1^2\mu(1+\kappa_1)}\right)^{l+k+z+\mu-1} \\
&\quad \times \exp\left\{\frac{-\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)}\right\} \left(\frac{\rho_2^2\mu(1+\kappa_2)}{1-\delta^2}\right)^{w+k+z+\mu-1} \left(\frac{1-\delta^2}{\rho_2^2\mu(1+\kappa_2)}\right)^{w+k+z+\mu-1} \\
&\quad \times \left(\frac{\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu^2\rho_1^2(1+\kappa_1)(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \\
&\quad \times \left(\frac{\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu^2\rho_2^2(1+\kappa_2)(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w \\
&\quad \times \left(\frac{\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\delta^2\mu^2\rho_1^2\rho_2^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2}\right)^z d\rho_1 d\rho_2 \\
&= \int_0^{q_1} \int_0^{q_2} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu-1) (1-\delta^2)^{2\mu-3} (1+\kappa_1)(1+\kappa_2)}{[\delta\mu(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})]^{\mu-1}} \\
&\quad \times \exp\left\{\frac{-\mu}{(1-\delta^2)}(\kappa_1+\kappa_2-2\delta\sqrt{\kappa_1\kappa_2})\right\} \\
&\quad \times \frac{(\mu+k-1)(2\mu-2)_k}{k!z!w!!\Gamma(\mu+k+l)\Gamma(\mu+k+w)\Gamma(\mu+k+z)} \\
&\quad \times \left(\frac{1-\delta^2}{\rho_1^2\mu(1+\kappa_1)}\right)^{l+k+z+\mu-1} \left(\frac{1-\delta^2}{\rho_2^2\mu(1+\kappa_2)}\right)^{w+k+z+\mu-1} \\
&\quad \times \left(\frac{\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu^2\rho_1^2(1+\kappa_1)(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \\
&\quad \times \left(\frac{\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu^2\rho_2^2(1+\kappa_2)(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w
\end{aligned}$$

4. BIVARIATE MODELS

4.2. The bivariate $\kappa - \mu$ distribution

$$\begin{aligned} & \times \left(\frac{\delta \mu \rho_1 \rho_2 \sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{\delta^2 \mu^2 \rho_1^2 \rho_2^2 (1 + \kappa_1)(1 + \kappa_2)}{(1 - \delta^2)^2} \right)^z \\ & \times \exp \left\{ -\frac{\mu \rho_1^2 (1 + \kappa_1)}{(1 - \delta^2)} \right\} \left(\frac{\mu \rho_1^2 (1 + \kappa_1)}{(1 - \delta^2)} \right)^{l+k+z+\mu-1} \\ & \times \exp \left\{ -\frac{\mu \rho_2^2 (1 + \kappa_2)}{(1 - \delta^2)} \right\} \left(\frac{\mu \rho_2^2 (1 + \kappa_2)}{(1 - \delta^2)} \right)^{w+k+z+\mu-1} d\rho_1 d\rho_2. \end{aligned}$$

After simplifying it follows,

$$\begin{aligned} F_{P_1, P_2}(p_1, p_2) &= \int_0^{p_1} \int_0^{p_2} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{4\mu^2 \Gamma(\mu-1) (1-\delta^2)^{2\mu-3} (1+\kappa_1)(1+\kappa_2)}{[\delta \mu (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})]^{\mu-1}} \\ & \times \exp \left\{ \frac{-\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta \sqrt{\kappa_1 \kappa_2}) \right\} \\ & \times \frac{(\mu+k-1)(2\mu-2)_k}{k!z!w!l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\ & \times \left(\frac{1-\delta^2}{\mu(1+\kappa_1)} \right)^{l+k+z+\mu-1} \left(\frac{1-\delta^2}{\mu(1+\kappa_2)} \right)^{w+k+z+\mu-1} \\ & \times \left(\frac{\mu \sqrt{1+\kappa_1} (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})}{(1-\delta^2)} \right)^{\mu-1+k} \left(\frac{\mu^2 (1+\kappa_1) (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})^2}{(1-\delta^2)^2} \right)^l \\ & \times \left(\frac{\mu \sqrt{1+\kappa_2} (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})}{(1-\delta^2)} \right)^{\mu-1+k} \left(\frac{\mu^2 (1+\kappa_2) (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})^2}{(1-\delta^2)^2} \right)^w \\ & \times \left(\frac{\delta \mu \sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)} \right)^{\mu-1+k} \left(\frac{\delta^2 \mu^2 (1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2} \right)^z \\ & \times \left(\frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right)^{l+k+z+\mu-1} \left(\frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+z+\mu-1} \\ & \times \rho_1^{2\mu+2k+2z+2l-1} \rho_2^{2\mu+2k+2z+2w-1} \exp \left\{ -\frac{\rho_1^2 \mu (1+\kappa_1)}{(1-\delta^2)} \right\} \\ & \times \exp \left\{ -\frac{\rho_2^2 \mu (1+\kappa_2)}{(1-\delta^2)} \right\} d\rho_1 d\rho_2. \end{aligned}$$

4. BIVARIATE MODELS

4.2. The bivariate $\kappa - \mu$ distribution

Using (2.20), the jcdf is

$$\begin{aligned}
 F_{P_1, P_2}(\rho_1, \rho_2) &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{\mu^2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu - 3} (1 + \kappa_1) (1 + \kappa_2)}{[\delta \mu (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})]^{\mu - 1}} \\
 &\quad \times \frac{(\mu + k - 1) (2\mu - 2)_k}{k! z! w! l! \Gamma(\mu + k + l) \Gamma(\mu + k + w) \Gamma(\mu + k + z)} \\
 &\quad \times \exp \left\{ \frac{-\mu}{(1 - \delta^2)} (\kappa_1 + \kappa_2 - 2\delta \sqrt{\kappa_1 \kappa_2}) \right\} \left(\frac{1 - \delta^2}{\mu (1 + \kappa_1)} \right)^{l+k+z+\mu-1} \\
 &\quad \times \left(\frac{1 - \delta^2}{\mu (1 + \kappa_2)} \right)^{w+k+z+\mu-1} \left(\frac{\mu \sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})}{(1 - \delta^2)} \right)^{\mu - 1 + k} \\
 &\quad \times \left(\frac{\mu^2 (1 + \kappa_1) (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})^2}{(1 - \delta^2)^2} \right)^l \left(\frac{\mu \sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})}{(1 - \delta^2)} \right)^{\mu - 1 + k} \\
 &\quad \times \left(\frac{\mu^2 (1 + \kappa_2) (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})^2}{(1 - \delta^2)^2} \right)^w \left(\frac{\delta \mu \sqrt{(1 + \kappa_1) (1 + \kappa_2)}}{(1 - \delta^2)} \right)^{\mu - 1 + k} \\
 &\quad \times \left(\frac{\delta^2 \mu^2 (1 + \kappa_1) (1 + \kappa_2)}{(1 - \delta^2)^2} \right)^z \left(\frac{\mu (1 + \kappa_1)}{(1 - \delta^2)} \right)^{l+k+z+\mu-1} \\
 &\quad \times \left(\frac{\mu (1 + \kappa_2)}{(1 - \delta^2)} \right)^{w+k+z+\mu-1} \left[\frac{\mu (1 + \kappa_1)}{(1 - \delta^2)} \right]^{-(\mu+k+z+l)} \left[\frac{\mu (1 + \kappa_2)}{(1 - \delta^2)} \right]^{-(\mu+k+z+w)} \\
 &\quad \times \gamma \left(\mu + k + z + l, \frac{\mu (1 + \kappa_1) \rho_1^2}{(1 - \delta^2)} \right) \gamma \left(\mu + k + z + w, \frac{\mu (1 + \kappa_2) \rho_2^2}{(1 - \delta^2)} \right) dt.
 \end{aligned}$$

Further simplification results in

$$\begin{aligned}
 F_{P_1, P_2}(\rho_1, \rho_2) &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{\Gamma(\mu - 1) (\mu + k - 1) (2\mu - 2)_k (1 - \delta^2)^{\mu} \delta^{k+2z}}{k! z! l! w! \Gamma(\mu + k + l) \Gamma(\mu + k + w) \Gamma(\mu + k + z)} \\
 &\quad \times \exp \left\{ \frac{-\mu}{(1 - \delta^2)} (\kappa_1 + \kappa_2 - 2\delta \sqrt{\kappa_1 \kappa_2}) \right\} \\
 &\quad \times \gamma \left(l + k + z + \mu, \frac{\mu \rho_1^2 (1 + \kappa_1)}{(1 - \delta^2)} \right) \gamma \left(w + k + z + \mu, \frac{\mu \rho_2^2 (1 + \kappa_2)}{(1 - \delta^2)} \right) \\
 &\quad \times \left(\frac{\mu (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})}{(1 - \delta^2)} \right)^{l + \frac{k}{2}} \left(\frac{\mu (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})}{(1 - \delta^2)} \right)^{w + \frac{k}{2}}. \tag{4.16}
 \end{aligned}$$

4. BIVARIATE MODELS

 4.2. The bivariate $\kappa - \mu$ distribution

4.2.3.2 Moments

Using the jpdf (4.14), the joint arbitrary moments can be found as

$$\begin{aligned}
 E[P_1^{n_1} P_2^{n_2}] &= \int_0^\infty \int_0^\infty \rho_1^{n_1} \rho_2^{n_2} f_{\rho_1, \rho_2}(\rho_1, \rho_2) d\rho_1 d\rho_2 \\
 &= \int_0^\infty \int_0^\infty \rho_1^{n_1} \rho_2^{n_2} \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1) (1 + \kappa_2)}{[\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^\mu} \\
 &\quad \times \exp \left\{ \frac{-2\mu}{2(1 - \delta^2)} [\rho_1^2(1 + \kappa_1) + \rho_2^2(1 + \kappa_2) + \kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}] \right\} \\
 &\quad \times \sum_{k=0}^\infty \frac{(\mu + k - 1)(2\mu - 2)_k}{k!} I_{\mu-1+k} \left(\frac{2\delta\mu\rho_1\rho_2\sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right) \\
 &\quad \times I_{\mu-1+k} \left(\frac{2\mu\rho_2\sqrt{1 + \kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right) \\
 &\quad \times I_{\mu-1+k} \left(\frac{2\mu\rho_1\sqrt{1 + \kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right) d\rho_1 d\rho_2 \\
 &= \int_0^\infty \int_0^\infty \rho_1^{n_1} \rho_2^{n_2} \sum_{z=0}^\infty \sum_{k=0}^\infty \sum_{l=0}^\infty \sum_{w=0}^\infty \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1) (1 + \kappa_2)}{[\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^\mu} \\
 &\quad \times \frac{(\mu + k - 1)(2\mu - 2)_k}{k!z!w!l!\Gamma(\mu - 1 + k + l + 1)\Gamma(\mu - 1 + k + w + 1)\Gamma(\mu - 1 + k + z + 1)} \\
 &\quad \times \exp \left\{ \frac{-\mu}{(1 - \delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}] \right\} \exp \left\{ \frac{-\mu\rho_1^2(1 + \kappa_1)}{(1 - \delta^2)} \right\} \exp \left\{ \frac{-\mu\rho_2^2(1 + \kappa_2)}{(1 - \delta^2)} \right\} \\
 &\quad \times \left(\frac{\mu\rho_1\sqrt{1 + \kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{\mu^2\rho_1^2(1 + \kappa_1)(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1 - \delta^2)^2} \right)^l \\
 &\quad \times \left(\frac{\mu\rho_2\sqrt{1 + \kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{\mu^2\rho_2^2(1 + \kappa_2)(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1 - \delta^2)^2} \right)^w \\
 &\quad \times \left(\frac{\delta\mu\rho_1\rho_2\sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{\delta^2\mu^2\rho_1^2\rho_2^2(1 + \kappa_1)(1 + \kappa_2)}{(1 - \delta^2)^2} \right)^z d\rho_1 d\rho_2.
 \end{aligned}$$

Following a method similar to the derivation of the jcdf the joint arbitrary moments are obtained. Using (2.13) and interchange the summation and integral it follows that

4. BIVARIATE MODELS
4.2. The bivariate $\kappa - \mu$ distribution

$$\begin{aligned}
E[P_1^{n_1} P_2^{n_2}] &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \int_0^{\infty} \int_0^{\infty} \frac{4\mu^2 \Gamma(\mu-1) (\mu+k-1) (2\mu-2)_k (1-\delta^2)^{2\mu-3}}{k!z!w!l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
&\quad \times \frac{(1+\kappa_1)(1+\kappa_2) \exp\left\{\frac{-\mu}{(1-\delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}]\right\}}{[\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \\
&\quad \times \rho_1^{n_1+1} \rho_2^{n_2+1} \exp\left\{\frac{-\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)}\right\} \exp\left\{\frac{-\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)}\right\} \\
&\quad \times \left(\frac{\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \\
&\quad \times \left(\frac{\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu^2\rho_2^2(1+\kappa_2)(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w \\
&\quad \times \left(\frac{\mu^2\rho_1^2(1+\kappa_1)(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \left(\frac{\delta^2\mu^2\rho_1^2\rho_2^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2}\right)^z d\rho_1 d\rho_2.
\end{aligned}$$

Further simplification results in

$$\begin{aligned}
E[P_1^{n_1} P_2^{n_2}] &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \int_0^{\infty} \int_0^{\infty} \frac{4\mu^2 \Gamma(\mu-1) (\mu+k-1) (2\mu-2)_k (1-\delta^2)^{2\mu-3}}{k!z!w!l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
&\quad \times \frac{(1+\kappa_1)(1+\kappa_2) \exp\left\{\frac{-\mu}{(1-\delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}]\right\}}{[\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \exp\left\{\frac{-\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)}\right\} \\
&\quad \times \rho_1^{n_1+1} \rho_2^{n_2+1} \exp\left\{\frac{-\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)}\right\} \left(\frac{\mu\rho_1^2(1+\kappa_1)}{1-\delta^2}\right)^{l+k+z+\mu+\frac{n_1}{2}-1} \\
&\quad \times \left(\frac{1-\delta^2}{\mu\rho_1^2(1+\kappa_1)}\right)^{l+k+z+\mu+\frac{n_1}{2}-1} \left(\frac{\mu\rho_2^2(1+\kappa_2)}{1-\delta^2}\right)^{w+k+z+\mu+\frac{n_2}{2}-1} \\
&\quad \times \left(\frac{1-\delta^2}{\mu\rho_2^2(1+\kappa_2)}\right)^{w+k+z+\mu+\frac{n_2}{2}-1} \left(\frac{\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)}\right)^{\mu-1+k} \\
&\quad \times \left(\frac{\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \\
&\quad \times \left(\frac{\mu^2\rho_2^2(1+\kappa_2)(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w \left(\frac{\mu^2\rho_1^2(1+\kappa_1)(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \\
&\quad \times \left(\frac{\delta^2\mu^2\rho_1^2\rho_2^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2}\right)^z d\rho_1 d\rho_2.
\end{aligned}$$

4. BIVARIATE MODELS

4.2. The bivariate $\kappa - \mu$ distribution

Therefore,

$$\begin{aligned}
 E[P_1^{n_1} P_2^{n_2}] &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \int_0^{\infty} \int_0^{\infty} \frac{\Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k}{k!z!w!l!\Gamma(\mu+k+l)\Gamma(\mu+k+w)\Gamma(\mu+k+z)} \\
 &\times \left(\frac{1-\delta^2}{\mu\rho_1^2(1+\kappa_1)}\right)^{l+k+z+\mu+\frac{n_1}{2}-1} \left(\frac{1-\delta^2}{\mu\rho_2^2(1+\kappa_2)}\right)^{w+k+z+\mu+\frac{n_2}{2}-1} \\
 &\times \left(\frac{\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\times \left(\frac{\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{\mu^2\rho_2^2(1+\kappa_2)(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w \\
 &\times \left(\frac{\mu^2\rho_1^2(1+\kappa_1)(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \left(\frac{\delta^2\mu^2\rho_1^2\rho_2^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2}\right)^z \\
 &\times \exp\left\{\frac{-\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)}\right\} \left(\frac{\mu\rho_1^2(1+\kappa_1)}{1-\delta^2}\right)^{l+k+z+\mu+\frac{n_1}{2}-1} \\
 &\times \exp\left\{\frac{-\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)}\right\} \left(\frac{\mu\rho_2^2(1+\kappa_2)}{1-\delta^2}\right)^{w+k+z+\mu+\frac{n_2}{2}-1} d\rho_1 d\rho_2.
 \end{aligned}$$

Simplifying the above expression and using (2.1) the joint arbitrary moments are,

$$\begin{aligned}
 E[P_1^{n_1} P_2^{n_2}] &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{\Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k(1-\delta^2)^\mu \delta^{k+2z}}{k!z!w!l!\Gamma(\mu+k+l)\Gamma(\mu+k+w)\Gamma(\mu+k+z)} \\
 &\times \exp\left\{\frac{-\mu}{(1-\delta^2)}[\kappa_1+\kappa_2-2\delta\sqrt{\kappa_1\kappa_2}]\right\} \left(\frac{\mu(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})^2}{(1-\delta^2)}\right)^{l+\frac{k}{2}} \\
 &\times \left(\frac{\mu(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^{w+\frac{k}{2}} \left(\frac{\mu(1+\kappa_1)}{(1-\delta^2)}\right)^{-\frac{n_1}{2}} \left(\frac{\mu(1+\kappa_2)}{(1-\delta^2)}\right)^{-\frac{n_2}{2}} \\
 &\times \Gamma\left(l+k+z+\mu+\frac{n_1}{2}\right) \Gamma\left(w+k+z+\mu+\frac{n_2}{2}\right). \tag{4.17}
 \end{aligned}$$

4.2.4 Outage Probability

The parameter of interest is the power, Ω_i , at each diversity branch i . For the dual branch selection combining (SC) scheme the outage probability is the jcdf of the power evaluated at $\vartheta_i = \vartheta$, with $\vartheta = \Omega_i$ and $\bar{\vartheta}_i = \frac{E_b}{N_0} E[R_i^2]$ being the average power (see Section 4.2.1). By expressing P_i in terms of R_i ,

$$P_i^2 = \frac{R_i^2}{E[R_i^2]} = \frac{R_i^2 \frac{E_b}{N_0}}{E[R_i^2] \frac{E_b}{N_0}} = \frac{\vartheta}{\bar{\vartheta}_i}.$$

4. BIVARIATE MODELS

4.3. The bivariate $\kappa - \mu$ type distribution

Thus, from equation (4.16) the outage probability is defined as

$$\begin{aligned}
 F_{SC}(\vartheta) &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{\Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^\mu \delta^{k+2z}}{k!z!l!w!\Gamma(\mu+k+l)\Gamma(\mu+k+w)\Gamma(\mu+k+z)} \\
 &\times \exp\left\{\frac{-\mu}{(1-\delta^2)}[\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}]\right\} \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{l+\frac{k}{2}} \\
 &\times \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{w+\frac{k}{2}} \gamma\left(l+k+z+\mu, \frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)}\right) \\
 &\times \gamma\left(w+k+z+\mu, \frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)}\right). \tag{4.18}
 \end{aligned}$$

Remark 4.1 This equation (4.18) corresponds to equation (21) of Villavicencio, et al., (2016) [33].

4.3 The bivariate $\kappa - \mu$ type distribution

The aim of this section is to derive a bivariate $\kappa - \mu$ type distribution with as the underlying model, the elliptical class. Exact formulae for the jpdf, jcdf and joint arbitrary moments will be explored. Special cases of the elliptical model will receive attention. A weight approximation as outlined by Arashi and Nadarajah (2016) [2] will also receive attention.

4.3.1 Description

Let R_1 and R_2 be 2 $\kappa - \mu$ envelopes then,

$$R_1^2 = \sum_{i=1}^{2\mu} X_i^2$$

$$R_2^2 = \sum_{i=1}^{2\mu} Y_i^2,$$

where X_i and Y_i are mutually independent elliptical random variables with mean equal to a and variances σ_1^2 and σ_2^2 respectively. [Note: $2\mu =$ number of clusters of multipath]. From (2.28) $X_i|t \sim N(0, a_t)$ and $Y_i|t \sim N(0, a_t)$ where $a_t = t^{-1}\sigma^2$. If $\{X|t, Y|t, A\}$ be 2μ - dimensional vectors then $X|t = [X_1|t \ X_2|t \ \dots \ X_{2\mu}|t]^T$, $Y|t = [Y_1|t \ Y_2|t \ \dots \ Y_{2\mu}|t]^T$ and $A = [a \ a \ \dots \ a]^T$. Let $V_i|t = [X_i|t \ Y_i|t]^T$ where $X_i|t$ and $Y_i|t$ are correlated if and only if $i = j$, $i = 1, 2, \dots, 2\mu$. Note that A is the vector of the dominant component of the cluster. The interest is the jpdf of the envelope R_1 and R_2 , as well as, the normalized envelopes, P_1 and P_2 with $P_i = \frac{R_i}{\sqrt{\hat{r}_i}}$ and \hat{r}_i as defined in (3.13).

4. BIVARIATE MODELS

4.3. The bivariate $\kappa - \mu$ type distribution

4.3.2 Derivation

The positive definite covariance matrix is

$$t^{-1}\Sigma = \begin{bmatrix} t^{-1}\Sigma_{11} & t^{-1}\Sigma_{12} \\ t^{-1}\Sigma_{21} & t^{-1}\Sigma_{22} \end{bmatrix} = \begin{bmatrix} t^{-1}\sigma_1^2 & \delta t^{-1}\sigma_1\sigma_2 \\ \delta t^{-1}\sigma_1\sigma_2 & t^{-1}\sigma_2^2 \end{bmatrix},$$

where δ is the correlation coefficient of the elliptical component,

$$\delta = \frac{t^{-1}\Sigma_{12}}{\sqrt{(t^{-1}\Sigma_{11})(t^{-1}\Sigma_{22})}} = \frac{\Sigma_{12}}{\sigma_1\sigma_2}$$

$$\therefore \Sigma_{12} = \delta\sigma_1\sigma_2.$$

The inverse of the covariance matrix is

$$\begin{aligned} (t^{-1}\Sigma)^{-1} &= \begin{bmatrix} (t^{-1}\Sigma_{11})^{-1} & (t^{-1}\Sigma_{12})^{-1} \\ (t^{-1}\Sigma_{21})^{-1} & (t^{-1}\Sigma_{22})^{-1} \end{bmatrix} = \begin{bmatrix} t^{-1}\sigma_1^2 & \delta t^{-1}\sigma_1\sigma_2 \\ \delta t^{-1}\sigma_1\sigma_2 & t^{-1}\sigma_2^2 \end{bmatrix}^{-1} \\ &= \frac{1}{t^{-2}\sigma_1^2\sigma_2^2 - t^{-2}\delta^2\sigma_1^2\sigma_2^2} \begin{bmatrix} t^{-1}\sigma_2^2 & -\delta t^{-1}\sigma_1\sigma_2 \\ -\delta t^{-1}\sigma_1\sigma_2 & t^{-1}\sigma_1^2 \end{bmatrix} \\ &= \frac{1}{t^{-2}\sigma_1^2\sigma_2^2(1-\delta^2)} \begin{bmatrix} t^{-1}\sigma_2^2 & -\delta t^{-1}\sigma_1\sigma_2 \\ -\delta t^{-1}\sigma_1\sigma_2 & \sigma_1^2 \end{bmatrix} \\ \Sigma^{-1} &= \frac{1}{(1-\delta^2)} \begin{bmatrix} t\sigma_1^{-2} & -\delta t\sigma_1^{-1}\sigma_2^{-1} \\ -\delta t\sigma_1^{-1}\sigma_2^{-1} & t\sigma_2^{-2} \end{bmatrix} \\ (t^{-1}\Sigma)^{-1} &= \begin{bmatrix} t\Sigma_{11}^{-1} & t\Sigma_{21}^{-1} \\ t\Sigma_{21}^{-1} & t\Sigma_{22}^{-1} \end{bmatrix}. \end{aligned} \quad (4.19)$$

The pdf of the multivariate elliptical distribution is given by (2.29). Thus, the jpdf of X and Y is given by

$$\begin{aligned} f_{X,Y}(X,Y) &= \prod_{i=1}^{2\mu} \int_0^\infty \frac{W(t)}{(2\pi) \det(t^{-1}\Sigma)^{\frac{1}{2}}} \exp\left(-\frac{1}{2}[(X_i - \mu)(Y_i - \mu)]^T (t^{-1}\Sigma)^{-1} \begin{bmatrix} (X_i - \mu) \\ (Y_i - \mu) \end{bmatrix}\right) dt \\ &= \prod_{i=1}^{2\mu} \int_0^\infty \frac{W(t)}{(2\pi) \det(t^{-1}\Sigma)^{\frac{1}{2}}} \exp\left(-\frac{1}{2}[V_i - C_i]^T (t^{-1}\Sigma)^{-1} [V_i - C_i]\right) dt \\ &= \prod_{i=1}^{2\mu} f_i(V_i) \\ &= \int_0^\infty \frac{W(t)}{(2\pi)^{2\mu} \det(t^{-1}\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} \sum_{i=1}^{2\mu} [V_i - C_i]^T (t^{-1}\Sigma)^{-1} [V_i - C_i]\right) dt, \end{aligned} \quad (4.20)$$

where $V_i = [X_i \ Y_i]^T$ and $C_i = [a \ a]^T$.

4. BIVARIATE MODELS

4.3. The bivariate $\kappa - \mu$ type distribution

Expanding the quadratic form and replacing the V_i by X_i and Y_i it follows that

$$\begin{aligned} & \sum_{i=1}^{2\mu} [V_i - C_i]^T (t^{-1}\Sigma)^{-1} [V_i - C_i] \\ &= \sum_{i=1}^{2\mu} [(X_i - a)(Y_i - a)] (t^{-1}\Sigma)^{-1} \begin{bmatrix} (X_i - a) \\ (Y_i - a) \end{bmatrix}. \end{aligned}$$

Then from the above and (4.19) it follows

$$\begin{aligned} & \sum_{i=1}^{2\mu} [(X_i - a)(Y_i - a)] \begin{bmatrix} t\Sigma_{11}^{-1} & t\Sigma_{12}^{-1} \\ t\Sigma_{21}^{-1} & t\Sigma_{22}^{-1} \end{bmatrix} \begin{bmatrix} (X_i - a) \\ (Y_i - a) \end{bmatrix} \\ &= \sum_{i=1}^{2\mu} [(X_i - a)t\Sigma_{11}^{-1} + (Y_i - a)t\Sigma_{21}^{-1}] (X_i - a) + [(X_i - a)t\Sigma_{21}^{-1} + (Y_i - a)t\Sigma_{22}^{-1}] (Y_i - a) \\ &= \sum_{i=1}^{2\mu} (X_i - a)t\Sigma_{11}^{-1}(X_i - a) + (Y_i - a)t\Sigma_{21}^{-1}(X_i - a) + (X_i - a)t\Sigma_{21}^{-1}(Y_i - a) \\ & \quad + (Y_i - a)t\Sigma_{22}^{-1}(Y_i - a) \\ &= \sum_{i=1}^{2\mu} X_i t\Sigma_{11}^{-1} X_i - X_i t\Sigma_{11}^{-1} a - a t\Sigma_{11}^{-1} X_i + a t\Sigma_{11}^{-1} a + Y_i t\Sigma_{21}^{-1} X_i - Y_i t\Sigma_{21}^{-1} a - a t\Sigma_{21}^{-1} X_i \\ & \quad + a t\Sigma_{21}^{-1} a + X_i t\Sigma_{21}^{-1} Y_i - X_i t\Sigma_{21}^{-1} a - a t\Sigma_{21}^{-1} X_i + a t\Sigma_{21}^{-1} a + Y_i t\Sigma_{22}^{-1} Y_i - Y_i t\Sigma_{22}^{-1} a \\ & \quad - a t\Sigma_{22}^{-1} Y_i + a t\Sigma_{22}^{-1} a \\ &= \sum_{i=1}^{2\mu} a (t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) a - a (t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) X_i - X_i (t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) a \\ & \quad - a (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) Y_i - Y_i (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) a + X_i t\Sigma_{11}^{-1} X_i + X_i t\Sigma_{21}^{-1} Y_i + Y_i t\Sigma_{21}^{-1} X_i + Y_i t\Sigma_{22}^{-1} Y_i. \end{aligned}$$

Since the inverse of the covariance matrix are constant the above simplifies to

$$\begin{aligned} & \sum_{i=1}^{2\mu} a (t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) a - a (t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) X_i - X_i (t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) a \\ & \quad - a (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) Y_i - Y_i (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) a + X_i t\Sigma_{11}^{-1} X_i + X_i t\Sigma_{21}^{-1} Y_i + Y_i t\Sigma_{21}^{-1} X_i + Y_i t\Sigma_{22}^{-1} Y_i \\ &= (t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A|^2 + t\Sigma_{11}^{-1} |X|^2 + t\Sigma_{22}^{-1} |Y|^2 \\ & \quad + \sum_{i=1}^{2\mu} [X_i t\Sigma_{21}^{-1} Y_i + Y_i t\Sigma_{21}^{-1} X_i - a (t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) X_i - X_i (t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) a \\ & \quad - a (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) Y_i - Y_i (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) a] \\ &= (t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A|^2 + t\Sigma_{11}^{-1} |X|^2 + t\Sigma_{22}^{-1} |Y|^2 + 2t\Sigma_{21}^{-1} XY \\ & \quad - 2(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) AX - 2(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) AY. \end{aligned} \tag{4.21}$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

Substituting (4.21) into (4.20) gives

$$\begin{aligned}
 & \int_0^\infty \frac{W(t)}{(2\pi)^{2\mu} \det(t^{-1}\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} \sum_{i=1}^{2\mu} [V_i - C_i]^T (t^{-1}\Sigma)^{-1} [V_i - C_i]\right) dt \\
 = & \int_0^\infty \frac{W(t)}{(2\pi)^{2\mu} \det(t^{-1}\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} [(t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A|^2 + t\Sigma_{11}^{-1}|X|^2 + t\Sigma_{22}^{-1}|Y|^2 + 2t\Sigma_{21}^{-1}XY]\right) \\
 & \times \exp\left(-\frac{1}{2} [-2(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1})AX - 2(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})AY]\right) dt \\
 = & \int_0^\infty \frac{W(t)}{(2\pi)^{2\mu} \det(t^{-1}\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} [(t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A|^2 + t\Sigma_{11}^{-1}|X|^2 + t\Sigma_{22}^{-1}|Y|^2]\right) \\
 & \times \exp[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1})AX + (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})AY - t\Sigma_{21}^{-1}XY] dt \\
 = & \int_0^\infty \frac{W(t)}{(2\pi)^{2\mu} \det(t^{-1}\Sigma)^{\frac{2\mu}{2}}} \exp\left(-\frac{1}{2} [(t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A|^2 + t\Sigma_{11}^{-1}|X|^2 + t\Sigma_{22}^{-1}|Y|^2]\right) \\
 & \times \exp[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1})AX + Y((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})A - t\Sigma_{21}^{-1}X)] dt.
 \end{aligned}$$

Hence, the jpdf can be written as follows

$$\begin{aligned}
 f_{X,Y}(X, Y) &= \int_0^\infty \frac{W(t)}{(2\pi)^{2\mu} \det(t^{-1}\Sigma)^\mu} \exp\left(-\frac{1}{2} [(t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A|^2 + t\Sigma_{11}^{-1}|X|^2 + t\Sigma_{22}^{-1}|Y|^2]\right) \\
 & \times \exp[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1})AX + Y((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})A - t\Sigma_{21}^{-1}X)] dt.
 \end{aligned}$$

As before in order to obtain the jpdf of the envelopes, an integration with respect to X and Y subject to the constraint $|X| = R_1$, and $|Y| = R_2$, is required (see (2.45)). That is

$$\begin{aligned}
 f_{X,Y}(X, Y) &= \int_0^\infty W(t) \frac{\exp\left(-\frac{1}{2} [(t\Sigma_{11}^{-1} + 2t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A|^2 + t\Sigma_{11}^{-1}|X|^2 + t\Sigma_{22}^{-1}|Y|^2]\right)}{(2\pi)^{2\mu} \det(t^{-1}\Sigma)^\mu} \\
 & \times \int_{|X|=R_1} \exp[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1})X^T A] ds_1 \\
 & \times \int_{|Y|=R_2} \exp[Y^T ((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})A - t\Sigma_{21}^{-1}X)] ds_2 dt. \tag{4.22}
 \end{aligned}$$

where ds_1 and ds_2 are the elements of the integration surface area (see (2.45)). The second integral in equation (4.22) can be expressed as

$$\begin{aligned}
 & \int_{|Y|} \exp[Y^T ((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})A - t\Sigma_{21}^{-1}X)] ds_2 \\
 = & (2\pi R_2)^{\frac{2\mu}{2}} |(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})A - t\Sigma_{21}^{-1}R_1|^{1-\mu} \\
 & \times I_{\frac{1}{2}(2\mu-2)}(R_2 |(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})A - t\Sigma_{21}^{-1}R_1|). \tag{4.23}
 \end{aligned}$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

Using (2.16), the expression (4.23) can be written as

$$\begin{aligned}
 & (2\pi R_2)^{\frac{2\mu}{2}} |(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) A - t\Sigma_{21}^{-1} R_1|^{1-\mu} I_{\frac{1}{2}(2\mu-2)}(R_2 |(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) A - t\Sigma_{21}^{-1} R_1|) \\
 = & \frac{(2\pi R_2)^\mu}{(R_2)^{1-\mu}} (R_2 |(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) A - t\Sigma_{21}^{-1} R_1|)^{1-\mu} I_{\frac{1}{2}(2\mu-2)}(R_2 |(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) A - t\Sigma_{21}^{-1} R_1|) \\
 = & \frac{(2\pi R_2)^\mu}{(R_2)^{1-\mu}} 2^{\mu-1} \Gamma(\mu-1) \sum_{k=0}^{\infty} (-1)^k (\mu-1+k) \frac{I_{\mu-1+k}((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A| R_2)}{(R_2)^{\mu-1} (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1})^{\mu-1} |A|^{\mu-1}} \\
 & \times \frac{I_{\mu-1+k}(t\Sigma_{21}^{-1} R_1 R_2)}{(t\Sigma_{21}^{-1})^{\mu-1} (R_1 R_2)^{\mu-1}} C_k^{\mu-1}(\cos \phi_1) \\
 & \times \frac{2^{2\mu-1} (\pi R_2)^\mu}{[(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) (t\Sigma_{21}^{-1}) R_2]^{\mu-1} |A|^{\mu-1} R_1^{\mu-1}} \Gamma(\mu-1) \sum_{k=0}^{\infty} (-1)^k (\mu-1+k) \\
 & \times I_{\mu-1+k}((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A| R_2) I_{\mu-1+k}(t\Sigma_{21}^{-1} R_1 R_2) C_k^{\mu-1}(\cos \phi_1).
 \end{aligned}$$

where ϕ_1 is the angle between X and A .

Thus, the jpdf of the envelopes is

$$\begin{aligned}
 & \int_0^\infty \frac{2^{2\mu-1} (\pi R_2)^\mu \Gamma(\mu-1) \exp\left(-\frac{t}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} R_1^2 + \Sigma_{22}^{-1} R_2^2]\right)}{(2\pi)^{2\mu} |t^{-1}\Sigma|^\mu [(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) (t\Sigma_{21}^{-1}) R_2]^{\mu-1} |A|^{\mu-1} R_1^{\mu-1}} \\
 & \times W(t) \sum_{k=0}^{\infty} (-1)^k (\mu-1+k) I_{\mu-1+k}((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A| R_2) I_{\mu-1+k}(t\Sigma_{21}^{-1} R_1 R_2) \\
 & \times \int_{|X|} \exp[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) X'A] C_k^{\mu-1}(\cos \phi_1) ds_1 dt. \tag{4.24}
 \end{aligned}$$

To evaluate the remaining integral in equation (4.24) express the integral in generalized spherical coordinates

$$\begin{aligned}
 & \int_{|X|} \exp[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) X'A] C_k^{\mu-1}(\cos \phi_1) ds_1 \\
 = & \int_{|X|} \exp[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A \cos \phi_1] C_k^{\mu-1}(\cos \phi_1) ds_1 \\
 = & R_1^{2\mu-1} \int_0^{2\pi} d\theta \int_0^\pi \exp[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A \cos \phi_1] C_k^{\mu-1}(\cos \phi_1) \sin^{2\mu-2} d\phi_1 \\
 & \times \prod_{k=2}^{2\mu-2} \int_0^\pi \sin^{2\mu-1-k} \phi_k d\phi_k. \tag{4.25}
 \end{aligned}$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

Applying (2.17)

$$\begin{aligned}
 & \int_0^\pi \exp \left[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A \cos \phi_1 \right] C_k^{\mu-1}(\cos \phi_1) \sin^{2\mu-2} d\phi_1 \\
 &= \frac{2^{\mu-1} \Gamma(\mu - 1 + \frac{1}{2}) \Gamma(\frac{1}{2}) \Gamma(2(\mu - 1) + k)}{((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A)^{\mu-1} k! \Gamma(2(\mu - 1))} I_{\mu-1+k}((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A) \quad (4.26)
 \end{aligned}$$

and using (2.18),

$$\begin{aligned}
 & R_1^{2\mu-1} \int_0^{2\pi} d\theta \prod_{k=2}^{2\mu-2} \int_0^\pi \sin^{2\mu-1-k} \phi_k d\phi_k \\
 &= \frac{2\pi^{\frac{1}{2}(2\mu-1)} R_1^{2\mu-1}}{\Gamma(\frac{1}{2}(2\mu-1))}, \quad (4.27)
 \end{aligned}$$

it follows from (4.25), (4.26) and (4.27) that

$$\begin{aligned}
 & \int_{|X|} \exp \left[(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) X A \right] C_k^{\mu-1}(\cos \phi_1) ds_1 \\
 &= \frac{2\pi^{\frac{1}{2}(2\mu-1)} R_1^{2\mu-1}}{\Gamma(\frac{1}{2}(2\mu-1))} \frac{2^{\mu-1} \Gamma(\mu - 1 + \frac{1}{2}) \Gamma(\frac{1}{2}) \Gamma(2(\mu - 1) + k)}{((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A)^{\mu-1} k! \Gamma(2(\mu - 1))} I_{\mu-1+k}((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A) \\
 &= \frac{2^\mu \pi^{\mu-\frac{1}{2}} \Gamma(\frac{1}{2}) R_1^{2\mu-1}}{((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A)^{\mu-1}} \frac{\Gamma(2\mu - 2 + k)}{k! \Gamma(2\mu - 2)} I_{\mu-1+k}((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A) \\
 &= \frac{2^\mu \pi^\mu R_1^{2\mu-1}}{((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A)^{\mu-1}} \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k}((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A). \quad (4.28)
 \end{aligned}$$

Substituting (4.28) into (4.24) the jpdf of the envelopes is

$$\begin{aligned}
 f_{R_1, R_2}(R_1, R_2) &= \int_0^\infty \frac{2^{2\mu-1} R_2 \Gamma(\mu - 1) \exp \left(-\frac{t}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} R_1^2 + \Sigma_{22}^{-1} R_2^2] \right)}{\pi^\mu \det(t^{-1}\Sigma)^\mu [(t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) (t\Sigma_{21}^{-1})]^{\mu-1} |A|^{\mu-1} R_1^{\mu-1}} \\
 &\quad \times \sum_{k=0}^\infty (-1)^k (\mu - 1 + k) I_{\mu-1+k}((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A| R_2) I_{\mu-1+k}(t\Sigma_{21}^{-1} R_1 R_2) \\
 &\quad \times \frac{2^\mu \pi^\mu R_1^{2\mu-1}}{((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A)^{\mu-1}} \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k}((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A) W(t) dt \\
 &= \int_0^\infty \frac{2^{\mu-1} R_1 R_2 \Gamma(\mu - 1) \exp \left(-\frac{t}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} R_1^2 + \Sigma_{22}^{-1} R_2^2] \right)}{\det(t^{-1}\Sigma)^\mu [(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) (t\Sigma_{21}^{-1}) |A|^2]^{\mu-1}} \\
 &\quad \times \sum_{k=0}^\infty (-1)^k (\mu - 1 + k) \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k}((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A| R_2) \\
 &\quad \times I_{\mu-1+k}(t\Sigma_{21}^{-1} R_1 R_2) I_{\mu-1+k}((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A) W(t) dt. \quad (4.29)
 \end{aligned}$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

Expression (4.29) can be written in terms of the coefficients of the inverse of the covariance matrix by using (4.19). Thus, the jpdf is

$$\begin{aligned}
 f_{R_1, R_2}(R_1, R_2) &= \int_0^\infty \frac{2^{\mu-1} R_1 R_2 \Gamma(\mu-1) \exp\left(-\frac{t}{2} [(\Sigma_{11}^{-1} + 2\Sigma_{21}^{-1} + \Sigma_{22}^{-1}) |A|^2 + \Sigma_{11}^{-1} R_1^2 + \Sigma_{22}^{-1} R_2^2]\right)}{\det(t^{-1}\Sigma)^\mu [(t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) (t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) (t\Sigma_{21}^{-1}) |A|^2]^\mu} \\
 &\times \sum_{k=0}^\infty (-1)^k (\mu-1+k) \binom{2\mu+k-3}{2\mu-3} I_{\mu-1+k}((t\Sigma_{21}^{-1} + t\Sigma_{22}^{-1}) |A| R_2) \\
 &\times I_{\mu-1+k}(t\Sigma_{21}^{-1} R_1 R_2) I_{\mu-1+k}((t\Sigma_{11}^{-1} + t\Sigma_{21}^{-1}) R_1 A) W(t) dt \\
 &= \int_0^\infty \sum_{k=0}^\infty W(t) \frac{2^{\mu-1} R_1 R_2 \Gamma(\mu-1) (-1)^k (\mu-1+k) \binom{2\mu+k-3}{2\mu-3}}{[t^{-2} \sigma_1^2 \sigma_2^2 (1-\delta^2)]^\mu \left[\frac{|A| \delta (1-\delta \frac{\sigma_2}{\sigma_1}) (1-\delta \frac{\sigma_1}{\sigma_2})}{-(1-\delta^2)^3 t^3 \sigma_1^3 \sigma_2^3} \right]^\mu} \\
 &\times \exp\left\{-\frac{1}{2} \left(\frac{1}{(1-\delta^2)} \left[\frac{tR_1^2}{\sigma_1^2} + \frac{tR_2^2}{\sigma_2^2} + \left(\frac{t\sigma_1^2 + t\sigma_2^2 - 2\delta t\sigma_1\sigma_2}{\sigma_1^2 \sigma_2^2} \right) |A|^2 \right] \right)\right\} \\
 &\times I_{\mu-1+k} \left(\frac{t|A| R_2 (1-\delta \frac{\sigma_2}{\sigma_1})}{(1-\delta^2) \sigma_2^2} \right) I_{\mu-1+k} \left(\frac{t|A| R_1 (1-\delta \frac{\sigma_1}{\sigma_2})}{(1-\delta^2) \sigma_1^2} \right) \\
 &\times I_{\mu-1+k} \left(\frac{tR_1 R_2 \delta}{(1-\delta^2) \sigma_1 \sigma_2} \right) dt \\
 &= \int_0^\infty \frac{W(t) R_1 R_2 \Gamma(\mu-1)}{t^{-2\mu} (\sigma_1^2 \sigma_2^2)^\mu (1-\delta^2)^\mu} \left[\frac{-2(1-\delta^2)^3 \sigma_1^3 \sigma_2^3}{|A|^2 \delta t^3 (1-\delta \frac{\sigma_2}{\sigma_1}) (1-\delta \frac{\sigma_1}{\sigma_2})} \right]^\mu \\
 &\times \exp\left\{-\frac{1}{2} \left(\frac{1}{(1-\delta^2)} \left[\frac{tR_1^2}{\sigma_1^2} + \frac{tR_2^2}{\sigma_2^2} + \left(\frac{t\sigma_1^2 + t\sigma_2^2 - 2\delta t\sigma_1\sigma_2}{\sigma_1^2 \sigma_2^2} \right) |A|^2 \right] \right)\right\} \\
 &\times \sum_{k=0}^\infty (-1)^k (\mu-1+k) \binom{2\mu+k-3}{2\mu-3} I_{\mu-1+k} \left(\frac{tR_1 R_2 \delta}{(1-\delta^2) \sigma_1 \sigma_2} \right) \\
 &\times I_{\mu-1+k} \left(\frac{t|A| R_2 (1-\delta \frac{\sigma_2}{\sigma_1})}{(1-\delta^2) \sigma_2^2} \right) I_{\mu-1+k} \left(\frac{t|A| R_1 (1-\delta \frac{\sigma_1}{\sigma_2})}{(1-\delta^2) \sigma_1^2} \right) dt. \tag{4.30}
 \end{aligned}$$

Define $\kappa_i = \frac{|A|^2}{2\mu\sigma_i^2} = \frac{a^2}{\sigma_i^2}$, $i = 1, 2$. Then by normalizing the two $\kappa - \mu$ envelopes, R_1 and R_2 (see equation (3.13)) where $P_i = \frac{R_i}{\sqrt{\kappa_i}}$. The Jacobian of the transformation from $(R_1, R_2) \longrightarrow (P_1, P_2)$ is

$$\begin{aligned}
 |J((R_1, R_2) \longrightarrow (P_1, P_2))| &= \begin{vmatrix} \sigma_1 \sqrt{2\mu(1+\kappa_1)} & 0 \\ 0 & \sigma_2 \sqrt{2\mu(1+\kappa_2)} \end{vmatrix} \\
 &= \sigma_1 \sigma_2 2\mu \sqrt{(1+\kappa_1)(1+\kappa_2)}.
 \end{aligned}$$

4. BIVARIATE MODELS

4.3. The bivariate $\kappa - \mu$ type distribution

The final expression for the normalized jpdf of the envelope is

$$\begin{aligned}
 f_{P_1, P_2}(\rho_1, \rho_2) &= \int_0^\infty W(t) \frac{\Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} 2\mu\rho_1\rho_2\sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{t^{\mu-3} (\sigma_1\sigma_2) [\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \\
 &\quad \times \exp \left\{ \frac{-2\mu t}{2(1 - \delta^2)} [\rho_1^2(1 + \kappa_1) + \rho_2^2(1 + \kappa_2) + \kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}] \right\} \\
 &\quad \times \sum_{k=0}^{\infty} (-1)^k (\mu + k - 1) \binom{2\mu + k - 3}{2\mu - 3} I_{\mu-1+k} \left(\frac{2t\delta\mu\rho_1\rho_2\sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right) \\
 &\quad \times I_{\mu-1+k} \left(\frac{2t\mu\rho_2\sqrt{1 + \kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right) I_{\mu-1+k} \left(\frac{2t\mu\rho_1\sqrt{1 + \kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right) \\
 &\quad \times 2\mu\sigma_1\sigma_2\sqrt{(1 + \kappa_1)(1 + \kappa_2)} dt \\
 &= \int_0^\infty W(t) \frac{4\mu^2\rho_1\rho_2\Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1)(1 + \kappa_2)}{t^{\mu-3} [\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \\
 &\quad \times \exp \left\{ \frac{-\mu t}{(1 - \delta^2)} [\rho_1^2(1 + \kappa_1) + \rho_2^2(1 + \kappa_2) + \kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}] \right\} \\
 &\quad \times \sum_{k=0}^{\infty} \frac{(\mu + k - 1) (2\mu - 2)_k}{k!} I_{\mu-1+k} \left(\frac{2t\delta\mu\rho_1\rho_2\sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right) \\
 &\quad \times I_{\mu-1+k} \left(\frac{2t\mu\rho_2\sqrt{1 + \kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right) \\
 &\quad \times I_{\mu-1+k} \left(\frac{2t\mu\rho_1\sqrt{1 + \kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right) dt. \tag{4.31}
 \end{aligned}$$

4. BIVARIATE MODELS

4.3. The bivariate $\kappa - \mu$ type distribution

4.3.3 Characteristics

4.3.3.1 Joint cumulative distribution function

To find the jcdf of P_1 and P_2 , we expand the Bessel functions as infinite series using (2.13) and then integrate term-wise. Hence, the jpdf can be written as

$$f_{P_1, P_2}(\rho_1, \rho_2) = \int_0^\infty W(t) \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1) (1 + \kappa_2)}{t^{\mu-3} [\delta\mu (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \quad (4.32)$$

$$\times \exp \left\{ \frac{-\mu t}{(1 - \delta^2)} [\rho_1^2 (1 + \kappa_1) + \rho_2^2 (1 + \kappa_2) + \kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1 \kappa_2}] \right\} \quad (4.33)$$

$$\times \sum_{k=0}^\infty \frac{(\mu + k - 1) (2\mu - 2)_k}{k!} \sum_{z=0}^\infty \left(\frac{2t\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{2(1-\delta^2)} \right)^{\mu-1+k} \quad (4.34)$$

$$\times \frac{\left(\frac{1}{4} \left(\frac{2t\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)} \right)^2 \right)^z}{z! \Gamma(\mu - 1 + k + z + 1)} \sum_{w=0}^\infty \left(\frac{2t\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{2(1-\delta^2)} \right)^{\mu-1+k} \quad (4.35)$$

$$\times \frac{\left(\frac{1}{4} \left(\frac{2t\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^2 \right)^w}{w! \Gamma(\mu - 1 + k + w + 1)} \sum_{l=0}^\infty \left(\frac{2t\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{2(1-\delta^2)} \right)^{\mu-1+k} \quad (4.36)$$

$$\times \frac{\left(\frac{1}{4} \left(\frac{2t\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^2 \right)^l}{l! \Gamma(\mu - 1 + k + l + 1)} dt \quad (4.37)$$

$$= \int_0^\infty \sum_{z=0}^\infty \sum_{k=0}^\infty \sum_{l=0}^\infty \sum_{w=0}^\infty W(t) \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1) (1 + \kappa_2)}{t^{\mu-3} [\delta\mu (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \quad (4.38)$$

$$\times \exp \left\{ \frac{-\mu t}{(1 - \delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1 \kappa_2}) \right\}$$

$$\times \frac{(\mu + k - 1) (2\mu - 2)_k}{k! z! w! l! \Gamma(\mu - 1 + k + l + 1) \Gamma(\mu - 1 + k + w + 1) \Gamma(\mu - 1 + k + z + 1)}$$

$$\times \exp \left\{ \frac{-t\mu\rho_1^2(1 + \kappa_1)}{(1 - \delta^2)} \right\} \left(\frac{\rho_1^2 \mu t (1 + \kappa_1)}{1 - \delta^2} \right)^{l+k+z+\mu-1} \left(\frac{1 - \delta^2}{\rho_1^2 \mu t (1 + \kappa_1)} \right)^{l+k+z+\mu-1}$$

$$\times \exp \left\{ \frac{-t\mu\rho_2^2(1 + \kappa_2)}{(1 - \delta^2)} \right\} \left(\frac{\rho_2^2 \mu t (1 + \kappa_2)}{1 - \delta^2} \right)^{w+k+z+\mu-1} \left(\frac{1 - \delta^2}{\rho_2^2 \mu t (1 + \kappa_2)} \right)^{w+k+z+\mu-1}$$

$$\times \left(\frac{t\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{t^2 \mu^2 \rho_1^2 (1 + \kappa_1) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1 - \delta^2)^2} \right)^l$$

$$\times \left(\frac{t\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{t^2 \mu^2 \rho_2^2 (1 + \kappa_2) (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1 - \delta^2)^2} \right)^w$$

$$\times \left(\frac{t\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{t^2 \delta^2 \mu^2 \rho_1^2 \rho_2^2 (1 + \kappa_1) (1 + \kappa_2)}{(1 - \delta^2)^2} \right)^z dt. \quad (4.39)$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

Thus from (4.39), jcdf of P_1 and P_2 is

$$\begin{aligned}
 F_{P_1, P_2}(p_1, p_2) &= \int_0^{p_1} \int_0^{p_2} f_{\rho_1, \rho_2}(\rho_1, \rho_2) d\rho_1 d\rho_2 \\
 &= \int_0^{p_1} \int_0^{p_2} \int_0^\infty \sum_{z=0}^\infty \sum_{k=0}^\infty \sum_{l=0}^\infty \sum_{w=0}^\infty W(t) \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu - 3} (1 + \kappa_1) (1 + \kappa_2)}{t^{\mu - 3} [\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu - 1}} \\
 &\quad \times \exp\left\{\frac{-\mu t}{(1 - \delta^2)}(\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1 \kappa_2})\right\} \\
 &\quad \times \frac{(\mu + k - 1)(2\mu - 2)_k}{k!z!w!!\Gamma(\mu + k + l)\Gamma(\mu + k + w)\Gamma(\mu + k + z)} \\
 &\quad \times \exp\left\{\frac{-t\mu\rho_1^2(1 + \kappa_1)}{(1 - \delta^2)}\right\} \left(\frac{t^2\rho_1^2\mu(1 + \kappa_1)}{1 - \delta^2}\right)^{l+k+z+\mu-1} \\
 &\quad \times \left(\frac{1 - \delta^2}{t\rho_1^2\mu(1 + \kappa_1)}\right)^{l+k+z+\mu-1} \exp\left\{\frac{-t\mu\rho_2^2(1 + \kappa_2)}{(1 - \delta^2)}\right\} \\
 &\quad \times \left(\frac{\rho_2^2\mu t(1 + \kappa_2)}{1 - \delta^2}\right)^{w+k+z+\mu-1} \left(\frac{1 - \delta^2}{t\rho_2^2\mu(1 + \kappa_2)}\right)^{w+k+z+\mu-1} \\
 &\quad \times \left(\frac{t\mu\rho_1\sqrt{1 + \kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)}\right)^{\mu-1+k} \left(\frac{t^2\mu^2\rho_1^2(1 + \kappa_1)(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1 - \delta^2)^2}\right)^l \\
 &\quad \times \left(\frac{t\mu\rho_2\sqrt{1 + \kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)}\right)^{\mu-1+k} \left(\frac{t^2\mu^2\rho_2^2(1 + \kappa_2)(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1 - \delta^2)^2}\right)^w \\
 &\quad \times \left(\frac{t\delta\mu\rho_1\rho_2\sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)}\right)^{\mu-1+k} \left(\frac{t^2\delta^2\mu^2\rho_1^2\rho_2^2(1 + \kappa_1)(1 + \kappa_2)}{(1 - \delta^2)^2}\right)^z dt d\rho_1 d\rho_2,
 \end{aligned}$$

4. BIVARIATE MODELS

4.3. The bivariate $\kappa - \mu$ type distribution

which simplifies to

$$\begin{aligned}
F_{P_1, P_2}(p_1, p_2) &= \int_0^{p_1} \int_0^{p_2} \int_0^\infty \sum_{z=0}^\infty \sum_{k=0}^\infty \sum_{l=0}^\infty \sum_{w=0}^\infty W(t) \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu-1) (1-\delta^2)^{2\mu-3} (1+\kappa_1) (1+\kappa_2)}{t^{\mu-3} [\delta\mu (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \\
&\times \frac{(\mu+k-1) (2\mu-2)_k}{k!z!w!l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
&\times \exp \left\{ \frac{-\mu t}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \left(\frac{1-\delta^2}{t\rho_1^2\mu(1+\kappa_1)} \right)^{l+k+z+\mu-1} \\
&\times \left(\frac{1-\delta^2}{t\rho_2^2\mu(1+\kappa_2)} \right)^{w+k+z+\mu-1} \left(\frac{t\mu\rho_1\sqrt{1+\kappa_1} (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{\mu-1+k} \\
&\times \left(\frac{t^2\mu^2\rho_1^2(1+\kappa_1) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2} \right)^l \left(\frac{t\mu\rho_2\sqrt{1+\kappa_2} (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{\mu-1+k} \\
&\times \left(\frac{t^2\mu^2\rho_2^2(1+\kappa_2) (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2} \right)^w \left(\frac{t\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)} \right)^{\mu-1+k} \\
&\times \left(\frac{t^2\delta^2\mu^2\rho_1^2\rho_2^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2} \right)^z \\
&\times \exp \left\{ -\frac{t\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)} \right\} \left(\frac{t\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)} \right)^{l+k+z+\mu-1} \\
&\times \exp \left\{ -\frac{t\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)} \right\} \left(\frac{t\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+z+\mu-1} dt d\rho_1 d\rho_2,
\end{aligned}$$

4. BIVARIATE MODELS

4.3. The bivariate $\kappa - \mu$ type distribution

and

$$\begin{aligned}
 F_{P_1, P_2}(p_1, p_2) &= \int_0^{p_1} \int_0^{p_2} \int_0^\infty \sum_{z=0}^\infty \sum_{k=0}^\infty \sum_{l=0}^\infty \sum_{w=0}^\infty W(t) \frac{4\mu^2 \Gamma(\mu-1) (1-\delta^2)^{2\mu-3} (1+\kappa_1)(1+\kappa_2)}{t^{\mu-3} [\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \\
 &\times \frac{(\mu+k-1)(2\mu-2)_k}{k!z!w!l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\times \exp\left\{\frac{-\mu t}{(1-\delta^2)}(\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2})\right\} \left(\frac{1-\delta^2}{t\mu(1+\kappa_1)}\right)^{l+k+z+\mu-1} \\
 &\times \left(\frac{1-\delta^2}{t\mu(1+\kappa_2)}\right)^{w+k+z+\mu-1} \left(\frac{t\mu\sqrt{1+\kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\times \left(\frac{t^2\mu^2(1+\kappa_1)(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \left(\frac{t\mu\sqrt{1+\kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\times \left(\frac{t^2\mu^2(1+\kappa_2)(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w \left(\frac{t\delta\mu\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\times \left(\frac{t^2\delta^2\mu^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2}\right)^z \left(\frac{t\mu(1+\kappa_1)}{(1-\delta^2)}\right)^{l+k+z+\mu-1} \\
 &\times \left(\frac{t\mu(1+\kappa_2)}{(1-\delta^2)}\right)^{w+k+z+\mu-1} \rho_1^{2\mu+2k+2z+2l-1} \rho_2^{2\mu+2k+2z+2w-1} \\
 &\times \exp\left\{-\frac{\rho_1^2 t\mu(1+\kappa_1)}{(1-\delta^2)}\right\} \exp\left\{-\frac{\rho_2^2 t\mu(1+\kappa_2)}{(1-\delta^2)}\right\} dt d\rho_1 d\rho_2.
 \end{aligned}$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

If we interchange the integrals then using (2.20), the jcdf is

$$\begin{aligned}
 F_{P_1, P_2}(\rho_1, \rho_2) &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} W(t) \frac{\mu^2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu - 3} (1 + \kappa_1) (1 + \kappa_2)}{t^{\mu - 3} [\delta \mu (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})]^{\mu - 1}} \\
 &\times \frac{(\mu + k - 1) (2\mu - 2)_k}{k! z! w! l! \Gamma(\mu + k + l) \Gamma(\mu + k + w) \Gamma(\mu + k + z)} \\
 &\times \exp \left\{ \frac{-\mu t}{(1 - \delta^2)} (\kappa_1 + \kappa_2 - 2\delta \sqrt{\kappa_1 \kappa_2}) \right\} \left(\frac{1 - \delta^2}{t\mu(1 + \kappa_1)} \right)^{l+k+z+\mu-1} \\
 &\times \left(\frac{1 - \delta^2}{t\mu(1 + \kappa_2)} \right)^{w+k+z+\mu-1} \left(\frac{t\mu\sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})}{(1 - \delta^2)} \right)^{\mu-1+k} \\
 &\times \left(\frac{t^2 \mu^2 (1 + \kappa_1) (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})^2}{(1 - \delta^2)^2} \right)^l \left(\frac{t\mu\sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})}{(1 - \delta^2)} \right)^{\mu-1+k} \\
 &\times \left(\frac{t^2 \mu^2 (1 + \kappa_2) (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})^2}{(1 - \delta^2)^2} \right)^w \left(\frac{t\delta\mu\sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right)^{\mu-1+k} \\
 &\times \left(\frac{t^2 \delta^2 \mu^2 (1 + \kappa_1) (1 + \kappa_2)}{(1 - \delta^2)^2} \right)^z \left(\frac{t\mu(1 + \kappa_1)}{(1 - \delta^2)} \right)^{l+k+z+\mu-1} \\
 &\times \left(\frac{t\mu(1 + \kappa_2)}{(1 - \delta^2)} \right)^{w+k+z+\mu-1} \left[\frac{t\mu(1 + \kappa_1)}{(1 - \delta^2)} \right]^{-(\mu+k+z+l)} \left[\frac{t\mu(1 + \kappa_2)}{(1 - \delta^2)} \right]^{-(\mu+k+z+w)} \\
 &\gamma \left(\mu + k + z + l, \frac{t\mu(1 + \kappa_1) \rho_1^2}{(1 - \delta^2)} \right) \gamma \left(\mu + k + z + w, \frac{t\mu(1 + \kappa_2) \rho_2^2}{(1 - \delta^2)} \right) dt.
 \end{aligned}$$

4. BIVARIATE MODELS

4.3. The bivariate $\kappa - \mu$ type distribution

Further simplification results in

$$\begin{aligned}
 F_{P_1, P_2}(\rho_1, \rho_2) &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} W(t) \frac{t^{k+l+w} \mu^{k+z+l+w-4} \Gamma(\mu-1) \delta^{k+2z} (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})^{2z+k}}{(1-\delta^2)^{2k+2z+l+w-1}} \\
 &\times \frac{(\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})^{2l+k} (\mu+k-1) (2\mu-2)_k}{k!z!w!l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\times \exp \left\{ \frac{-\mu t}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta \sqrt{\kappa_1 \kappa_2}) \right\} \\
 &\times \gamma \left(\mu+k+z+l, \frac{t\mu(1+\kappa_1)\rho_1^2}{(1-\delta^2)} \right) \gamma \left(\mu+k+z+w, \frac{t\mu(1+\kappa_2)\rho_2^2}{(1-\delta^2)} \right) dt \\
 &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} W(t) \frac{t^{k+l+w} \Gamma(\mu-1) (\mu+k-1) (2\mu-2)_k (1-\delta^2)^{\mu} \delta^{k+2z}}{k!z!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\times \exp \left\{ \frac{-t\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta \sqrt{\kappa_1 \kappa_2}) \right\} \left(\frac{\mu (\sqrt{\kappa_1} - \delta \sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \\
 &\times \left(\frac{\mu (\sqrt{\kappa_2} - \delta \sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \gamma \left(l+k+z+\mu, \frac{t\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)} \right) \\
 &\times \gamma \left(w+k+z+\mu, \frac{t\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)} \right) dt. \tag{4.40}
 \end{aligned}$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

4.3.3.2 Moments

The joint arbitrary moments can be found using the jpdf (4.31) as

$$\begin{aligned}
 E[P_1^{n_1} P_2^{n_2}] &= \int_0^\infty \int_0^\infty \rho_1^{n_1} \rho_2^{n_2} f_{\rho_1, \rho_2}(\rho_1, \rho_2) d\rho_1 d\rho_2 \\
 &= \int_0^\infty \int_0^\infty \int_0^\infty \rho_1^{n_1} \rho_2^{n_2} W(t) \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1) (1 + \kappa_2)}{t^{\mu-3} [\delta\mu (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^\mu} \\
 &\quad \times \exp \left\{ \frac{-\mu t}{(1 - \delta^2)} [\rho_1^2 (1 + \kappa_1) + \rho_2^2 (1 + \kappa_2) + \kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1 \kappa_2}] \right\} \\
 &\quad \times \sum_{k=0}^\infty \frac{(\mu + k - 1) (2\mu - 2)_k}{k!} I_{\mu-1+k} \left(\frac{2t\delta\mu\rho_1\rho_2\sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right) \\
 &\quad \times I_{\mu-1+k} \left(\frac{2t\mu\rho_2\sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right) \\
 &\quad \times I_{\mu-1+k} \left(\frac{2t\mu\rho_1\sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right) dt d\rho_1 d\rho_2 \\
 &= \int_0^\infty \int_0^\infty \int_0^\infty \rho_1^{n_1} \rho_2^{n_2} \sum_{z=0}^\infty \sum_{k=0}^\infty \sum_{l=0}^\infty \sum_{w=0}^\infty W(t) \frac{4\mu^2 \rho_1 \rho_2 \Gamma(\mu - 1) (1 - \delta^2)^{2\mu-3} (1 + \kappa_1) (1 + \kappa_2)}{t^{\mu-3} [\delta\mu (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1}) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^\mu} \\
 &\quad \times \frac{(\mu + k - 1) (2\mu - 2)_k}{k! z! w! l! \Gamma(\mu + k + l) \Gamma(\mu + k + w) \Gamma(\mu + k + z)} \\
 &\quad \times \exp \left\{ \frac{-\mu t}{(1 - \delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1 \kappa_2}] \right\} \exp \left\{ \frac{-t\mu\rho_1^2 (1 + \kappa_1)}{(1 - \delta^2)} \right\} \exp \left\{ \frac{-t\mu\rho_2^2 (1 + \kappa_2)}{(1 - \delta^2)} \right\} \\
 &\quad \times \left(\frac{t\mu\rho_1\sqrt{1 + \kappa_1} (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{t^2\mu^2\rho_1^2 (1 + \kappa_1) (\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1 - \delta^2)^2} \right)^l \\
 &\quad \times \left(\frac{t\mu\rho_2\sqrt{1 + \kappa_2} (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{t^2\mu^2\rho_2^2 (1 + \kappa_2) (\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1 - \delta^2)^2} \right)^w \\
 &\quad \times \left(\frac{\delta t \mu \rho_1 \rho_2 \sqrt{(1 + \kappa_1)(1 + \kappa_2)}}{(1 - \delta^2)} \right)^{\mu-1+k} \left(\frac{\delta^2 t^2 \mu^2 \rho_1^2 \rho_2^2 (1 + \kappa_1) (1 + \kappa_2)}{(1 - \delta^2)^2} \right)^z dt d\rho_1 d\rho_2.
 \end{aligned}$$

Following a method similar to the derivation of the jcdf the joint arbitrary moments are obtained. Using (2.13), and interchange the summation and integral, it follows that

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

$$\begin{aligned}
 E[P_1^{n_1} P_2^{n_2}] &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \int_0^{\infty} \int_0^{\infty} \int_0^{\infty} \frac{W(t) 4\mu^2 \Gamma(\mu-1) (\mu+k-1) (2\mu-2)_k (1-\delta^2)^{2\mu-3}}{t^{\mu-3} k! z! w! l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\quad \times \frac{(1+\kappa_1)(1+\kappa_2) \exp\left\{\frac{-\mu t}{(1-\delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}]\right\}}{[\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \times \rho_1^{n_1+1} \rho_2^{n_2+1} \\
 &\quad \times \exp\left\{\frac{-t\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)}\right\} \exp\left\{\frac{-t\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)}\right\} \left(\frac{\delta t\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\quad \times \left(\frac{t\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\quad \times \left(\frac{t\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{t^2\mu^2\rho_2^2(1+\kappa_2)(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w \\
 &\quad \times \left(\frac{t^2\mu^2\rho_1^2(1+\kappa_1)(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \left(\frac{t^2\delta^2\mu^2\rho_1^2\rho_2^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2}\right)^z d\rho_1 d\rho_2 dt \\
 &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \int_0^{\infty} \int_0^{\infty} \int_0^{\infty} \frac{W(t) 4\mu^2 \Gamma(\mu-1) (\mu+k-1) (2\mu-2)_k (1-\delta^2)^{2\mu-3}}{t^{\mu-3} k! z! w! l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\quad \times \frac{(1+\kappa_1)(1+\kappa_2) \exp\left\{\frac{-\mu t}{(1-\delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}]\right\}}{[\delta\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})]^{\mu-1}} \times \exp\left\{\frac{-t\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)}\right\} \\
 &\quad \times \rho_1^{n_1+1} \rho_2^{n_2+1} \exp\left\{\frac{-t\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)}\right\} \left(\frac{t\mu\rho_1^2(1+\kappa_1)}{1-\delta^2}\right)^{l+k+z+\mu+\frac{n_1}{2}-1} \\
 &\quad \times \left(\frac{1-\delta^2}{t\mu\rho_1^2(1+\kappa_1)}\right)^{l+k+z+\mu+\frac{n_1}{2}-1} \left(\frac{t\mu\rho_2^2(1+\kappa_2)}{1-\delta^2}\right)^{w+k+z+\mu+\frac{n_2}{2}-1} \\
 &\quad \times \left(\frac{1-\delta^2}{t\mu\rho_2^2(1+\kappa_2)}\right)^{w+k+z+\mu+\frac{n_2}{2}-1} \left(\frac{\delta t\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\quad \times \left(\frac{t\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{t\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\quad \times \left(\frac{t^2\mu^2\rho_2^2(1+\kappa_2)(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w \left(\frac{t^2\mu^2\rho_1^2(1+\kappa_1)(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \\
 &\quad \times \left(\frac{t^2\delta^2\mu^2\rho_1^2\rho_2^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2}\right)^z d\rho_1 d\rho_2 dt.
 \end{aligned}$$

Further simplification results in

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

$$\begin{aligned}
 E[P_1^{n_1} P_2^{n_2}] &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \int_0^{\infty} \int_0^{\infty} \int_0^{\infty} \frac{W(t) \Gamma(\mu-1) (\mu+k-1) (2\mu-2)_k}{t^{\mu-3} k! z! w! l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\times \frac{4\mu^2 (1-\delta^2)^{2\mu-3} (1+\kappa_1) (1+\kappa_2) \rho_1^{n_1+1} \rho_2^{n_2+1}}{[\delta\mu(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})]^{\mu-1}} \times \exp\left\{\frac{-\mu t}{(1-\delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}]\right\} \\
 &\times \left(\frac{1-\delta^2}{t\mu\rho_1^2(1+\kappa_1)}\right)^{l+k+z+\mu+\frac{n_1}{2}-1} \left(\frac{1-\delta^2}{t\mu\rho_2^2(1+\kappa_2)}\right)^{w+k+z+\mu+\frac{n_2}{2}-1} \\
 &\times \left(\frac{t\delta\mu\rho_1\rho_2\sqrt{(1+\kappa_1)(1+\kappa_2)}}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{t\mu\rho_1\sqrt{1+\kappa_1}(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{\mu-1+k} \\
 &\times \left(\frac{t\mu\rho_2\sqrt{1+\kappa_2}(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{\mu-1+k} \left(\frac{t^2\mu^2\rho_2^2(1+\kappa_2)(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^w \\
 &\times \left(\frac{t^2\mu^2\rho_1^2(1+\kappa_1)(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})^2}{(1-\delta^2)^2}\right)^l \left(\frac{t^2\delta^2\mu^2\rho_1^2\rho_2^2(1+\kappa_1)(1+\kappa_2)}{(1-\delta^2)^2}\right)^z \\
 &\times \exp\left\{\frac{-t\mu\rho_1^2(1+\kappa_1)}{(1-\delta^2)}\right\} \left(\frac{t\mu\rho_1^2(1+\kappa_1)}{1-\delta^2}\right)^{l+k+z+\mu+\frac{n_1}{2}-1} \\
 &\times \exp\left\{\frac{-t\mu\rho_2^2(1+\kappa_2)}{(1-\delta^2)}\right\} \left(\frac{t\mu\rho_2^2(1+\kappa_2)}{1-\delta^2}\right)^{w+k+z+\mu+\frac{n_2}{2}-1} d\rho_1 d\rho_2 dt.
 \end{aligned}$$

Simplifying the above expression and using (2.1) the following expression is obtained

$$\begin{aligned}
 E[P_1^{n_1} P_2^{n_2}] &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{W(t) \Gamma(\mu-1) (\mu+k-1) (2\mu-2)_k (1-\delta^2)^{\mu} \delta^{k+2z}}{k! z! w! l! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\times \exp\left\{\frac{-\mu t}{(1-\delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}]\right\} t^{2\mu+2l+2z+2w+3k-1} \\
 &\times \left(\frac{\mu(\sqrt{\kappa_1}-\delta\sqrt{\kappa_2})^2}{(1-\delta^2)}\right)^{l+\frac{k}{2}} \left(\frac{\mu(\sqrt{\kappa_2}-\delta\sqrt{\kappa_1})^2}{(1-\delta^2)^2}\right)^{w+\frac{k}{2}} \\
 &\times \left(\frac{\mu(1+\kappa_1)}{(1-\delta^2)}\right)^{-\frac{n_1}{2}} \left(\frac{\mu(1+\kappa_2)}{(1-\delta^2)}\right)^{-\frac{n_2}{2}} \\
 &\times \Gamma\left(l+k+z+\mu+\frac{n_1}{2}\right) \Gamma\left(w+k+z+\mu+\frac{n_2}{2}\right) dt. \tag{4.41}
 \end{aligned}$$

4.3.4 Outage Probability

As before (see Section (4.2.4)), the outage probability is defined, using (4.40), as

$$\begin{aligned}
 F_{SC}(\vartheta) &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} W(t) \frac{t^{k+l+w} \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^{\mu} \delta^{k+2z}}{k!z!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\times \exp \left\{ \frac{-t\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \\
 &\times \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \gamma \left(l+k+z+\mu, \frac{\vartheta t\mu(1+\kappa_1)}{\vartheta_1(1-\delta^2)} \right) \\
 &\times \gamma \left(w+k+z+\mu, \frac{\vartheta t\mu(1+\kappa_2)}{\vartheta_2(1-\delta^2)} \right) dt. \tag{4.42}
 \end{aligned}$$

4.3.4.1 Special cases

Normal case:

Assume the normal distribution weight function (2.35) is substituted in the outage probability function (4.42), then the outage probability (4.42) becomes,

$$\begin{aligned}
 F_{SC}(\vartheta) &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \delta(t-1) \frac{t^{k+l+w} \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^{\mu} \delta^{k+2z}}{k!z!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\times \exp \left\{ \frac{-t\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \\
 &\times \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \gamma \left(l+k+z+\mu, \frac{\vartheta t\mu(1+\kappa_1)}{\vartheta_1(1-\delta^2)} \right) \\
 &\times \gamma \left(w+k+z+\mu, \frac{\vartheta t\mu(1+\kappa_2)}{\vartheta_2(1-\delta^2)} \right) dt.
 \end{aligned}$$

Consider the result,

$$\begin{aligned}
 F_{SC}(\vartheta, x) &= \int_0^{\infty} \delta(x) F_{SC}(\vartheta, x+1) dx \\
 &= F_{SC}(\vartheta, 1).
 \end{aligned}$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

Let $q = t - 1$ then $t = q + 1$, we get that the outage probability is

$$\begin{aligned}
 F_{SC}(\vartheta, t) &= F_{SC}(\vartheta, q + 1) \\
 &= \int_0^\infty \sum_{z=0}^\infty \sum_{k=0}^\infty \sum_{l=0}^\infty \sum_{w=0}^\infty \delta(q) \frac{(q+1)^{k+l+w} \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^\mu \delta^{k+2z}}{k!z!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\quad \times \exp \left\{ \frac{-(q+1)\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \\
 &\quad \times \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \gamma \left(l+k+z+\mu, \frac{\vartheta(q+1)\mu(1+\kappa_1)}{\vartheta_1(1-\delta^2)} \right) \\
 &\quad \times \gamma \left(w+k+z+\mu, \frac{\vartheta(q+1)\mu(1+\kappa_2)}{\vartheta_2(1-\delta^2)} \right) dq \\
 &= F_{SC}(\vartheta, 1).
 \end{aligned}$$

Thus,

$$\begin{aligned}
 F_{SC}(\vartheta) &= \sum_{z=0}^\infty \sum_{k=0}^\infty \sum_{l=0}^\infty \sum_{w=0}^\infty \frac{\Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^\mu \delta^{k+2z}}{k!z!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\quad \times \exp \left\{ \frac{-\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \\
 &\quad \times \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \gamma \left(l+k+z+\mu, \frac{\vartheta\mu(1+\kappa_1)}{\vartheta_1(1-\delta^2)} \right) \\
 &\quad \times \gamma \left(w+k+z+\mu, \frac{\vartheta\mu(1+\kappa_2)}{\vartheta_2(1-\delta^2)} \right). \tag{4.43}
 \end{aligned}$$

Note that (4.43) is obtained by substituting the normal distribution weight function in the elliptical model yields the same result as the case when the underlying distribution is assumed to be normal (4.18).

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

t-distribution case:

Assume the t- distribution (with v degrees of freedom), therefore the weight function (2.36) is substituted in the outage probability function (4.42). Then the outage probability (4.42) becomes,

$$\begin{aligned}
 F_{SC}(\vartheta) &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{t^{k+l+w} \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^{\mu} \delta^{k+2z}}{k!z!!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\times \frac{v \left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right)}{2\Gamma\left(\frac{v}{2}\right)} \exp\left\{\frac{-t\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2})\right\} \\
 &\times \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{l+\frac{k}{2}} \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{w+\frac{k}{2}} \\
 &\times \gamma\left(l+k+z+\mu, \frac{\vartheta}{\vartheta_1} \frac{t\mu(1+\kappa_1)}{(1-\delta^2)}\right) \gamma\left(w+k+z+\mu, \frac{\vartheta}{\vartheta_2} \frac{t\mu(1+\kappa_2)}{(1-\delta^2)}\right) dt \\
 &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{v^{\frac{v}{2}} \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^{\mu} \delta^{k+2z}}{2^{\frac{v}{2}} k!z!!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z) \Gamma\left(\frac{v}{2}\right)} \\
 &\times \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)}\right)^{l+\frac{k}{2}} \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)}\right)^{w+\frac{k}{2}} \\
 &\times t^{k+l+w+\frac{v}{2}-1} \exp\left\{-t \left[\frac{v}{2} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2})\right]\right\} \\
 &\times \gamma\left(l+k+z+\mu, \frac{\vartheta}{\vartheta_1} \frac{t\mu(1+\kappa_1)}{(1-\delta^2)}\right) \gamma\left(w+k+z+\mu, \frac{\vartheta}{\vartheta_2} \frac{t\mu(1+\kappa_2)}{(1-\delta^2)}\right) dt.
 \end{aligned} \tag{4.44}$$

Evaluating the integral in equation (4.44) using (2.6) and (2.21), we obtain

$$\begin{aligned}
 &\int_0^{\infty} t^{k+l+w+\frac{v}{2}-1} \exp\left\{-t \left[\frac{v}{2} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2})\right]\right\} \\
 &\times \gamma\left(l+k+z+\mu, t \frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)}\right) \gamma\left(w+k+z+\mu, t \frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)}\right) dt \\
 &= \int_0^{\infty} \sum_{n=0}^{\infty} \frac{(-1)^n \left(t \frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)}\right)^{w+k+\mu+z+n}}{n!(w+k+\mu+n)} t^{k+l+w+\frac{v}{2}-1} \\
 &\times \exp\left\{-t \left[\frac{v}{2} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2})\right]\right\} \gamma\left(l+k+z+\mu, t \frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)}\right) dt
 \end{aligned}$$

4. BIVARIATE MODELS

 4.3. The bivariate $\kappa - \mu$ type distribution

$$\begin{aligned}
 &= \sum_{n=0}^{\infty} \frac{(-1)^n \left(\frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+\mu+z+n}}{n!(w+k+\mu+z+n)} \int_0^{\infty} t^{2w+2k+l+\mu+z+n+\frac{v}{2}-1} \\
 &\quad \times \exp \left\{ -t \left[\frac{v}{2} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right] \right\} \gamma \left(l+k+z+\mu, t \frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right) dt \\
 &= \sum_{n=0}^{\infty} \frac{(-1)^n \left(\frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+\mu+z+n}}{n!(w+k+\mu+z+n)} \frac{\Gamma \left(n + \frac{v}{2} + 2l + 2w + 3k + 3\mu + 3z \right)}{(l+k+\mu+z)} \\
 &\quad \times \frac{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right)^{l+k+\mu+z}}{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} + \frac{v}{2} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right)^{n+\frac{v}{2}+2l+2w+3k+3\mu+3z}} \\
 &\quad \times {}_2F_1 \left(1, n + \frac{v}{2} + 2l + 2w + 3k + 3\mu + 3z; l+k+\mu+z+1; \tau \right),
 \end{aligned}$$

where $\tau = \frac{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right)}{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} + \frac{v}{2} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right)}$. Thus, the outage probability is

$$\begin{aligned}
 F_{SC}(\vartheta) &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \frac{v^{\frac{v}{2}} \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^\mu \delta^{k+2z}}{2^{\frac{v}{2}} k! z! l! w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z) \Gamma\left(\frac{v}{2}\right)} \\
 &\quad \times \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \\
 &\quad \times \sum_{n=0}^{\infty} \frac{(-1)^n \left(\frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+\mu+z+n}}{n!(w+k+\mu+z+n)} \times \frac{\Gamma \left(\frac{v}{2} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right)}{(l+k+\mu+z)} \\
 &\quad \times \frac{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right)^{l+k+\mu+z}}{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} + \frac{v}{2} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right)^{n+\frac{v}{2}+2l+2w+3k+3\mu+3z}} \\
 &\quad \times {}_2F_1 \left(1, n + \frac{v}{2} + 2l + 2w + 3k + 3\mu + 3z; l+k+\mu+z+1; \tau \right),
 \end{aligned} \tag{4.45}$$

where $\frac{v}{2} > \frac{\mu(\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2})}{(1-\delta^2)}$.

4.4 Weight approximation function

An expansion for the jcdf (4.40) is given, based on the assumption that the weighting function admits the expansion. For the weight function (2.37), the outage probability (4.42) becomes,

$$\begin{aligned}
 F_{SC}(\vartheta) &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \left(\sum_{j=1}^{\infty} a_j t^j \right) \frac{t^{k+l+w} \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^{\mu} \delta^{k+2z}}{k!z!!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\quad \times \exp \left\{ \frac{-t\mu}{(1-\delta^2)} [\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}] \right\} \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \\
 &\quad \times \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \gamma \left(l+k+z+\mu, \frac{\vartheta}{\vartheta_1} \frac{t\mu(1+\kappa_1)}{(1-\delta^2)} \right) \\
 &\quad \times \gamma \left(w+k+z+\mu, \frac{\vartheta}{\vartheta_{21}} \frac{t\mu(1+\kappa_2)}{(1-\delta^2)} \right) dt \\
 &= \int_0^{\infty} \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \sum_{j=1}^{\infty} \frac{a_j \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^{\mu} \delta^{k+2z}}{k!z!!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\quad \times \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \\
 &\quad \times t^{k+l+w+j} \exp \left\{ \frac{-t\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \\
 &\quad \times \gamma \left(l+k+z+\mu, \frac{\vartheta}{\vartheta_1} \frac{t\mu(1+\kappa_1)}{(1-\delta^2)} \right) \gamma \left(w+k+z+\mu, \frac{\vartheta}{\vartheta_2} \frac{t\mu(1+\kappa_2)}{(1-\delta^2)} \right) dt \quad (4.46)
 \end{aligned}$$

Evaluating the integral in equation (4.46) using (2.6) and (2.21), we obtain

$$\begin{aligned}
 &\int_0^{\infty} t^{k+l+w+j} \exp \left\{ -\frac{t\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \\
 &\quad \times \gamma \left(l+k+z+\mu, t \frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right) \gamma \left(w+k+z+\mu, t \frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right) dt \\
 &= \int_0^{\infty} \sum_{n=0}^{\infty} \frac{(-1)^n \left(t \frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+\mu+z+n}}{n! (w+k+\mu+n)} t^{k+l+w+j} \\
 &\quad \times \exp \left\{ -\frac{t\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \gamma \left(l+k+z+\mu, t \frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right) dt
 \end{aligned}$$

4. BIVARIATE MODELS
 4.5. Performance Measures

$$\begin{aligned}
 &= \sum_{n=0}^{\infty} \frac{(-1)^n \left(\frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+\mu+z+n}}{n! (w+k+\mu+z+n)} \int_0^{\infty} t^{2k+2w+\mu+l+z+j+n} \\
 &\quad \times \exp \left\{ -\frac{t\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right\} \gamma \left(l+k+z+\mu, t \frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right) dt \\
 &= \sum_{n=0}^{\infty} \frac{(-1)^n \left(\frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+\mu+z+n}}{n! (w+k+\mu+z+n)} \frac{\Gamma(3k+2w+2\mu+2l+2z+j+n+1)}{(l+k+\mu+z)} \\
 &\quad \times \frac{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right)^{l+k+\mu+z}}{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right)^{3k+2w+2\mu+2l+2z+j+n+1}} \\
 &\quad \times {}_2F_1(1, 3k+2w+2\mu+2l+2z+j+n+1; l+k+\mu+z+1; \pi),
 \end{aligned}$$

where $\pi = \frac{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right)}{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right)}$. Thus, the outage probability is

$$\begin{aligned}
 F_{SC}(\vartheta) &= \sum_{z=0}^{\infty} \sum_{k=0}^{\infty} \sum_{l=0}^{\infty} \sum_{w=0}^{\infty} \sum_{j=1}^{\infty} \frac{a_j \Gamma(\mu-1)(\mu+k-1)(2\mu-2)_k (1-\delta^2)^\mu \delta^{k+2z}}{k!z!l!w! \Gamma(\mu+k+l) \Gamma(\mu+k+w) \Gamma(\mu+k+z)} \\
 &\quad \times \left(\frac{\mu(\sqrt{\kappa_1} - \delta\sqrt{\kappa_2})}{(1-\delta^2)} \right)^{l+\frac{k}{2}} \left(\frac{\mu(\sqrt{\kappa_2} - \delta\sqrt{\kappa_1})}{(1-\delta^2)} \right)^{w+\frac{k}{2}} \\
 &\quad \times \sum_{n=0}^{\infty} \frac{(-1)^n \left(\frac{\vartheta}{\vartheta_2} \frac{\mu(1+\kappa_2)}{(1-\delta^2)} \right)^{w+k+\mu+z+n}}{n! (w+k+\mu+z+n)} \frac{\Gamma(3k+2w+2\mu+2l+2z+j+n+1)}{(l+k+\mu+z)} \\
 &\quad \times \frac{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} \right)^{l+k+\mu+z}}{\left(\frac{\vartheta}{\vartheta_1} \frac{\mu(1+\kappa_1)}{(1-\delta^2)} - \frac{\mu}{(1-\delta^2)} (\kappa_1 + \kappa_2 - 2\delta\sqrt{\kappa_1\kappa_2}) \right)^{3k+2w+2\mu+2l+2z+j+n+1}} \\
 &\quad \times {}_2F_1(1, 3k+2w+2\mu+2l+2z+j+n+1; l+k+\mu+z+1; \pi),
 \end{aligned} \tag{4.47}$$

where $|\pi| < 1$.

4.5 Performance Measures

Following a similar approach as Villavicencio, et al.,(2016) [33], the bivariate $\kappa - \mu$ type model is introduced. Graphical displays of (4.31) for special cases ((2.35) and (2.36)) are

4. BIVARIATE MODELS

4.5. Performance Measures

shown below for $\mu = 0.5$ and $\kappa_1 = \kappa_2 = 5$, on left and $\mu = 2$ and $\kappa_1 = \kappa_2 = 1$, on right.

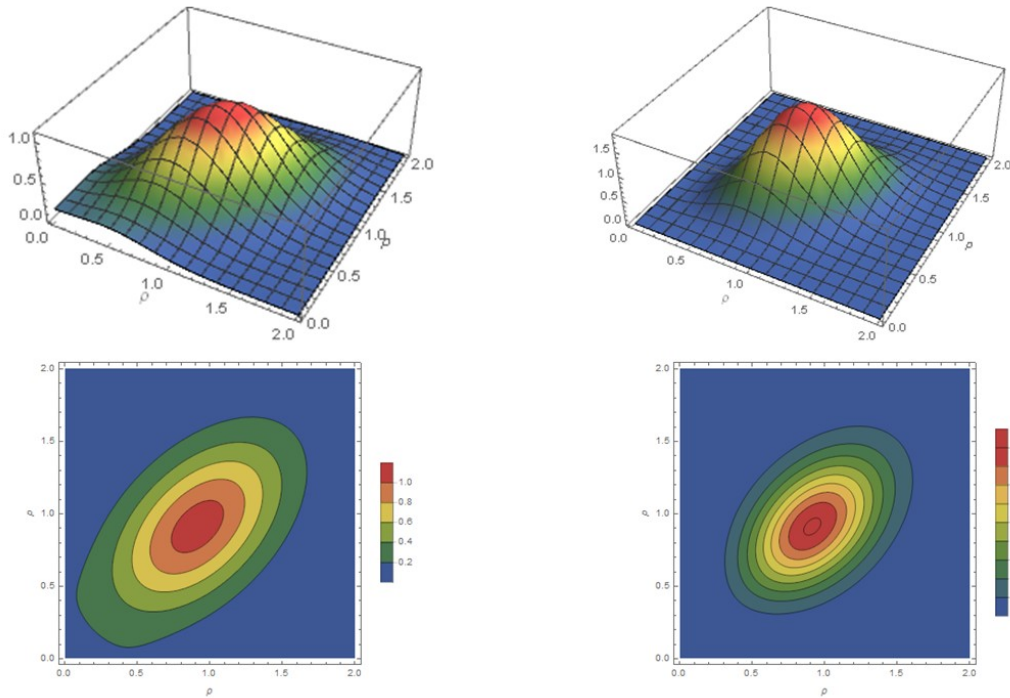


Figure 4.5.1 displays the normal case.

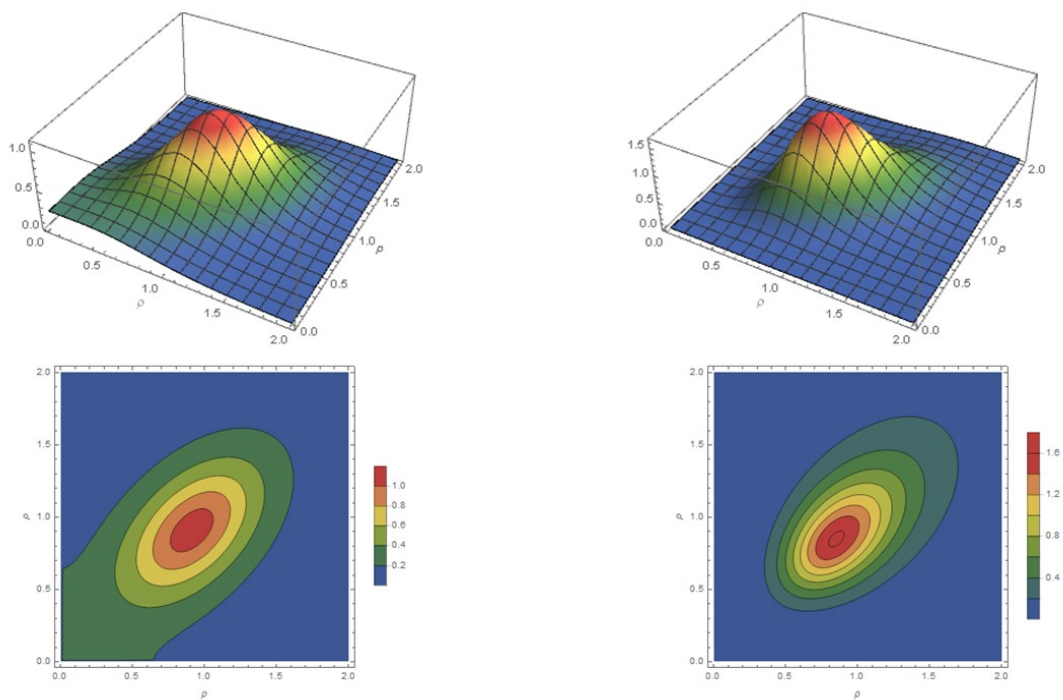


Figure 4.5.2 shows the t-distribution case, where $v = 3$ degrees of freedom were considered.

4. BIVARIATE MODELS

4.5. Performance Measures

The following figures illustrate the impact of the bivariate $\kappa - \mu$ -type distribution in the context of outage probability where the two cases of the elliptical model, namely normal distribution (2.35) and t-distribution (2.36) is considered. The effect of assuming the t-distribution as the underlying model is visible in Figure 4.5.3 and Figure 4.5.4 where the t-distribution outperforms the normal distribution. In the study of [11] it was shown that superior performance in terms of increased capacity of the communication system is observed when considering other distributions than the usual normal assumption.

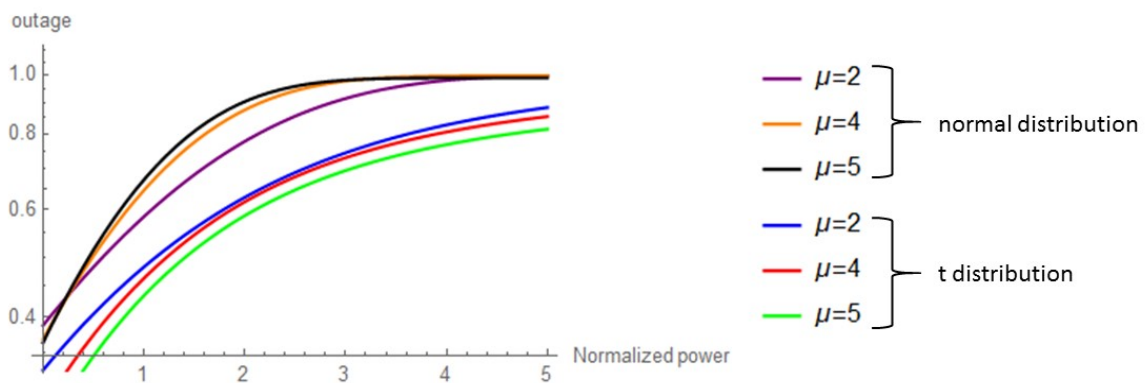


Figure 4.5.3 shows the outage probability for $\kappa_1 = \kappa_2 = 2$ and $\delta = 0.5$ for the normal case and t-distribution with $v = 3$ degrees of freedom.

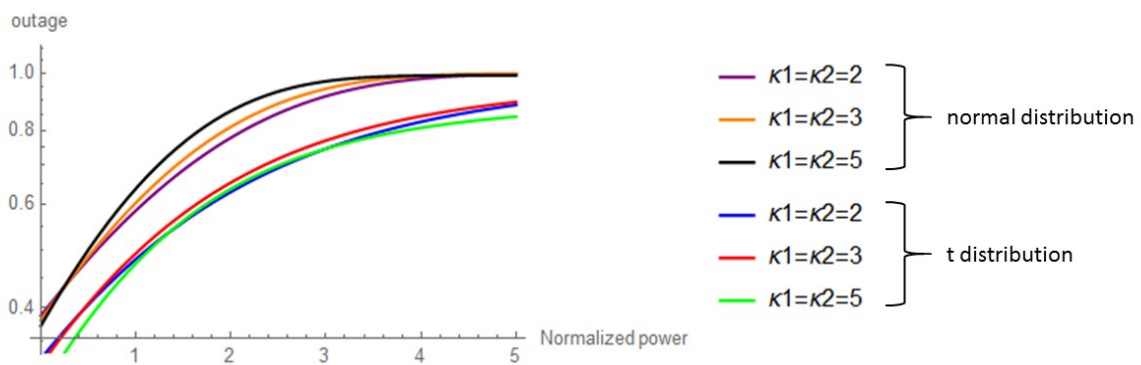


Figure 4.5.4 shows the outage probability for $\mu = 2$ and $\delta = 0.5$ for the normal case and t-distribution with $v = 3$ degrees of freedom.

Chapter 5

Composite models

5.1 Introduction

Composite models are used to account for the simultaneous presence of fading and shadowing (see Section 1.2.2). Shadowing can be incorporated in multipath fading models in various ways. The $\kappa - \mu$ shadowed fading model proposed by Paris (2014) [23] assumes the $\kappa - \mu$ distribution for the multipath fading and the Nakagami- m distribution for the shadowing. Right after Paris (2014) [23] in an independent work Cotton (2015) [7] proposed the same model. Figure 5.1 illustrates the physical model that motivates the development by Cotton (2015) [7] where shadowing events are caused not only by local environment but also by the user's body. The $\kappa - \mu$ shadowed fading distribution relies on a generalization of the physical model for the $\kappa - \mu$ distribution presented by Yacoub (2007) [35]. The $\kappa - \mu$ shadowed fading distribution has been derived assuming that the signals are structured in clusters of waves propagating in a nonhomogeneous environment. The $\kappa - \mu$ model Yacoub (2007) [35] assumes a fixed dominant component within each cluster whereas, the $\kappa - \mu$ shadowed model assumes that the dominant components of all the clusters can vary randomly as a result of shadowing.

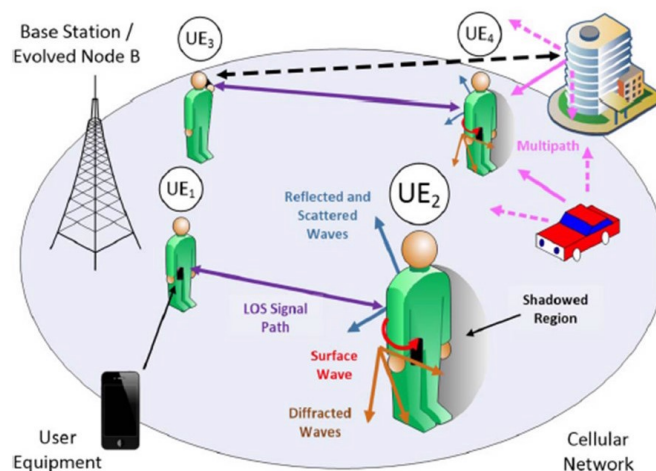


Figure 5.1: Illustration of structures associated with shadowing [7]

5. COMPOSITE MODELS

5.2. The $\kappa - \mu$ composite distribution

This chapter is organized as follows. In Section 5.2 the $\kappa - \mu$ shadowed fading model is characterised in terms of measurable physical parameters. The pdf of the SNR and the statistical properties of the $\kappa - \mu$ shadowed fading model will receive attention in Sections 5.2 and 5.3, respectively. The $\kappa - \mu$ type shadowed fading model that emanates from the elliptical assumption will be derived in Section 5.3; followed by Section 5.4 where the outage probability under the different members of the elliptical model will be compared.

5.2 The $\kappa - \mu$ composite distribution

5.2.1 Description

Let X_i and Y_i be mutually independent normal processes with $E(X_i) = E(Y_i) = 0$ and $var(X_i) = var(Y_i) = \sigma^2$. Then the envelope, R , of the physical model for the $\kappa - \mu$ distribution can be written in terms of the in-phase and quadrature components of the fading signal as

$$W = R^2 = \sum_{i=1}^n (X_i + \xi p_i)^2 + \sum_{i=1}^n (Y_i + \xi q_i)^2, \quad (5.1)$$

where ξp_i and ξq_i are the mean values of the in-phase and quadrature components of the multipath waves of cluster i and n is the number of clusters of multipath, i.e.

$$W_i = R_i = (X_i + \xi p_i)^2 + (Y_i + \xi q_i)^2, \quad i = 1, 2, \dots, n.$$

Define $\xi^2 \delta^2 = \sum_{i=1}^n (\xi^2 p_i^2 + \xi^2 q_i^2)$ to represent the mean power of the dominant component where we assume ξ to be a Nakagami- m random variable with shaping parameter m and $E(\xi^2) = \bar{\xi}$. For given ξ let $X_i \sim N(0, \sigma^2)$ and $Y_i \sim N(0, \sigma^2)$, thus $(X_i + \xi p_i) \sim N(\xi p_i, \sigma^2)$ and $X_i^* = \left(\frac{X_i + \xi p_i}{\sigma}\right) \sim N\left(\frac{\xi p_i}{\sigma}, 1\right)$. Therefore $(X_i^*)^2$ follows a $\chi_{\frac{2p_i^2}{\sigma^2}}^2(1)$ with non-centrality parameter $\frac{\xi^2 p_i^2}{\sigma^2}$. Similarly, for given ξ let $Y_i^* = \left(\frac{Y_i + \xi q_i}{\sigma}\right)$ and $(Y_i^*)^2$ follows a $\chi_{\frac{2q_i^2}{\sigma^2}}^2(1)$ with non-centrality parameter $\frac{\xi^2 q_i^2}{\sigma^2}$. As a result for a given ξ , $\frac{R_i^2}{\sigma^2}$ follows a $\chi_{\frac{2\delta_i^2}{\sigma^2}}^2(2)$ with non-centrality parameter $\frac{\xi^2 \delta_i^2}{\sigma^2} = \frac{\xi^2 p_i^2 + \xi^2 q_i^2}{\sigma^2}$. The focus is to derive the pdf of the power of the fading signal.

5.2.2 Derivation

For the given model (5.1), the pdf of the power, given the shadowing amplitude ξ , follows a $\kappa - \mu$ distribution with pdf (see equation (3.5))

$$f_{W|\xi}(w|\xi) = \frac{1}{2\sigma^2} \left(\frac{w}{\xi^2 \delta^2}\right)^{\frac{n-1}{2}} \exp\left\{-\frac{(w + \xi^2 \delta^2)}{2\sigma^2}\right\} I_{n-1}\left(\frac{\xi \delta}{\sigma^2} \sqrt{w}\right), \quad (5.2)$$

where $w > 0$.

5. COMPOSITE MODELS

 5.2. The $\kappa - \mu$ composite distribution

From (5.2) and (2.32) the unconditional pdf of the power is then obtained as follows

$$\begin{aligned}
 f_W(w) &= \int_0^{\infty} f_{W|\xi}(w|\xi) f_{\xi}(\xi) d\xi \\
 &= \int_0^{\infty} \frac{1}{2\sigma^2} \left(\frac{w}{\xi^2 \delta^2} \right)^{\frac{n-1}{2}} \exp \left\{ -\frac{(w + \xi^2 \delta^2)}{2\sigma^2} \right\} I_{n-1} \left(\frac{\xi \delta}{\sigma^2} \sqrt{w} \right) \\
 &\quad \times \frac{2m^m}{\theta^m \Gamma(m)} \xi^{2m-1} \exp \left\{ -\frac{m\xi^2}{\theta} \right\} d\xi \\
 &= \frac{m^m}{\sigma^2 \theta^m \Gamma(m)} \left(\frac{w}{\delta^2} \right)^{\frac{n-1}{2}} \exp \left\{ -\frac{w}{2\sigma^2} \right\} \int_0^{\infty} \xi^{2m-n} \\
 &\quad \exp \left\{ -\xi^2 \left(\frac{\delta^2}{2\sigma^2} + \frac{m}{\theta} \right) \right\} I_{n-1} \left(\frac{\xi \delta}{\sigma^2} \sqrt{w} \right) d\xi.
 \end{aligned}$$

Using (2.22), the unconditional pdf of the power is

$$\begin{aligned}
 f_W(w) &= \frac{m^m}{\sigma^2 \theta^m \Gamma(m)} \left(\frac{w}{\delta^2} \right)^{\frac{n-1}{2}} \exp \left\{ -\frac{w}{2\sigma^2} \right\} 2^{-n+1-1} \left(\frac{\delta \sqrt{w}}{\sigma^2} \right)^{n-1} \\
 &\quad \times \left(\frac{\delta^2}{2\sigma^2} + \frac{m}{\theta} \right)^{-\frac{(2m-n+1+n-1)}{2}} \Gamma \left[\frac{(2m-n+1+n-1)}{2} \right] \\
 &\quad \times {}_1F_1 \left(\frac{(2m-n+1+n-1)}{2}; n-1+1; \frac{\left(\frac{\delta \sqrt{w}}{\sigma^2} \right)^2}{4 \left(\frac{\delta^2}{2\sigma^2} + \frac{m}{\theta} \right)} \right) \\
 &= \frac{m^m \Gamma(m) w^{n-1}}{(2\sigma^2)^n \theta^m \Gamma(m) \Gamma(n) \left(\frac{\delta^2}{2\sigma^2} + \frac{m}{\theta} \right)^m} \exp \left\{ -\frac{w}{2\sigma^2} \right\} \\
 &\quad \times {}_1F_1 \left(m; n; \frac{\delta^2 w}{4\sigma^4 \left(\frac{\delta^2}{2\sigma^2} + \frac{m}{\theta} \right)} \right). \tag{5.3}
 \end{aligned}$$

As before, define the parameter $\kappa = \frac{\delta^2}{2\sigma^2\mu}$ to be the ratio between the total power of the dominant components and the total power of the scattered waves (see remark 3.3). Replace n with μ and let Ω represent the normalized power for the fading channel then $\Omega = \frac{W}{\bar{w}}$, where $\bar{w} = E(W) = \delta^2 + 2\sigma^2\mu = 2\sigma^2\mu(1 + \kappa)$ (see (3.8)). Thus, the pdf (5.3)

5. COMPOSITE MODELS

 5.2. The $\kappa - \mu$ composite distribution

can be written in terms of κ and μ as follows

$$\begin{aligned}
 f_{\Omega}(\omega) &= \frac{m^m \Gamma(m) (\omega \bar{w})^{\mu-1}}{(2\sigma^2)^{\mu} \theta^m \Gamma(m) \Gamma(\mu) \left(\frac{(2\sigma^2 \mu \kappa)}{2\sigma^2} + \frac{m}{\theta}\right)^m} \exp\left\{-\frac{(\omega \bar{w})}{2\sigma^2}\right\} \\
 &\times {}_1F_1\left(m; \mu; \frac{(2\sigma^2 \mu \kappa) (\omega \bar{w})}{4\sigma^4 \left(\frac{(2\sigma^2 \mu \kappa)}{2\sigma^2} + \frac{m}{\theta}\right)}\right) \bar{w} \\
 &= \frac{m^m \Gamma(m) (\omega (2\sigma^2 \mu (1 + \kappa)))^{\mu-1} (2\sigma^2 \mu (1 + \kappa))}{(2\sigma^2)^{\mu} \theta^m \Gamma(m) \Gamma(\mu) \left(\frac{(2\sigma^2 \mu \kappa)}{2\sigma^2} + \frac{m}{\theta}\right)^m} \exp\left\{-\frac{(\omega (2\sigma^2 \mu (1 + \kappa)))}{2\sigma^2}\right\} \\
 &\times {}_1F_1\left(m; \mu; \frac{(2\sigma^2 \mu \kappa) (\omega (2\sigma^2 \mu (1 + \kappa)))}{4\sigma^4 \left(\frac{(2\sigma^2 \mu \kappa)}{2\sigma^2} + \frac{m}{\theta}\right)}\right) \\
 &= \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu} \omega^{\mu-1}}{\theta^m \Gamma(\mu) (\mu \kappa + \frac{m}{\theta})^m} \exp\{-\mu (1 + \kappa) \omega\} {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) \omega}{(\mu \kappa + \frac{m}{\theta})}\right).
 \end{aligned} \tag{5.4}$$

5.2.3 Characteristics

5.2.3.1 Cumulative distribution function

The cdf of Ω is as follows

$$\begin{aligned}
 F_{\Omega}(Z_T) &= \int_0^{Z_T} \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu} \omega^{\mu-1}}{\theta^m \Gamma(\mu) (\mu \kappa + \frac{m}{\theta})^m} \exp\{-\mu (1 + \kappa) \omega\} {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) \omega}{(\mu \kappa + \frac{m}{\theta})}\right) d\omega \\
 &= \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu}}{\theta^m \Gamma(\mu) (\mu \kappa + \frac{m}{\theta})^m} \int_0^{Z_T} \omega^{\mu-1} \exp\{-\mu (1 + \kappa) \omega\} {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) \omega}{(\mu \kappa + \frac{m}{\theta})}\right) d\omega.
 \end{aligned}$$

Using (2.10) the confluent hypergeometric function can be expressed as an infinite series. Thus, the cdf can be written as

$$\begin{aligned}
 F_{\Omega}(Z_T) &= \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu}}{\theta^m \Gamma(\mu) (\mu \kappa + \frac{m}{\theta})^m} \int_0^{Z_T} \omega^{\mu-1} \exp\{-\mu (1 + \kappa) \omega\} \\
 &\times \sum_{l=0}^{\infty} \frac{(m)_l}{l! (\mu)_l} \left(\frac{\mu^2 \kappa (1 + \kappa) \omega}{(\mu \kappa + \frac{m}{\theta})}\right)^l d\omega \\
 &= \sum_{l=0}^{\infty} \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu} (m)_l}{\theta^m \Gamma(\mu) (\mu \kappa + \frac{m}{\theta})^m l! (\mu)_l} \left(\frac{\mu^2 \kappa (1 + \kappa)}{(\mu \kappa + \frac{m}{\theta})}\right)^l \\
 &\times \int_0^{Z_T} \omega^{\mu+l-1} \exp\{-\mu (1 + \kappa) \omega\} d\omega.
 \end{aligned}$$

5. COMPOSITE MODELS

 5.2. The $\kappa - \mu$ composite distribution

Using (2.20), the cdf becomes

$$\begin{aligned}
 F_{\Omega}(Z_T) &= \sum_{l=0}^{\infty} \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu} (m)_l}{\theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m l! (\mu)_l} \left(\frac{\mu^2 \kappa (1 + \kappa)}{(\mu\kappa + \frac{m}{\theta})} \right)^l \\
 &\quad \times [\mu(1 + \kappa)]^{-(\mu+l)} \gamma(\mu + l, \mu(1 + \kappa) Z_T) \\
 &= \sum_{l=0}^{\infty} \frac{m^m \mu^l \kappa^l (m)_l}{\theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^{m+l} l! (\mu)_l} \gamma(\mu + l, \mu(1 + \kappa) Z_T). \quad (5.5)
 \end{aligned}$$

5.2.3.2 Moments

From (5.4) the j^{th} moment of the normalized power, Ω , is

$$\begin{aligned}
 E(\Omega^j) &= \int_0^{\infty} \omega^j f_{\Omega}(\omega) d\gamma \\
 &= \int_0^{\infty} \omega^j \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu} \omega^{\mu-1}}{\theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m} \exp\{-\mu(1 + \kappa)\omega\} \\
 &\quad \times {}_1F_1\left(m; \mu+; \frac{\mu^2 \kappa (1 + \kappa) \omega}{(\mu\kappa + \frac{m}{\theta})}\right) d\omega \\
 &= \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu}}{\theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m} \int_0^{\infty} \omega^{\mu+j-1} \exp\{-\mu(1 + \kappa)\omega\} \\
 &\quad \times {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) \omega}{(\mu\kappa + \frac{m}{\theta})}\right) d\omega.
 \end{aligned}$$

Using (2.23), the j^{th} moment is

$$\begin{aligned}
 E(\Omega^j) &= \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu}}{\theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m} \times \frac{\Gamma(\mu + j)}{(\mu(1 + \kappa))^{\mu+j}} \\
 &\quad \times {}_2F_1\left(m, \mu + j; \mu; \frac{\left(\frac{\mu^2 \kappa (1 + \kappa)}{(\mu\kappa + \frac{m}{\theta})}\right)}{(\mu(1 + \kappa))}\right) \\
 &= \frac{m^m \Gamma(\mu + j)}{\theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m (\mu(1 + \kappa))^j} {}_2F_1\left(m, \mu + j; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right). \quad (5.6)
 \end{aligned}$$

5.2.3.3 Amount of fading

The amount of fading is given as (see equation (1.1))

$$\begin{aligned}
 AF &= \frac{E(\Omega^2)}{[E(\Omega)]^2} - 1 \\
 &= \frac{\left[\frac{m^m \Gamma(\mu+2)}{\theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m (\mu(1+\kappa))^2} {}_2F_1\left(m, \mu+2; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right) \right]}{\left[\frac{m^m \Gamma(\mu+1)}{\theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m (\mu(1+\kappa))} {}_2F_1\left(m, \mu+1; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right) \right]^2} - 1 \\
 &= \left(\frac{\theta^m (\mu+1) \left(\mu\kappa + \frac{m}{\theta}\right)^m {}_2F_1\left(m, \mu+2; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right)}{m^m \mu \left[{}_2F_1\left(m, \mu+1; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right) \right]^2} \right) - 1. \quad (5.7)
 \end{aligned}$$

5.3 The $\kappa - \mu$ type composite model

5.3.1 Description

Let X_i and Y_i be mutually independent elliptical processes with $E(X_i) = E(Y_i) = 0$ and $var(X_i) = var(Y_i) = -2\Psi'(0)$. Then the envelope, R , of the physical model for the $\kappa - \mu$ distribution can be written in terms of the in-phase and quadrature components of the fading signal as

$$W = R^2 = \sum_{i=1}^n (X_i + \xi p_i)^2 + \sum_{i=1}^n (Y_i + \xi q_i)^2, \quad (5.8)$$

where ξp_i and ξq_i are the mean values of the in-phase and quadrature components of the multipath waves of cluster i and n is the number of clusters of multipath, i.e.

$$W_i = R_i^2 = (X_i + \xi p_i)^2 + (Y_i + \xi q_i)^2, \quad i = 1, 2, \dots, n.$$

Define $\xi^2 \delta^2 = \sum_{i=1}^n (\xi^2 p_i^2 + \xi^2 q_i^2)$ to represent the mean power of the dominant component where we assume ξ to be a Nakagami- m random variable with shaping parameter m and $E(\xi^2) = \bar{\xi}$. For given ξ let $X_i \sim E(0, \sigma^2)$ and $Y_i \sim E(0, \sigma^2)$, therefore from (2.28), $X_i|t \sim N(0, a_t)$ and $Y_i|t \sim N(0, a_t)$ where $a_t = t^{-1}\sigma^2$. Hence, $(X_i + \xi p_i) | (t, \xi) \sim N(\xi p_i, a_t)$ and $X_i^* | (t, \xi) = \left(\frac{X_i + \xi p_i}{\sqrt{a_t}} | (t, \xi) \right) \sim N\left(\frac{\xi p_i}{\sqrt{a_t}}, 1\right)$. Thus $(X_i^*)^2 | (t, \xi)$ follows a $\chi_{\frac{\xi^2 p_i^2}{a_t}}^2(1)$ with non-centrality parameter $\frac{\xi^2 p_i^2}{a_t}$. Similarly $Y_i^* | (t, \xi) = \left(\frac{Y_i + \xi q_i}{\sqrt{a_t}} | (t, \xi) \right) \sim N\left(\frac{\xi q_i}{\sqrt{a_t}}, 1\right)$ and $(Y_i^*)^2 | (t, \xi)$ follows a $\chi_{\frac{\xi^2 q_i^2}{a_t}}^2(1)$ with non-centrality parameter $\frac{\xi^2 q_i^2}{a_t}$. As a result $\left(\frac{R_i^2}{a_t} | (t, \xi) \right)$ follows a $\chi_{\frac{\xi^2 \delta_i^2}{a_t}}^2(2)$ with non-centrality parameter $\frac{\xi^2 \delta_i^2}{a_t} = \frac{\xi^2 p_i^2 + \xi^2 q_i^2}{a_t}$. The focus is to derive the pdf of the power of the fading signal $W = \sum_{i=1}^n W_i$.

5.3.2 Derivation

For the given model (5.8), the conditional pdf of the power, given the shadowing amplitude ξ , follows a $\kappa - \mu$ distribution with pdf (see equation (3.23))

$$f_{W|t,\xi}(w|t, \xi) = \frac{1}{2a_t} \left(\frac{w}{\xi^2 \delta^2} \right)^{\frac{n-1}{2}} \exp \left\{ -\frac{(w + \xi^2 \delta^2)}{2a_t} \right\} I_{n-1} \left(\frac{\xi \delta}{a_t} \sqrt{w} \right), \quad (5.9)$$

where $w > 0$.

From (5.9) the conditional pdf of the power is obtained as follows

$$\begin{aligned} f_{W|t}(w|t) &= \int_0^\infty f_{W|t,\xi}(w|t, \xi) f_\xi(\xi) d\xi \\ &= \int_0^\infty \frac{1}{2a_t} \left(\frac{w}{\xi^2 \delta^2} \right)^{\frac{n-1}{2}} \exp \left\{ -\frac{(w + \xi^2 \delta^2)}{2a_t} \right\} I_{n-1} \left(\frac{\xi \delta}{a_t} \sqrt{w} \right) \\ &\quad \times \frac{2m^m}{\theta^m \Gamma(m)} \xi^{2m-1} \exp \left\{ -\frac{m\xi^2}{\theta} \right\} d\xi \\ &= \frac{m^m}{a_t \theta^m \Gamma(m)} \left(\frac{w}{\delta^2} \right)^{\frac{n-1}{2}} \exp \left\{ -\frac{w}{2a_t} \right\} \int_0^\infty \xi^{2m-n} \\ &\quad \times \exp \left\{ -\xi^2 \left(\frac{\delta^2}{2a_t} + \frac{m}{\theta} \right) \right\} I_{n-1} \left(\frac{\xi \delta}{a_t} \sqrt{w} \right) d\xi. \end{aligned}$$

Using (2.22), the conditional pdf of the power is

$$\begin{aligned} f_{W|t}(w|t) &= \frac{m^m}{a_t \theta^m \Gamma(m)} \left(\frac{w}{\delta^2} \right)^{\frac{n-1}{2}} \exp \left\{ -\frac{w}{2a_t} \right\} 2^{-n+1-1} \\ &\quad \times \left(\frac{\delta \sqrt{w}}{a_t} \right)^{n-1} \left(\frac{\delta^2}{2a_t} + \frac{m}{\theta} \right)^{-\frac{(2m-n+1+n-1)}{2}} \Gamma \left[\frac{(2m-n+1+n-1)}{2} \right] \\ &\quad \times {}_1F_1 \left(\frac{(2m-n+1+n-1)}{2}; n-1+1; \frac{\left(\frac{\delta \sqrt{w}}{a_t} \right)^2}{4 \left(\frac{\delta^2}{2a_t} + \frac{m}{\theta} \right)} \right) \\ &= \frac{m^m \Gamma(m) w^{n-1}}{(2a_t)^n \theta^m \Gamma(m) \Gamma(n) \left(\frac{\delta^2}{2a_t} + \frac{m}{\theta} \right)^m} \exp \left\{ -\frac{w}{2a_t} \right\} \\ &\quad \times {}_1F_1 \left(m; n; \frac{\delta^2 w}{4a_t^2 \left(\frac{\delta^2}{2a_t} + \frac{m}{\theta} \right)} \right). \quad (5.10) \end{aligned}$$

Replace n with μ and define the parameter $\kappa = \frac{\delta^2}{2t^{-1}\sigma^2\mu}$. Let Ω_t represent the conditional normalized power for the fading channel then $\Omega_t \equiv \frac{w}{\bar{w}_t}$, where $\bar{w}_t \equiv E(W|t) = \delta^2 + 2t^{-1}\sigma^2\mu = 2t^{-1}\sigma^2\mu(1 + \kappa) = t^{-1}\bar{w}$ (see (3.26)). Thus, the pdf (5.10) can be written

5. COMPOSITE MODELS

 5.3. The $\kappa - \mu$ type composite model

in terms of κ and μ as

$$\begin{aligned}
 f_{\Omega_t}(\omega_t) &= \frac{m^m \Gamma(m) (\omega_t \bar{w}_t)^{\mu-1}}{(2a_t)^\mu \theta^m \Gamma(m) \Gamma(\mu) \left(\frac{2a_t \mu \kappa}{2a_t} + \frac{m}{\theta}\right)^m} \exp\left\{-\frac{(\omega_t \bar{w}_t)}{2a_t}\right\} \\
 &\times {}_1F_1\left(m; \mu; \frac{(2a_t \mu \kappa) (\omega_t \bar{w}_t)}{4a_t^2 \left(\frac{2a_t \mu \kappa}{2a_t} + \frac{m}{\theta}\right)}\right) \bar{w}_t \\
 &= \frac{m^m \Gamma(m) (\omega_t (2a_t \mu (1 + \kappa)))^{\mu-1} (2a_t \mu (1 + \kappa))}{(2a_t)^\mu \theta^m \Gamma(m) \Gamma(\mu) \left(\frac{2a_t \mu \kappa}{2a_t} + \frac{m}{\theta}\right)^m} \exp\left\{-\frac{(\omega_t (2a_t \mu (1 + \kappa)))}{2a_t}\right\} \\
 &\times {}_1F_1\left(m; \mu; \frac{(2a_t \mu \kappa) (\omega_t (2a_t \mu (1 + \kappa)))}{4a_t^2 \left(\frac{2a_t \mu \kappa}{2a_t} + \frac{m}{\theta}\right)}\right) \\
 &= \frac{m^m \mu^\mu (1 + \kappa)^\mu \omega_t^{\mu-1}}{\theta^m \Gamma(\mu) \left(\mu \kappa + \frac{m}{\theta}\right)^m} \exp\{-\mu (1 + \kappa) \omega_t\} \\
 &\times {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) \omega_t}{\left(\mu \kappa + \frac{m}{\theta}\right)}\right). \tag{5.11}
 \end{aligned}$$

Hence the unconditional pdf of the normalized power, $\Omega = \frac{W}{t^{-1}\bar{w}} = t\omega$ (see section 3.3) is given by

$$\begin{aligned}
 f_{\Omega}(\omega) &= \int_0^{\infty} W(t) f_{\Omega_t}(\omega_t) dt \\
 &= \int_0^{\infty} W(t) \frac{m^m \mu^\mu (1 + \kappa)^\mu (t\omega)^{\mu-1}}{\theta^m \Gamma(\mu) \left(\mu \kappa + \frac{m}{\theta}\right)^m} \exp\{-\mu (1 + \kappa) t\omega\} \\
 &\quad \times {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) t\omega}{\left(\mu \kappa + \frac{m}{\theta}\right)}\right) dt. \tag{5.12}
 \end{aligned}$$

Particular cases of $W(t)$ in equation (5.12) are focused on since it forms part of the investigation in section 5.4.

Normal case

Assume (2.35), then (5.12) simplifies to (5.4).

t-distribution

Assume (2.36), then it follows from (5.4) that

$$\begin{aligned}
 f_{\Omega}(\omega) &= \int_0^{\infty} \frac{v \left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right)}{2\Gamma\left(\frac{v}{2}\right)} \frac{m^m \mu^\mu (1 + \kappa)^\mu (t\omega)^{\mu-1}}{\theta^m \Gamma(\mu) \left(\mu \kappa + \frac{m}{\theta}\right)^m} \\
 &\quad \times \exp\{-\mu (1 + \kappa) t\omega\} {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) t\omega}{\left(\mu \kappa + \frac{m}{\theta}\right)}\right) dt
 \end{aligned}$$

$$\begin{aligned}
 &= \int_0^\infty \frac{v^{\frac{v}{2}} m^m \mu^\mu (1 + \kappa)^\mu \omega^{\mu-1}}{2^{\frac{v}{2}} \theta^m \Gamma(\mu) \Gamma\left(\frac{v}{2}\right) \left(\mu\kappa + \frac{m}{\theta}\right)^m} \\
 &\quad \times t^{\frac{v}{2} + \mu - 2} \exp\left\{-t\mu(1 + \kappa)\omega - \frac{tv}{2}\right\} \\
 &\quad \times {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) t\omega}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right) dt.
 \end{aligned}$$

Using (2.23), the pdf of the normalized power follows as

$$\begin{aligned}
 f_\Omega(\omega) &= \frac{v^{\frac{v}{2}} m^m \mu^\mu (1 + \kappa)^\mu \omega^{\mu-1}}{2^{\frac{v}{2}} \theta^m \Gamma(\mu) \Gamma\left(\frac{v}{2}\right) \left(\mu\kappa + \frac{m}{\theta}\right)^m} \Gamma\left(\frac{v}{2} + \mu - 1\right) \\
 &\quad \times \left[\mu(1 + \kappa)\omega + \frac{v}{2}\right]^{-\left(\frac{v}{2} + \mu - 1\right)} \\
 &\quad \times {}_2F_1\left(m; \frac{v}{2} + \mu - 1; \mu; \frac{\frac{\mu^2 \kappa (1 + \kappa) \omega}{\left(\mu\kappa + \frac{m}{\theta}\right)}}{\left[\mu(1 + \kappa)\omega + \frac{v}{2}\right]}\right), \quad (5.13)
 \end{aligned}$$

where $\omega > 0$.

5.3.3 Characteristics

5.3.3.1 Cumulative distribution function

The cdf of Ω is

$$\begin{aligned}
 F_\Omega(Z_T) &= \int_0^{Z_T} \int_0^\infty W(t) \frac{m^m \mu^\mu (1 + \kappa)^\mu t^{\mu-1}}{\theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m} \omega^{\mu-1} \exp\{-\mu(1 + \kappa)t\omega\} \\
 &\quad \times {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) t\omega}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right) dt d\omega. \quad (5.14)
 \end{aligned}$$

Subsequently special cases ((2.35) and (2.36)) for the outage probability will be derived.

Remark 5.1 For the normal distribution case with weight function (2.35), the outage probability is given by (5.5).

Remark 5.2 For the t -distribution case with weight function (2.36), the outage probability is obtained as follows:

$$\begin{aligned}
 F_\Omega(Z_T) &= \int_0^{Z_T} \int_0^\infty \frac{v\left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right)}{2\Gamma\left(\frac{v}{2}\right)} \frac{m^m \mu^\mu (1 + \kappa)^\mu t^{\mu-1}}{\theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m} \omega^{\mu-1} \\
 &\quad \times \exp\{-\mu(1 + \kappa)t\omega\} {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) t\omega}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right) dt d\omega.
 \end{aligned}$$

5. COMPOSITE MODELS

 5.3. The $\kappa - \mu$ type composite model

Using (2.10) the confluent hypergeometric function in equation can be expressed as an infinite series. Thus, the cdf can be written as

$$\begin{aligned}
 F_{\Omega}(Z_T) &= \int_0^{Z_T} \int_0^{\infty} \frac{v^{\frac{v}{2}} m^m \mu^{\mu} (1 + \kappa)^{\mu} \omega^{\mu-1} t^{\mu+\frac{v}{2}-2}}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m} \exp\left\{-\mu(1 + \kappa)t\omega - \frac{v}{2}t\right\} \\
 &\quad \times \sum_{l=0}^{\infty} \frac{(m)_l}{l! (\mu)_l} \left(\frac{\mu^2 \kappa (1 + \kappa) t\omega}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right)^l d\omega \\
 &= \int_0^{Z_T} \int_0^{\infty} \sum_{l=0}^{\infty} \frac{v^{\frac{v}{2}} m^m \mu^{\mu} (1 + \kappa)^{\mu}}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m} \left(\frac{\mu^2 \kappa (1 + \kappa)}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right)^l \\
 &\quad \times \omega^{\mu+l-1} t^{\mu+\frac{v}{2}+l-2} \exp\left\{-t\left[\mu(1 + \kappa)\omega + \frac{v}{2}\right]\right\} dt d\omega.
 \end{aligned}$$

Using (2.20), it follows

$$\begin{aligned}
 F_{\Omega}(Z_T) &= \int_0^{Z_T} \sum_{l=0}^{\infty} \frac{v^{\frac{v}{2}} m^m \mu^{\mu} (1 + \kappa)^{\mu}}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m} \left(\frac{\mu^2 \kappa (1 + \kappa)}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right)^l \\
 &\quad \times \omega^{\mu+l-1} \Gamma\left(\mu + \frac{v}{2} + l - 1\right) \left[\mu(1 + \kappa)\omega + \frac{v}{2}\right]^{-(\mu+\frac{v}{2}+l-1)} d\omega \\
 &= \int_0^{Z_T} \sum_{l=0}^{\infty} \frac{v^{\frac{v}{2}} m^m \mu^{\mu} (1 + \kappa)^{\mu} \left(\frac{\mu^2 \kappa (1 + \kappa)}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right)^l}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m \Gamma\left(\mu + \frac{v}{2} + l - 1\right)} \\
 &\quad \times \omega^{\mu+l-1} \left[\mu(1 + \kappa)\omega + \frac{v}{2}\right]^{-(\mu+\frac{v}{2}+l-1)} d\omega \\
 &= \int_0^{Z_T} \sum_{l=0}^{\infty} \frac{v^{\frac{v}{2}} m^m \mu^{\mu} (1 + \kappa)^{\mu} \left(\frac{\mu^2 \kappa (1 + \kappa)}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right)^l}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m \Gamma\left(\mu + \frac{v}{2} + l - 1\right)} \\
 &\quad \times \omega^{\mu+l-1} \left\{\frac{v}{2} \left[\frac{2\mu(1 + \kappa)\omega}{v} + 1\right]\right\}^{-(\mu+\frac{v}{2}+l-1)} d\omega \\
 &= \int_0^{Z_T} \sum_{l=0}^{\infty} \frac{v^{\frac{v}{2}} m^m \mu^{\mu} (1 + \kappa)^{\mu} \left(\frac{\mu^2 \kappa (1 + \kappa)}{\left(\mu\kappa + \frac{m}{\theta}\right)}\right)^l \left(\frac{v}{2}\right)^{-(\mu+\frac{v}{2}+l-1)}}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) \left(\mu\kappa + \frac{m}{\theta}\right)^m \Gamma\left(\mu + \frac{v}{2} + l - 1\right)} \\
 &\quad \times \omega^{\mu+l-1} \left[\frac{2\mu(1 + \kappa)\omega}{v} + 1\right]^{-(\mu+\frac{v}{2}+l-1)} d\omega.
 \end{aligned}$$

5. COMPOSITE MODELS

 5.3. The $\kappa - \mu$ type composite model

Using (2.24), the outage probability is

$$\begin{aligned}
 F_{\Omega}(Z_T) &= \sum_{l=0}^{\infty} \frac{v^{\frac{v}{2}} m^m \mu^{\mu} (1 + \kappa)^{\mu} \left(\frac{\mu^2 \kappa (1 + \kappa)}{\mu \kappa + \frac{m}{\theta}} \right)^l \left(\frac{v}{2} \right)^{-(\mu + \frac{v}{2} + l - 1)}}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) (\mu \kappa + \frac{m}{\theta})^m \Gamma\left(\mu + \frac{v}{2} + l - 1\right)} \\
 &\quad \times \frac{Z_T^{\mu+l}}{\mu+l} {}_2F_1\left(\mu + \frac{v}{2} + l - 1, \mu + l; 1 + \mu + l; -\frac{2\mu(1 + \kappa)}{v} Z_T\right) \\
 &= \sum_{l=0}^{\infty} \frac{\left(\frac{v}{2}\right)^{-\mu-l+1} m^m \mu^{\mu+2l} (1 + \kappa)^{\mu+l} \kappa^l Z_T^{\mu+l}}{\theta^m \Gamma\left(\frac{v}{2}\right) \Gamma(\mu) \Gamma\left(\mu + \frac{v}{2} + l - 1\right) (\mu \kappa + \frac{m}{\theta})^{m+l} (\mu + l)} \\
 &\quad \times {}_2F_1\left(\mu + \frac{v}{2} + l - 1, \mu + l; 1 + \mu + l; -\frac{2\mu(1 + \kappa) Z_T}{v}\right) \\
 &= \sum_{l=0}^{\infty} \frac{\left(\frac{v}{2}\right)^{-\mu-l+1} m^m \mu^{\mu+2l} (1 + \kappa)^{\mu+l} \kappa^l Z_T^{\mu+l} \left(1 - \frac{2\mu(1 + \kappa) Z_T}{v}\right)^{-\mu - \frac{v}{2} - l + 1}}{\theta^m \Gamma\left(\frac{v}{2}\right) \Gamma(\mu) \Gamma\left(\mu + \frac{v}{2} + l - 1\right) (\mu \kappa + \frac{m}{\theta})^{m+l} (\mu + l)} \\
 &\quad \times {}_2F_1\left(\mu + \frac{v}{2} + l - 1, \mu + l; 1 + \mu + l; \frac{2\mu(1 + \kappa) Z_T}{2\mu(1 + \kappa) Z_T + v}\right). \quad (5.15)
 \end{aligned}$$

5.3.3.2 Moments

From (5.12) the j^{th} moment of the normalized power, Ω , is

$$\begin{aligned}
 E(\Omega^j) &= \int_0^{\infty} \omega^j f_{\Omega}(\omega) d\omega \\
 &= \int_0^{\infty} \omega^j \int_0^{\infty} W(t) \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu} (t\omega)^{\mu-1}}{\theta^m \Gamma(\mu) (\mu \kappa + \frac{m}{\theta})^m} \exp\{-\mu(1 + \kappa)t\omega\} \\
 &\quad \times {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) (t\omega)}{(\mu \kappa + \frac{m}{\theta})}\right) dt d\omega \\
 &= \int_0^{\infty} W(t) \frac{m^m \mu^{\mu} (1 + \kappa)^{\mu} t^{\mu-1}}{\theta^m \Gamma(\mu) (\mu \kappa + \frac{m}{\theta})^m} \int_0^{\infty} \omega^{\mu+j-1} \\
 &\quad \times \exp\{-\mu(1 + \kappa)t\omega\} {}_1F_1\left(m; \mu; \frac{\mu^2 \kappa (1 + \kappa) t\omega}{(\mu \kappa + \frac{m}{\theta})}\right) d\omega
 \end{aligned}$$

5. COMPOSITE MODELS

 5.3. The $\kappa - \mu$ type composite model

Using (2.23), the j^{th} moment is

$$\begin{aligned}
 E(\Omega^j) &= \int_0^\infty W(t) \frac{m^m \mu^\mu (1+\kappa)^\mu t^{\mu-1}}{\theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m} \frac{\Gamma(\mu+j)}{(t\mu(1+\kappa))^{\mu+j}} \\
 &\quad \times {}_2F_1\left(m, \mu+j; \mu; \frac{\left(\frac{\mu^2 \kappa(1+\kappa)t}{(\mu\kappa + \frac{m}{\theta})}\right)}{(t\mu(1+\kappa))}\right) dt \\
 &= \int_0^\infty W(t) \frac{m^m \Gamma(\mu+j)}{(t)^{j+1} \theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m (\mu(1+\kappa))^j} \\
 &\quad \times {}_2F_1\left(m, \mu+j; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right) dt. \tag{5.16}
 \end{aligned}$$

Remark 5.3 Substituting the weight functions (2.35) and (2.36) into (5.16), respectively, the following expressions are derived.

(i) For the normal distribution from (5.16) the j^{th} moment is (see (5.6))

$$E(\Omega^j) = \frac{m^m \Gamma(\mu+j) {}_2F_1\left(m, \mu+j; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right)}{\theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m (\mu(1+\kappa))^j}.$$

(ii) For the t -distribution from (5.16) the j^{th} moment is

$$\begin{aligned}
 E(\Omega^j) &= \int_0^\infty \frac{v \left(\frac{vt}{2}\right)^{\frac{v}{2}-1} \exp\left(-\frac{vt}{2}\right)}{2\Gamma\left(\frac{v}{2}\right)} \frac{m^m \Gamma(\mu+j)}{(t)^{j+1} \theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m (\mu(1+\kappa))^j} \\
 &\quad \times {}_2F_1\left(m, \mu+j; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right) dt \\
 &= \int_0^\infty \frac{v^{\frac{v}{2}} m^m \Gamma(\mu+j) {}_2F_1\left(m, \mu+j; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right)}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m (\mu(1+\kappa))^j} \\
 &\quad \times t^{\frac{v}{2}-j-2} \exp\left(-\frac{vt}{2}\right) dt.
 \end{aligned}$$

Using (2.19),

$$\begin{aligned}
 E(\Omega^j) &= \frac{v^{\frac{v}{2}} m^m \Gamma(\mu+j) {}_2F_1\left(m, \mu+j; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right) \Gamma\left(\frac{v}{2} - j - 1\right)}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m (\mu(1+\kappa))^j \left(\frac{v}{2}\right)^{\frac{v}{2}-j-1}} \\
 &= \frac{v^{j+1} m^m \Gamma(\mu+j) \Gamma\left(\frac{v}{2} - j - 1\right) {}_2F_1\left(m, \mu+j; \mu; \frac{\mu\kappa}{\mu\kappa + \frac{m}{\theta}}\right)}{2^{j+1} \Gamma\left(\frac{v}{2}\right) \theta^m \Gamma(\mu) (\mu\kappa + \frac{m}{\theta})^m (\mu(1+\kappa))^j}, \tag{5.17}
 \end{aligned}$$

where $\text{Re}\left(\frac{v}{2} - j - 1\right) > 0$.

5.4 Performance Measures

In this section graphical displays of the composite $\kappa - \mu$ type model and some performance metrics will be shown.

The pdf for the normalized power (5.12) is shown Figure 5.4.1 for $\kappa = 2$, $m = 2$ and $\Omega = 1$, Figure 5.4.2 for $\mu = 0.5$, $m = 2$ and $\Omega = 1$, and Figure 5.4.3 for $\kappa = 2$, $\mu = 0.5$ and $\Omega = 1$ for both the special cases (normal (5.3), on left, and t-distribution, where $v = 3$ degrees of freedom was considered, (5.13), on right, respectively).

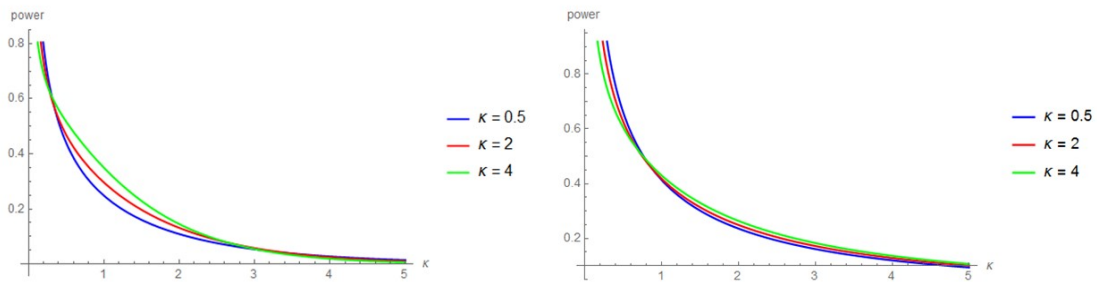


Figure 5.4.1

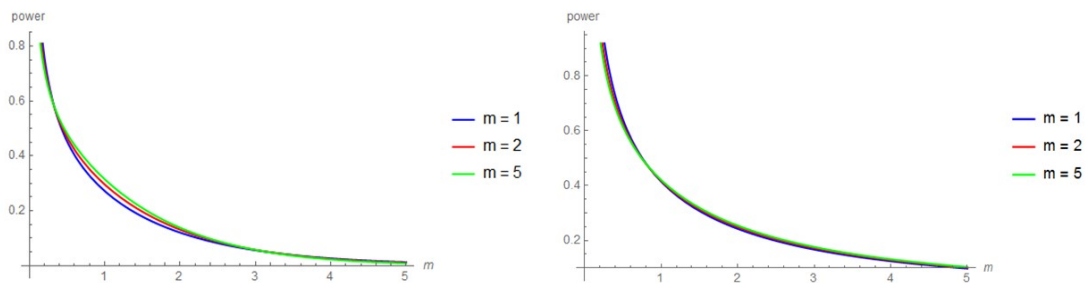


Figure 5.4.2

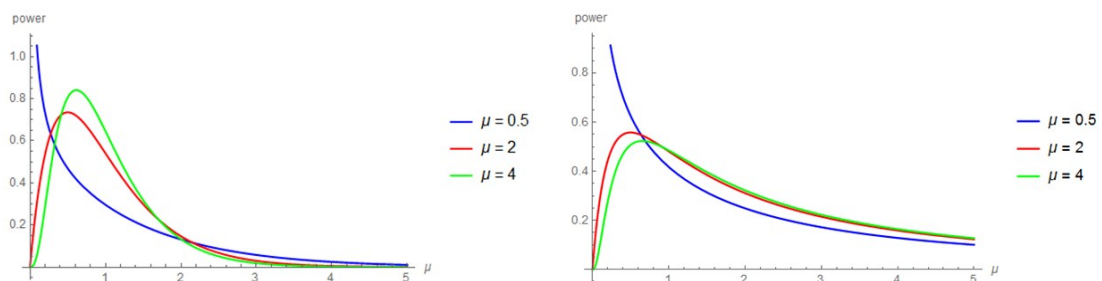


Figure 5.4.3

The outage probability (5.14) is illustrated in Figure 5.4.4 for $\kappa = 2$, $m = 2$ and $\Omega = 1$, Figure 5.4.5 for $\mu = 1$, $m = 2$ and $\Omega = 1$, and Figure 5.4.6 for $\kappa = 2$, $\mu = 2$ and

5. COMPOSITE MODELS

5.4. Performance Measures

$\Omega = 1$, respectively. Both the special cases normal (5.5), and t-distribution, where $v = 3$ degrees of freedom is considered, (5.15). For a fixed μ , the outage probability for the t-distribution outperforms the normal distribution as shown in Figure 5.4.4. For a fixed κ , the outage probability for the t-distribution only outperforms the normal distribution when μ (the number of clusters) is large as shown in Figure 5.4.5. Thus, as the number of clusters increase the assumption of the t-distribution as the underlying model is more appropriate.

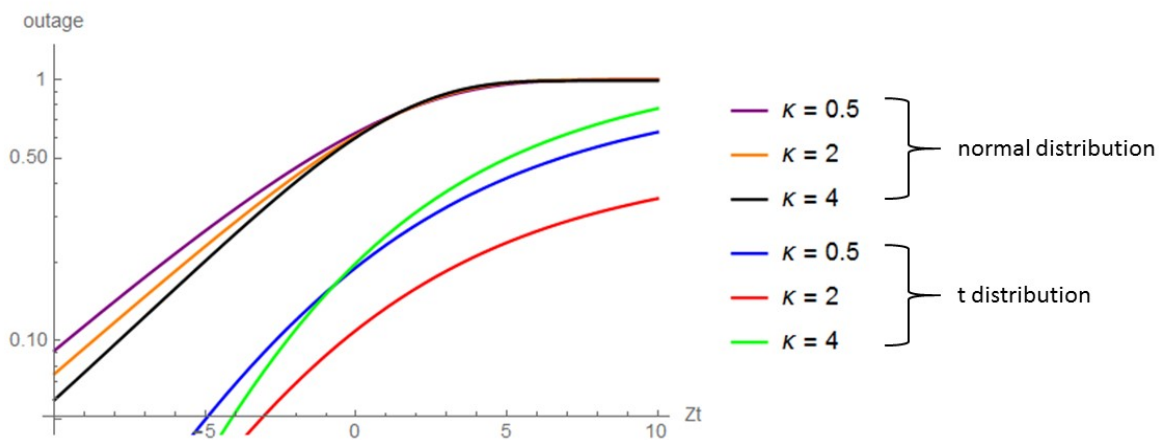


Figure 5.4.4

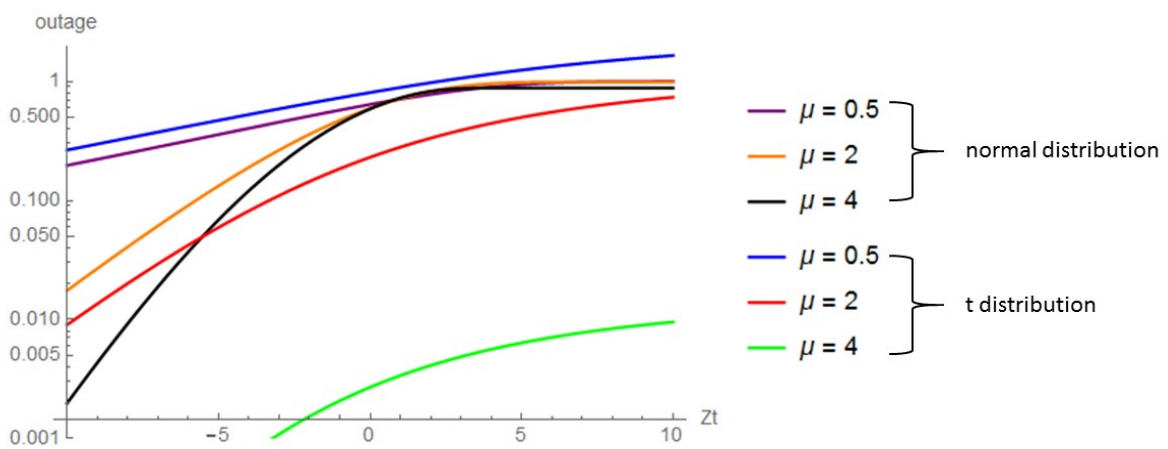


Figure 5.4.5

5. COMPOSITE MODELS
5.4. Performance Measures

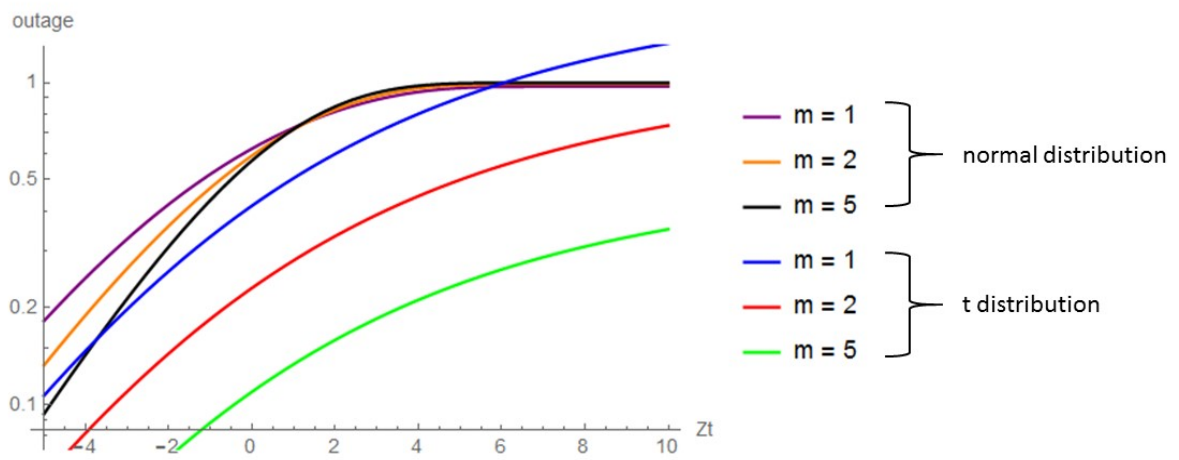


Figure 5.4.6

Chapter 6

Summary and future work

In this mini-dissertation the univariate and bivariate $\kappa - \mu$ type fading models as well as the composite $\kappa - \mu$ type fading model were developed; these distributions emanated from the elliptical assumption as the point of departure for the description of the envelope of the physical model. For each model the basic statistical characteristics and performance metrics were derived. By assuming the elliptical model various different distributions are included, but specific the t distribution is focused of this study, since Yacoub (2007) [35] commented about the inadequacy of tail fitting of some distributions to experimental data from the communications system domain.

Graphical illustrations accompanied these characteristics of the different models. The analytical framework that was used (stemming from an integral representation) provides computationally convenient forms of these distributions (Chu, 1973 [6]; Loots et al., 2013 [3]). By considering the t-distribution as a special member of the elliptical class, it was noted that the outage probability was lower in most cases. This has a significant impact on the assumption made by the researcher/practitioner in the communication systems field.

There are many opportunities for future research based on this study, to name a few:

- In a statistical context the univariate $\kappa - \mu$ type fading model can be further explored in terms of fitting this model to a real data case and investigating more characteristics, amongst others the skewness. Similarly for the bivariate $\kappa - \mu$ type model the correlation structure needs investigation. The physical contribution in the communications systems area needs to be investigated, the fit to field measurements in scenarios in this area, together with the study of other metrics such as the ergodic capacity (see Garcia-Corrales et al 2014 [12]).

- If the overall fading in a channel is the result of multiple scattering components, further research can include the assumption that the received SNR of the cascaded channel is expressed as the product of $\kappa - \mu$ type random variables. (See for the gamma case Shankar, 2011, p 232. [28])

- For the above scenario obtain derive, in a manageable form, asymptotic approximations for products and/or ratios of independent product of $\kappa - \mu$ type random variables which have pdf and cdf easily implementable and computationally appealing. (See Marques and Longeville, 2016 [17]; Nicolas and Florence Tupin, 2016 [20].)

- Extension of other models, for example the $\alpha - \kappa - \mu$ fading model with the elliptical assumption.

6. SUMMARY AND FUTURE WORK

- For the dependent components (X, Y) in the bivariate $\kappa - \mu$ type fading model the pdf of ratio of the components is also of interest, if X represents the signal power and Y the interference power.
- Exploring a gamma mixture approach (see Atapattu, et al 2011 [4]) for the univariate $\kappa - \mu$ type fading and the composite $\kappa - \mu$ type fading models, as well as exact and approximate moment generating functions (MGF) expression for these generalized fading models (see Salahat, et al 2017 [26])
- The generalised elliptical assumption (see Arashi and Nadarajah, 2016 [2]) can be assumed as the underlying model for the descriptions in this study.
- Investigate bivariate extensions such as the bivariate $\kappa - \mu$ -lognormal or the bivariate composite $\kappa - \mu$ type model, since Reig et al (2014) [25] described the usefulness of the bivariate Nakagami- lognormal in the assessment of performances in receivers with highly dynamical environments.

In conclusion, with the elliptical model as the point of departure, this study contributed to the distribution field by proposing new $\kappa - \mu$ type fading distributions that may serve as alternatives to the existing models in the literature.

Bibliography

- [1] Abramowitz, M. and Stegun, I.A., 1964. Handbook of mathematical functions: with formulas, graphs, and mathematical tables (Vol. 55). Courier Corporation.
- [2] Arashi, M. and Nadarajah, S., 2016. Generalized elliptical distributions. Communications in Statistics-Theory and Methods, (just-accepted).
- [3] Arashi, M., Bekker, A., Loots, M.T. and Roux, J.J., 2015. Integral representation of quaternion elliptical density and its applications. Communications in Statistics-Theory and Methods, 44(4), pp.778-789.
- [4] Atapattu, S., Tellambura, C. and Jiang, H., 2011. A mixture gamma distribution to model the SNR of wireless channels. IEEE transactions on wireless communications, 10(12), pp.4193-4203.
- [5] Badarneh, O.S. and Aloqlah, M.S., 2016. Performance Analysis of Digital Communication Systems Over $\alpha - \eta - \mu$ Fading Channels. IEEE Transactions on Vehicular Technology, 65(10), pp.7972-7981.
- [6] Chu, K.A.C., 1973. Estimation and decision for linear-systems with elliptical random processes. IEEE Transactions on Automatic Control, (5), pp.499-505.
- [7] Cotton, S.L, 2015. Human Body Shadowing in Cellular Device-to-Device Communications: Channel Modeling Using the Shadowed $\kappa - \mu$ Fading Model. IEEE Journal on Selected areas in Communications, 33(1), pp.111-119.
- [8] de Souza, R.A.A. and Yacoub, M.D., 2008. Bivariate Nakagami- m distribution with arbitrary correlation and fading parameters. IEEE Transactions on Wireless Communications, 7(12), pp.5227-5232.
- [9] de Souza, R.A.A., Yacoub, M.D. and Rabelo, G.S., 2012. Bivariate Hoyt (Nakagami- q) distribution. IEEE Transactions on Communications, 60(3), pp.714-723.
- [10] Ermolova, N.Y. and Tirkkonen, O., 2015. Cumulative Distribution Function of Bivariate Gamma Distribution With Arbitrary Parameters and Applications. IEEE Communications Letters, 19(2), pp.167-170.
- [11] Ferreira, J.T., Bekker, A. and Arashi, M. Advances in Wishart-type modelling of channel capacity. Submitted for publication to RevStat Statistical Journal.
- [12] García-Corrales, C., Cañete, F.J. and Paris, J.F., 2014. Capacity of $\kappa - \mu$ Shadowed Fading Channels. International Journal of Antennas and Propagation, 2014.
- [13] Gradshteyn, I.S. and Ryzhik, I.M. 2007. Table of Integral, Series, and Products, 7th Ed., Academic Press.

- [14] Lopez-Fernandez, J., Paris, J.F. and Martos-Naya, E., 2017. Bivariate Rician shadowed fading model. arXiv preprint arXiv:1701.02981.
- [15] Lopez-Martinez, F.J., Morales-Jimenez, D., Martos-Naya, E. and Paris, J.F., 2013. On the bivariate Nakagami- m cumulative distribution function: Closed-form expression and applications. *IEEE Transactions on Communications*, 61(4), pp.1404-1414.
- [16] Lopez-Martinez, F.J., Paris, J.F. and Romero-Jerez, J.M., 2016. The $\kappa - \mu$ Shadowed Fading Model with Integer Fading Parameters. arXiv preprint arXiv:1609.00317.
- [17] Marques, F.J. and Loingeville, F., 2016. Improved near-exact distributions for the product of independent Generalized Gamma random variables. *Computational Statistics & Data Analysis*, 102, pp.55-66.
- [18] Mendes, J.R. and Yacoub, M.D., 2007. A general bivariate Ricean model and its statistics. *IEEE transactions on vehicular technology*, 56(2), pp.404-415.
- [19] Miller, K. , *Multidimensional Gaussian Distributions*, SIAM Series on Applied Mathematics. New York, NY, USA:Wiley, 1964.
- [20] Nicolas, J.M. and Tupin, F., 2016, August. Statistical models for SAR amplitude data: A unified vision through Mellin transform and Meijer functions. In *Signal Processing Conference (EUSIPCO), 2016 24th European* (pp. 518-522). IEEE.
- [21] Ollila, E., Eriksson, J. and Koivunen, V., 2011. Complex elliptically symmetric random variables—generation, characterization, and circularity tests. *IEEE Transactions on Signal Processing*, 59(1), pp.58-69.
- [22] Moreno-Pozas, L., Lopez-Martinez, F.J., Paris, J.F. and Martos-Naya, E., 2016. The $\kappa - \mu$ Shadowed Fading Model: Unifying the $\kappa - \mu$ and $\eta - \mu$ Distributions. *IEEE Transactions on Vehicular Technology*, 65(12), pp.9630-9641.
- [23] Paris, J.F., 2014. Statistical Characterization of $\kappa - \mu$ Shadowed Fading. *IEEE Transactions on Vehicular Technology*, 63(2), pp.518-526.
- [24] Prudnikov, A.P., Brychkov, Y.A. and Marichev, O.I., 1988. *Integrals and Series, Volume 2: Special Functions*, Second Printing with corrections. Gordon and Breach Science Publishers.
- [25] Reig, J., Rubio, L. and Rodrigo-Peñarrocha, V.M., 2014. On the bivariate Nakagami-Lognormal distribution and its correlation properties. *International Journal of Antennas and Propagation*, 2014.
- [26] Salahat, E., Hakam, A., Ali, N. and Kulaib, A., *Moment Generating Functions of Generalized Wireless Fading Channels and Applications in Wireless Communication Theory*.
- [27] Shankar, P.M., 2005. Outage probabilities in shadowed fading channels using a compound statistical model. *IEEE Proceedings-Communications*, 152(6), pp.828-832.
- [28] Shankar, P.M., 2011. *Fading and shadowing in wireless systems*. Springer Science & Business Media.
- [29] Simon, M.K. and Alouini, M.S., 2005. *Digital communication over fading channels* (Vol. 95). John Wiley & Sons.

- [30] Sofotasios, P.C. and Freear, S., 2015. On the $\kappa - \mu$ /Gamma Generalized Multipath/Shadowing Fading Distribution. arXiv preprint arXiv:1505.04186.
- [31] Sofotasios, P.C., Tsiftsis, T.A., Van, K.H., Freear, S., Wilhelmsson, L.R. and Valkama, M., 2013, September. The $\kappa - \mu$ /Ig Composite Statistical Distribution in RF and FSO Wireless Channels. In Vehicular Technology Conference (VTC Fall), 2013 IEEE 78th (pp. 1-5).
- [32] Stein, S., 1987. Fading channel issues in system engineering. IEEE Journal on Selected Areas in Communications, 5(2), pp.68-89.
- [33] Villavicencio, M.A.G., de Souza, R.A.A., de Souza, G.C. and Yacoub, M.D., 2016. A Bivariate $\kappa - \mu$ Distribution. IEEE Transactions on Vehicular Technology, 65(7), pp.5737-5743.
- [34] Vural, M., Kurt, G.K. and Schneider, C., 2015, May. The effect of shadow fading distributions on outage probability and coverage area. In Vehicular Technology Conference (VTC Spring), 2015 IEEE 81st (pp. 1-6).
- [35] Yacoub, M.D., 2007. The $\kappa - \mu$ distribution and the $\eta - \mu$ distribution. IEEE Antennas and Propagation Magazine, 49(1), pp.68-81.
- [36] Yoo, S.K., Cotton, S.L., Sofotasios, P.C., Matthaiou, M., Valkama, M. and Karagiannidis, G.K., 2015, September. The $\kappa - \mu$ /Inverse gamma fading model. In IEEE PIMRC'15.