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# Appendix A

## List of Symbols

This appendix lists symbols used within this thesis.

### Chapter 2 – Clustering and Quality Measures

$P$	set of patterns that needs to be partitioned . . . . .	9
$N$	number of dimensions . . . . .	9
$\mathbf{p}_i$	$i$ -th pattern in $P$ . . . . .	9
$I$	size of $P$ . . . . .	9
$K$	number of clusters . . . . .	9
$C_k$	$k$ -th cluster . . . . .	9
$\sigma$	distance based similarity measure . . . . .	10
$\Gamma$	cosine similarity measure . . . . .	11
$\mathbf{Z}$	covariance matrix for the distribution of feature vectors . . . . .	11
$\mathbf{p}_i^T$	transpose of $\mathbf{p}_i$ . . . . .	11
$\mathbf{p}_{i,n}$	$n$ -th attribute of $i$ -th pattern in $P$ . . . . .	12
$\oplus$	exclusive-or between two bits . . . . .	12
$\ell$	linking distance measure . . . . .	14
$\mathbf{c}_k$	centroid of cluster $C_k$ . . . . .	16
$J_X$	function measuring different quantization errors . . . . .	15
$\mathbf{M}$	matrix of membership degrees . . . . .	17
$m_{i,k}$	function measuring the degree of membership . . . . .	17
$\phi$	radius of influence . . . . .	18
$C$	set of $K$ centroids . . . . .	18

$G$	sum of individual densities . . . . .	18
$g$	Gaussian density function . . . . .	19
$\chi_k$	mixing probability for $C_k$ . . . . .	19
$\Omega_k$	mean vector of $C_k$ . . . . .	19
$\xi_k$	distribution parameters for $C_k$ . $\xi_k = (\rho_k, \mu_k, \mathbf{Z}_k)$ . . . . .	19
$\Xi$	set of $\xi$ . . . . .	19
$Q_X$	cluster validity index of $X$ . . . . .	22
$v$	diameter of a cluster . . . . .	22
$E_i$	$i$ -th execution of an algorithm . . . . .	22
$C_k^i$	$k$ -th cluster in $E_i$ . . . . .	22
$inf$	information gain/loss . . . . .	23
$M$	magnitude of change . . . . .	23
$d$	direction of $M$ . . . . .	23
$\mathcal{E}$	ascending ordered set of feature vector pair distances . . . . .	24
$e$	$e \in \mathcal{E}$ ; feature vector pair distance . . . . .	25
$S_{min}$	sum of a number of smallest feature vector pair distances . . . . .	24
$S_{max}$	sum of a number of largest feature vector pair distances . . . . .	24
$\Psi$	variance of a set of vectors . . . . .	25
$\iota$	average standard deviation of all clusters . . . . .	26
$n$	function which determines whether a feature vector is within the neighbour- hood of another vector . . . . .	26
$inter_{min}$	minimum separation between clusters . . . . .	27
$\omega$	control parameters for non-stationary environment . . . . .	29
$f$	objective (fitness) function . . . . .	32
$best(t)$	best candidate solution in a population at time $t$ . . . . .	32
$f_{max}(t)$	best fitness value in search space at time $t$ . . . . .	32
$f_{min}(t)$	worst fitness value in search space at time $t$ . . . . .	32
$accuracy(t)$	accuracy of $best(t)$ at time $t$ . . . . .	32
$stab(t)$	stability of an algorithm at time $t$ . . . . .	32
$sens(t)$	sensitivity of an algorithm to a change at time $t$ . . . . .	33
$react_{\epsilon}(t)$	time taken to relocate $best(t)$ within a minimum time-ratio of $\epsilon$ . . . . .	33
$T$	number of time steps . . . . .	33
$currentBest(t)$	fitness of $best(t)$ . . . . .	34
$currentBestOff(t)$	maximum of $currentBest(t')$ where $1 \leq t' \leq t$ . . . . .	34
$currentAvg(t)$	the average accuracy of a population at time $t$ . . . . .	34

$windowAcc(t)$	the <i>best</i> ( $t$ ) within a certain time-span	34
$cmf$	collective mean fitness	35
$o$	set of outlier feature vectors	36
$S_o$	subset of potential outlier feature vectors	38
$sf$	smoothing factor	38
$D$	function of dissimilarity	38
$S_e$	exception set of feature vectors with highest $sf$ value	38
$S$	swarm or population of particles	39
$\mathbf{x}_i$	current feature vector or position of the $i$ -th particle in $S$	39
$\mathbf{b}_i$	the <i>personal best position</i> of the $i$ -th particle in $S$	39
$\mathbf{v}_i$	current velocity of the $i$ -th particle in $S$	39
$\mathbf{V}_{max}$	the maximum allowed velocity of any particle in $S$	39
$\mathbf{g}_i$	position of the <i>best</i> particle in the <i>neighbourhood</i> of the $i$ -th particle in $S$	39
$w$	inertia weight	40
$c_1, c_2$	acceleration constants	40
$\delta$	ratio of $\mathbf{V}_{max}$	41
$\mathbf{w}_{rc}$	codebook vector at row $r$ and column $c$	43
$\mathbf{w}'$	best matching neuron	44
$\Lambda$	neighbourhood of $\mathbf{w}'$	44
$\gamma$	learning rate	44

## Chapter 4 – Artificial Immune Systems

$\mathcal{B}$	ALC/ARB/antibody population	67
$\mathcal{A}$	antigen population	67
$\mathcal{S}$	set of selected ALCs ( $\mathcal{S} \subseteq \mathcal{B}$ )	68
$\mathcal{H}$	set of selected ALCs with highest affinity ( $\mathcal{H} \subseteq \mathcal{S}$ )	68
$\mathfrak{w}$	shape space (search space)	70
$V$	total volume of $\mathfrak{w}$	70
$\mathbf{b}$	ALC in $\mathcal{B}$	70
$\mathbf{a}$	antigen pattern in $\mathcal{A}$	70
$r$	antigen affinity threshold	70

$v$	average volume in $\varpi$ with radius $\phi$ . . . . .	70
$\mathbf{x}$	antigen detector . . . . .	74
$\Upsilon$	set of self patterns . . . . .	74
$\kappa$	alphabet size . . . . .	75
$\mathbf{x}^*$	length-limited/crossover hole . . . . .	74
$W_i^{\mathbf{x}}$	template window of pattern $\mathbf{x}$ starting at position $i$ in $\mathbf{x}$ . . . . .	74
$\mathbf{s}$	self pattern . . . . .	76
$\Theta$	multiplying factor . . . . .	82
$\mathcal{F}$	free-antibodies population . . . . .	84
$\mathcal{M}$	memory ALC population . . . . .	84
$\mathbf{b}^*$	clone of $\mathbf{b}$ . . . . .	85
$\mathbf{b}'$	mutated $\mathbf{b}^*$ . . . . .	85
$\eta$	number of clones . . . . .	85
$\sigma_{max}$	maximum distance/affinity between two patterns . . . . .	85
$\mathbf{m}_{min}$	memory ALC with lowest $\sigma$ . . . . .	85
$v$	concentration level . . . . .	87
$NAT$	network affinity threshold . . . . .	90
$\vartheta$	stimulation level . . . . .	90
$ps$	antigen stimulation level . . . . .	90
$ns$	network stimulation level . . . . .	90
$nn$	network suppression level . . . . .	90
$\alpha_{\mathbf{b}}$	set of antigen stimulation levels with ALC $\mathbf{b}$ . . . . .	90
$\lambda_{\mathbf{b}}$	set of network stimulation levels with ALC $\mathbf{b}$ . . . . .	90
$R$	number of resources allocated to an ALC . . . . .	92
$R_k$	resource constant . . . . .	92
$\vartheta'$	normalised stimulation level . . . . .	92
$R_{\gamma}$	decaying rate of resources . . . . .	94
$R_{\Lambda}$	mortality threshold . . . . .	94
$\mathbf{h}$	highest stimulated ALC . . . . .	94
$\mathcal{B}_{init}$	initial size of $\mathcal{B}$ . . . . .	95
$R_{init}$	initial value of $R$ . . . . .	95
$R_{max}$	maximum number of available resources . . . . .	96
$R_{decay}$	decayed resource level . . . . .	96
$s$	density of patterns in a set . . . . .	97

$\tau$	rate of affinity contribution	100
$\phi_{init}$	initial value of $\phi$	100
$k_{compress}$	number of sub-nets	102
$\alpha$	co-stimulation coefficient	102
$\beta$	network suppression coefficient	102
$m_{min}$	minimum degree for activation	103
$age_i$	age of $i$ -th ALC	103
$a_{min}$	minimum age of ALC	103
$a_{max}$	maximum age of ALC	103
$k_{clone}$	maximum number of clones	103
$\mathcal{B}_{max}$	maximum size of $\mathcal{B}$	103
$\zeta$	mutation rate	104
$\epsilon_{death}$	affinity death threshold	104
$\epsilon_{network}$	network suppression threshold	105
$\varphi$	number (ratio) of ALCs in $\mathcal{B}$ with the lowest affinity	105
$\epsilon_{fitness}$	ALC fitness threshold	105
$l$	number of levels in an idiotypic network of ALCs	116
$z$	number of network connections in an idiotypic network of ALCs	117
$v^j$	concentration level of an ALC in level $j$ of an idiotypic network	118
$\vartheta^i$	stimulation level of an ALC in level $i$ of an idiotypic network	118

## Chapter 5 – A Local Network Neighbourhood Artificial Immune System with Application to Unsupervised Data Clustering

$\mathbf{b}_h$	ALC with the highest antigen affinity at position $h$ in $\mathcal{B}$	127
$\mathbf{a}'$	antigen mutated clone	127
$\mathcal{C}_h$	set of antigen mutated clones of ALC $\mathbf{b}_h$	127
$ \mathcal{C}_h $	clonal level of ALC $\mathbf{b}_h$	127
$\epsilon_{clone}$	clonal level threshold	127
$\sigma^*$	normalised antigen affinity	127
$\theta$	vector difference	128



$\mathcal{N}$	ALC network neighbourhood .....	129
$\rho$	window size of $\mathcal{N}$ , i.e. neighbourhood size .....	129
$\mu$	average network affinity between ALCs .....	129
$i^*$	index position in $\mathcal{B}$ with the highest average network affinity .....	131
$\rho_r$	$\rho$ expressed as a ratio of $\mathcal{B}_{max}$ .....	151

## Chapter 7 – Data Clustering in Uncertain Environments using a Local Network Neighbourhood Artificial Immune System

$\bar{Q}_r$	mean clustering quality for run $r$ .....	208
$\hat{Q}$	collective mean clustering quality .....	208
$Q_{best}(t)$	clustering quality at time $t$ .....	208
$P^{(t)}$	data set $P$ at time $t$ .....	208
$K^{(t)}$	number of clusters in data set $P^{(t)}$ at time $t$ .....	208
$\tilde{s}$	severity of change in $P$ .....	208
$\tilde{f}$	frequency of change in $P$ .....	209

# Appendix B

## Derived Publications

This appendix provides a list of articles which were derived from the work introduced in this thesis. These articles have been published or are currently being reviewed.

### Book Chapters

A.J. Graaff and A.P. Engelbrecht. Chapter 18: Natural Immune System. *Computational Intelligence: An Introduction*, 2nd Edition, A.P. Engelbrecht (Author), John Wiley & Sons, October 2007.

A.J. Graaff and A.P. Engelbrecht. Chapter 19: Artificial Immune Models. *Computational Intelligence: An Introduction*, 2nd Edition, A.P. Engelbrecht (Author), John Wiley & Sons, October 2007.

## Journal Publications

A.J. Graaff and A.P. Engelbrecht. Optimised Coverage of Non-self with Evolved Lymphocytes in an Artificial Immune System. *International Journal of Computational Intelligence Research*, vol. 2, no. 2, pp. 127–150, 2006.

A.J. Graaff and A.P. Engelbrecht. Clustering Data in an Uncertain Environment using an Artificial Immune System. *Pattern Recognition Letters*, vol. 32, no. 2, pp. 342–351, January 2011.

A.J. Graaff and A.P. Engelbrecht. Using sequential deviation to dynamically determine the number of clusters found by a local network neighbourhood artificial immune system. *Applied Soft Computing*, vol. 11, pp. 2698–2713, March 2011.

A.J. Graaff and A.P. Engelbrecht. Clustering Data in Stationary Environments with a Local Network Neighborhood Artificial Immune System. *International Journal of Machine Learning and Cybernetics*, submitted May 2011.

## Conference Publications

A.J. Graaff and A.P. Engelbrecht. A local network neighbourhood artificial immune system for data clustering. In *IEEE Congress on Evolutionary Computation*, CEC 2007., pp. 260–267, 2007.

A.J. Graaff and A.P. Engelbrecht. Towards a self regulating local network neighbourhood artificial immune system for data clustering. In *IEEE Congress on Evolutionary Computation*, CEC 2008.(IEEE World Congress on Computational Intelligence), pp. 633–640, 2008.