

A Stochastic Space-Time Rainfall Model for Engineering Risk Assessment

by Michael Leonard

Submitted in fulfilment of the requirements for the degree of
DOCTOR OF PHILOSOPHY

April 14, 2010

FACULTY OF ENGINEERING, COMPUTER AND MATHEMATICAL SCIENCES

School of Civil Environmental and Mining Engineering



A Stochastic Space-Time Rainfall Model for Engineering Risk Assessment

By: Michael Leonard, *B.E. Civil (Hons)*

April 14, 2010

Thesis submitted in fulfilment of the requirements for the degree of
Doctor of Philosophy

School of Civil Environmental and Mining Engineering
Faculty of Engineering, Computer and Mathematical Sciences
The University of Adelaide SA 5005 Australia

Telephone: +61 8 8303 5451
Facsimile: +61 8 8303 4359
Web: www.ecms.adelaide.edu.au/civeng
Email: mleonard@civeng.adelaide.edu.au

Table of Contents

Table of Contents	iii
List of Figures	vii
List of Tables	xi
Abstract	xiii
Statement of Originality	xv
Acknowledgements	xvii
List of Symbols	xix
1 Introduction	1
1.1 The Rational Method	2
1.2 Design Events	2
1.2.1 AEP Neutrality	3
1.2.2 Intensity Frequency Duration curves	4
1.2.3 Temporal patterns	4
1.3 Monte Carlo Simulation	6
1.4 Continuous Simulation	7
1.5 The Australian Climate	8
1.6 Rainfall Measurement	9
1.6.1 Daily rainfall gauges	11
1.6.2 Pluviograph rainfall gauges	13
1.6.3 Radar rainfall reflectivity	14
1.6.4 Summary of rainfall data	15
1.7 Thesis Objectives	16
1.8 Thesis Structure	18

1.9	Overview of Australian Case studies	19
2	Literature Review	23
2.1	Rainfall Homogeneity	24
2.1.1	Spatial homogeneity	24
2.1.2	Temporal homogeneity	25
2.2	Single-site Rainfall Models	26
2.2.1	Event-based models	27
2.2.2	Poisson-cluster models	28
2.2.3	Disaggregation models	32
2.2.4	Regionalised single-site models	33
2.3	Multi-site Models	34
2.4	Space-time Models	35
2.4.1	Radar-based models	35
2.4.2	Gauge-based models	38
2.5	Summary of Literature	41
3	Model Formulation and Calibration	43
3.1	Model Formulation	43
3.2	Calibration Statistics	48
3.3	Calibration Summary	54
4	Simulation Involving a Regional Boundary	57
4.1	Introduction	57
4.2	Algorithm for Simulation of Cells Using a Buffer Region	58
4.3	Algorithm for Direct Simulation of Cells	61
4.3.1	Number of Outer Cells Intersecting Target	61
4.3.2	Cell Centre Conditioned on Intersecting Target	63
4.3.3	Cell Radius Conditioned on Location	63
4.4	Results and Discussion	64
4.5	Conclusion	66
5	Calibration Involving a Monthly Boundary	67
5.1	Introduction	67
5.2	Derivation of Monthly Bias	68
5.3	Case-study	71
5.4	Conclusion	73

6 Calibration Using Simulated Moments	75
6.1 Introduction	75
6.2 Calibration methodology	76
6.3 Case-study	78
6.4 Conclusion	84
7 Spatial Storm Extent	87
7.1 Introduction	87
7.2 Probability of storm occurrence at a point	87
7.3 Model having storm extent	89
7.4 Model calibration	92
7.5 Model simulation	93
7.6 Case-study application	94
7.6.1 Modelled region and calibrated parameters	94
7.6.2 Interpolation of Scale Parameters	96
7.6.3 Model Comparison to Calibration Statistics	100
7.6.4 Model Comparison to Other Statistics	101
7.7 Visualisation of Spatial Storm Process	103
7.8 Conclusion	107
8 Climatic and Seasonally Partitioned Extreme Rainfall	109
8.1 Introduction	110
8.2 Methodology	114
8.3 Urban Design Application - Scott Creek	117
8.4 Climate Conditioned SNSRP Methodology	123
8.5 Bourke Case Study	126
8.6 Observed Partitioned Extremes	127
8.7 Comparison of Observed and Simulated Extremes	134
8.8 Conclusions	135
9 Spatially Inhomogeneous Neyman-Scott Model	137
9.1 Introduction	137
9.2 Model Development	139
9.2.1 Spatial process	139
9.3 Integral Approximation	139
9.3.1 Temporal process	146

Table of Contents

9.3.2	Matrix Inversion to Find Spatial Rate Surface	149
9.4	Simulation Technique	154
9.5	Case Study	157
9.5.1	Observed Data	157
9.5.2	Calibration to Regional Statistics	163
9.5.3	Estimation of number of cells	164
9.6	Results	167
9.7	Conclusions	171
10	Summary and Recommendations	177
10.1	Model Review	177
10.2	Recommendations	181
References		185
Appendices		197
Appendix A	Spatial Storm Extent	199
A.1	List of rainfall gauges	199
A.2	Simulated Annual Totals	202
A.3	Simulated Extreme Values	212
Appendix B	Bourke Case Study	227
B.1	Observed SOI Partitioned Annual Extremes	227
B.2	Observed SOI Partitioned Summer Extremes	231
B.3	Observed SOI Partitioned Winter Extremes	235
B.4	Observed Seasonally Partitioned SOI+ Extremes	239
B.5	Observed Seasonally Partitioned SOI- Extremes	243
B.6	Comparison of Observed and Simulated Extremes	247

List of Figures

1.1	24 hour design storm temporal patterns for 8 different zones of Australia.	5
1.2	Mean rainfall in millimetres depth across Australia.	10
1.3	Climatic Zones of Australia.	11
1.4	Distribution of rainfall measuring devices about Australia.	12
1.5	Number of commissioned gauges/radars in time.	16
2.1	Schematic diagram of an alternating renewal model of storm events. . . .	27
2.2	Schematic diagrams for various point-process formulations.	30
3.1	Schematic diagram of the Neyman-Scott Rectangular Pulse model. . . .	44
3.2	Schematic diagram of spatial cell process.	46
4.1	Schematic diagram of cells generated inside and outside the target region.	58
4.2	Proportion reduction in simulated rainfall depth.	60
4.3	Efficiency of direct-algorithm, proportion of memory usage and run-time.	65
5.1	Schematic of rainfall landing outside the month boundary.	69
5.2	Bias in mean due to monthly boundary.	72
5.3	Change in parameters accounting for monthly bias.	74
6.1	Spatial coordinates of observation gauges about Launceston, Tasmania. .	78
6.2	Observed and fitted temporal statistics, Launceston case-study.	80
6.3	Observed and fitted spatial statistics, Launceston case-study.	81
6.4	Observed and simulated annual maxima, Launceston.	83
6.5	Observed and simulated annual totals, Launceston.	84
7.1	Schematic diagram showing the cause of spurious cross-correlations. . . .	88
7.2	Example of variation in cross correlation at the 24 hour aggregate.	91
7.3	Sydney 24 hr spurious cross-correlations.	93
7.4	Sydney case study, gauge locations.	95
7.5	Variation in mean rainfall across Sydney.	98

7.6	Simulated values for statistics used in calibration.	99
7.7	Comparison of model to observed statistics, Sydney.	102
7.8	Model visualisation without storm-envelope.	104
7.9	Model visualisation including storm-envelope.	105
8.1	Comparison of distributions of summer and winter seasonal maxima about Australia.	113
8.2	Average monthly rainfall, Adelaide, South Australia.	118
8.3	Maximum water depths in Scott Creek Basin.	119
8.4	Maximum rainfall intensities in neighbouring catchment to Scott Creek. .	120
8.5	Annual maximum water depths in Scott Creek Basin, composite distribution from seasonal maxima.	121
8.6	Maximum water depths in Scott Creek Basin, adjusted composite distribution.	122
8.7	Southern Oscillation Index.	124
8.8	ACF and PACF of the SOI	124
8.9	Pattern of two indicator states based on partition of SOI.	125
8.10	Locations of gauges used for Bourke case-study.	127
8.11	Fitted correlation statistics for Bourke.	129
8.12	Fitted coefficient of variation for Bourke.	130
8.13	Fitted skewness for Bourke.	131
8.14	SOI+/SOI- partitioned, observed annual extremes, site 4, Bourke.	133
8.15	SOI+/SOI- partitioned, observed extremes, site 4, Bourke.	133
8.16	Seasonally partitioned, observed SOI+ extremes, site 4, Bourke.	134
8.17	Comparison of observed and simulated extremes, site 4, Bourke.	135
9.1	Avon Basin from <i>Jothityangkoon et al.</i> [2000].	138
9.2	Schematic of rainfall process.	140
9.3	Schematic of two pixels from gridded rainfall data.	140
9.4	Schematic of distance between two points (u,v) and (x,y)	141
9.5	Integration regions used for h^- case with $i = k; j = l$	142
9.6	Integration regions used for h^- case with $ i - k = j - l \geq 0$	143
9.7	Effectiveness of hypotenuse approximations for various distances.	146
9.8	1-D line example of spatial cell process.	150
9.9	1-D circular line example of spatial cell process.	151
9.10	Inhomogeneous model simulation schematic	155

9.11	Inhomogeneous model accept-reject schematic	156
9.12	Location of gauges for the Avon basin case-study.	157
9.13	Variation in mean rainfall over the Avon Basin.	159
9.14	Variation in the coefficient of variability over the Avon Basin.	160
9.15	Variation in the probability of a dry day over the Avon Basin.	161
9.16	Variation in the daily auto-correlation over the Avon Basin.	162
9.17	Fitted regional coefficient of variation, Avon Basin.	163
9.18	Fitted regional autocorrelation, Avon Basin.	163
9.19	Fitted regional skewness, Avon Basin.	164
9.20	Fitted variation in the number of cells over the Avon Basin.	166
9.21	Spline fitted number of cells over the Avon Basin.	168
9.22	Residuals of the number of cells from spline fitting.	169
9.23	Comparison of simulated and observed daily dry probability.	170
9.24	Comparison of simulated and observed daily coefficient of variation.	172
9.25	Simulated spatial rate of cells over the Avon Basin.	173
9.26	Uncertainty in cross-correlation estimates.	174
9.27	Simulated and observed cross-correl. comparison, after spline interpolation.	175
B.1	SOI+/SOI- partitioned, observed annual extremes, sites 1 - 6, Bourke. . .	228
B.2	SOI+/SOI- partitioned, observed annual extremes, sites 7 - 12, Bourke. . .	229
B.3	SOI+/SOI- partitioned, observed annual extremes, sites 13 - 16, Bourke. .	230
B.4	SOI+/SOI- partitioned, observed summer extremes, sites 1 - 6, Bourke. . .	232
B.5	SOI+/SOI- partitioned, observed summer extremes, sites 7 - 12, Bourke. .	233
B.6	SOI+/SOI- partitioned, observed summer extremes, sites 13 - 16, Bourke. .	234
B.7	SOI+/SOI- partitioned, observed winter extremes, sites 1 - 6, Bourke. . .	236
B.8	SOI+/SOI- partitioned, observed winter extremes, sites 7 - 12, Bourke. . .	237
B.9	SOI+/SOI- partitioned, observed winter extremes, sites 13 - 16, Bourke. .	238
B.10	Seasonally partitioned, observed SOI+ extremes, sites 1 - 6, Bourke. . . .	240
B.11	Seasonally partitioned, observed SOI+ extremes, sites 7 - 12, Bourke. . . .	241
B.12	Seasonally partitioned, observed SOI+ extremes, sites 13 - 16, Bourke. . .	242
B.13	Seasonally partitioned, observed SOI- extremes, sites 1 - 6, Bourke. . . .	244
B.14	Seasonally partitioned, observed SOI- extremes, sites 7 - 12, Bourke. . . .	245
B.15	Seasonally partitioned, observed SOI- extremes, sites 13 - 16, Bourke. . . .	246
B.16	Comparison of observed and simulated extremes, sites 1 - 3, Bourke. . . .	248
B.17	Comparison of observed and simulated extremes, sites 4 - 6, Bourke. . . .	249
B.18	Comparison of observed and simulated extremes, sites 7 - 9, Bourke. . . .	250

List of Figures

B.19 Comparison of observed and simulated extremes, sites 10 - 12, Bourke. . .	251
B.20 Comparison of observed and simulated extremes, sites 13 - 15, Bourke. . .	252
B.21 Comparison of observed and simulated extremes, site 16, Bourke.	253

List of Tables

1.1	Summary statistics for varied case studies.	20
5.1	Bias in mean due to monthly boundary.	71
5.2	Bias in mean due to monthly boundary.	73
6.1	Launceston parameter estimates for each month, via analytic expressions.	82
6.2	Launceston parameter estimates for each month, via MCS.	82
7.1	Algorithm for efficiently simulating storm and cell properties.	94
7.2	Parameter estimates for each month	97
8.1	List of gauges used for Bourke case-study.	128
8.2	Bourke regional parameter estimates, entire record.	128
8.3	Bourke regional parameter estimates, SOI-.	132
8.4	Bourke regional parameter estimates, SOI+.	132
9.1	Computational requirements of matrix inversion.	149
9.2	Conjugate Gradient Algorithm.	153
9.3	Avon Basin regional parameter estimates for each month.	165
A.1	List of pluviograph gauges from Sydney Water Observation Network . . .	199
A.2	List of Bureau of Meteorology daily rainfall gauges	200

List of Tables

Abstract

The temporal and spatial variability of Australia's climate affects the quantity and quality of its water resources, the productivity of its agricultural systems, and the health of its ecosystems. This variability should be taken into account when assessing the risks associated with flooding. Continuous simulation rainfall models are one means for doing this, whereby sequences of storms are generated for an arbitrarily long time period and over some region of interest. The simulated rainfall should reproduce observed statistics in time and space so that it can be used as a suitable input for hydrologic models at the catchment scale, with particular emphasis on extreme events.

There are a variety of approaches to modelling rainfall, including a broad range of single-site and multi-site rainfall models. By way of contrast there are few models that aim to simulate rainfall across all points within a region at daily or sub-daily increments. This thesis focuses on models calibrated solely to rain gauges, and a specific type known as Neyman-Scott Rectangular Pulse (NSRP) models. Existing NSRP models have a mature history of modelling developments including calibration methodology and an ability to reproduce key statistics across a range of timescales. Nonetheless, these models also have several limitations (and other space-time models notwithstanding) that are addressed in this thesis. These developments include improvements to the conceptual representation of rainfall and improvements to calibration and simulation techniques. Specifically these improvements include (i) the development of an efficient simulation technique, (ii) assessing the impact of monthly parameter changes on rainfall statistics, (iii) the use of simulated statistics within calibration to overcome reliance on derived model properties (iv) incorporating a storm extent parameter to better match spatial correlations, (v) incorporating long term climatic variability and developing a methodology to assess climatic and seasonal variability in simulated extremes (vi) incorporating inhomogeneity of rainfall occurrence across a region. Numerous case studies are used at various locations about Australia to illustrate these improvements and highlight the applicability of the model under varied climatic conditions.

Abstract

Statement of Originality

This work contains no material which has been accepted for the award of any other degree or diploma in any university or other tertiary institution to Michael Leonard and, to the best of my knowledge and belief, contains no material previously published or written by another person, except where due reference has been made in the text. I give consent to this copy of my thesis, when deposited in the University Library, being made available for loan and photocopying, subject to the provisions of the Copyright Act 1968. I also give permission for the digital version of my thesis to be made available on the web, via the University's digital research repository, the Library catalogue, the Australasian Digital Theses Program (ADTP) and also through web search engines, unless permission has been granted by the University to restrict access for a period of time.

SIGNED: DATE:

Acknowledgements

Undertaking a thesis is a privileged opportunity. Amongst the many privileges, I consider the community of support I have received and the resulting friendships to be among the greatest.

In this spirit, I would like to thank my supervisors Prof. Martin Lambert and Assoc. Prof. Andrew Metcalfe and collaborators Prof. George Kuczera, Dr. Mark Thyer and Dr. Paul Cowpertwait for their friendship, time, advice and enthusiasm. For their numerous conversations, assistance and general interest I also thank Prof. Angus Simpson, Dr. Alan Seed, Dr. Theresa Heneker, Dr. Julian Whiting, Dr. Andrew Frost, Ms. Juan Qin, Mr. Alex Osti, Mr. Steven Need, Ms. Geraldine Wong. There are many other friends I would like to mention here who encouraged me but, for the sake of avoiding long lists, to whom I hope a simple ‘thank you’ suffices. I will single out only Mr. Rob May as he has been especially generous in his friendship.

I am thankful to all of the staff in the School of Civil, Environmental and Mining Engineering at the University of Adelaide for creating a genuinely positive and constructive research environment. I am also indebted to fellow postgraduate students for their positive and constructive research distractions.

I am grateful to my family for their ongoing support and to J.B. for her love and patient understanding. Lastly, I am thankful to a merciful God who ‘sends rain on the righteous and the unrighteous’ alike (Sermon on the Mount, Matthew 5:45).

Acknowledgements

List of Symbols

Symbol	Quantity	Units
Functions		
$A(h, l), B(h, l)$	Used to calculate SNSRP covariance	
$E[]$	Expected value	
$E_1()$	First-order, exponential integral function	
f_{R_c}	Probability density for random variable R_c	
$f_{R_{xy}}$	Probability density for random variable R_{xy}	
$f(), g()$	Used to calculate SNSRP skewness	
$\Gamma()$	Gamma function	
$p_h(t)$	Probability of no-rain in interval $(t, t+h)$	
$P(N_o = n_o)$	Probability of obtaining n_o cells	
$\psi()$	Prob. of a cell overlapping a point, given it overlaps a region	
$P(\phi, d)$	Prob. of a cell overlapping a point at a distance d away	
$Pr()$	Probability that ...	
English		
A	Rational method, catchment area	km^2
A_K	Area of target region	km^2
b_r	r^{th} order, probability weighted moment	
C	Random variable, number of cells	
C_Y	Rational method, dimensionless runoff coefficient	
d	Distance between two generic points	km
dN	Number of cells ‘alive’ at time t	
$D_{i,j}$	Domain of points defining a pixel	
$D_{k,l}$	Domain of points defining a pixel	
$D \equiv I_{h,i,j,k,l}$	Indicator function for data element $x_{h,i,j,k,l}$, (1=dry,0=wet)	
F_1, F_2	Least squares objective functions used in calibration	
h	Level of aggregation	hr

continued on next page

Symbol	Quantity	Units
h, h^-, h^+	Approximations for length of a hypotenuse	
$I \equiv I_{h,i,j,k,l}$	Indicator function for data element $x_{h,i,j,k,l}$, (1=valid,0=corrupt)	
$I_{t_c,Y}$	Rational method, intensity at critical period	mm/hr
k	Month index, $k = 1 \dots 12$	
K	Set of points defining a region	
l	Index of individual data interval, $l = 1 \dots 31(24/h)$	
$L, L_{i,j}$	Random variable, cell lifetime, i^{th} storm, j^{th} cell,	hr
L	Length scale of a pixel	
M	Number of observed rainfall sites in region, $i = 1 \dots M$	
n	Number of storm types in mixture of SNSRP processes	
$n_{h,k}$	Number of data elements all sites, aggregate h , month k	
$n_{h,i,k}$	Number of data elements, aggregate h , site i , month k	
N_i	Number of years in observed record at site i , $j = 1 \dots N_i$	
N_o	Number of successful Bernoulli trials	
N_{tot}	Total number of Bernoulli trials	
N_{xy}	Number of concurrent years in records at sites x , and y	
p_o	Arbitrary constant probability for Bernoulli trials	
Q_Y	Rational method, peak flow for ARI Y	$m^3 s^{-1}$
r_K	Radius of target region	km
R_C	Random Variable, radius of cell centre	km
R_{xy}	Random Variable, dist. from region edge to cell centre	km
$S, S_{i,j}$	Random variable, cell start-time, i^{th} storm, j^{th} cell,	hr
t_3	The L-skewness, calculated from L-moments	
t_c	Rational method, critical period	s
T, T_i	Random variable, arrival time of i^{th} storm	hr
u, v	Dummy integration variables for a pixel	
U	Random variable, uniform distribution, $U \sim [0, 1]$	
w	Mixture ratio	
w_ψ, \dots	Relative weight values for the method of moments	
x, y	Dummy integration variables for a pixel	
$x_{h,i,j,k,l}$	Individual element of data, agg. h , site i , year j , month k , interval l	
$X, X_{i,j}$	Random variable, cell intensity, i^{th} storm, j^{th} cell,	mm
X_j	j^{th} order statistic, arbitrary data X	
$Y(t)$	Instantaneous rainfall process at time t	mm
$Y_l(h)$	Aggregate rainfall process, l^{th} interval, agg. level h	mm
$Y_Z(t)$	Instantaneous rainfall process at time t , spatial location Z	mm
Z	Generic location in \mathbb{R}^2 domain	

continued on next page

Symbol	Quantity	Units
Z_C	Random Variable, location of cell centre in \mathbb{R}^2 domain	

Greek

α	Storm intensity parameter	
$\alpha^{(p)}$	Storm intensity parameter, $SNSRP^{(p)}$ mixture	
β	Cell dispersion parameter	hr^{-1}
$\beta^{(p)}$	Cell dispersion parameter, $SNSRP^{(p)}$ mixture	hr^{-1}
$\gamma_{d,h,\tau}$	Covariance, distance d (km), aggregation h , time-lag τ	mm^2
$\gamma_{d,h,\tau}^{(p)}$	Covariance of $SNSRP^{(p)}$ mixture	mm^2
$\gamma_{0,h,\tau}$	Auto-covariance, from $\gamma_{d,h,\tau}$	mm^2
$\gamma_{0,h,0}$	Variance, from $\gamma_{d,h,\tau}$	mm^2
δr	Elemental radius length	km
$\delta \xi$	Elemental angle	$rad.$
ϵ_i, ϵ	Rainfall falling beyond month i boundary	mm
ζ_h	Unstandardised skewness, aggregation h	mm^3
$\zeta_h^{(p)}$	Unstandardised skew of $SNSRP^{(p)}$ mixture, agg. h	mm^3
η	Cell lifetime parameter	hr^{-1}
$\eta^{(p)}$	Cell lifetime parameter, $SNSRP^{(p)}$ mixture	hr^{-1}
θ	Storm intensity parameter / scale parameter	mm
$\hat{\theta}_{h,i,k}$	Estimate of scale parameter, agg. h , site i , month k	mm
κ_h	Standardised skewness	
$\hat{\kappa}_{h,k}$	Non-dim., pooled skew estimate, agg. h , month k	
λ	Storm rate parameter	hr^{-1}
$\lambda^{(p)}$	Storm rate parameter, $SNSRP^{(p)}$ mixture	hr^{-1}
μ_C	Mean number of cells covering a point in space	
μ_C	Other interpretation: Mean number of cells per storm	
$\mu_C^{(p)}$	Mean number of cells per storm, $SNSRP^{(p)}$ mixture	
μ_h	Mean rainfall intensity at agg. level h	mm
$\mu_h^{(p)}$	Mean rainfall intensity of $SNSRP^{(p)}$ mixture, agg. level h	mm
$\hat{\mu}_{h,i,k}$	Estimate of mean, agg. h , site i , month k	mm
$\hat{\mu}_{h,k}$	Pooled estimate of mean, agg. h , month k	mm
μ_y	Mean rainfall of instantaneous rainfall intensity	mm
ν_h	Coefficient of variation	

continued on next page

Symbol	Quantity	Units
$\hat{\nu}_{h,k}$	Pooled estimate of coefficient of variation, agg. h , month k	
ξ	Arbitrary angle at which a cell is located	
Ξ	Random variable, uniform angle, $X_i \sim [0, 2\pi]$	
ρ_h	Lag-1 auto-correlation	
$\rho_{x,y,h}$	Lag-0 cross-correlation between site x and site y , agg. h	
$\hat{\rho}_{x,y,h,k}$	Estimate of lag-0 cross-correlation, sites x & y , agg. h , month k	
σ_h^2	Variance, aggregation h	mm^2
$\hat{\sigma}_{h,k}^2$	Non-dim., pooled variance estimate, agg. h , month k	
τ	Correlation lag in time	
φ_C	Spatial rate parameter	km^{-2}
ϕ	Generic radius parameter, either ϕ_c or ϕ_s	km^{-1}
ϕ_c	Cell radius parameter	km^{-1}
$\phi_c^{(p)}$	Cell radius parameter, $SNSRP^{(p)}$ mixture	km^{-1}
ϕ_s	Storm radius parameter	km^{-1}
$\phi_s^{(p)}$	Storm radius parameter, $SNSRP^{(p)}$ mixture	km^{-1}
ψ_h	Dry portion / probability of a dry interval	
$\psi_h^{(p)}$	Dry portion of $SNSRP^{(p)}$ mixture	
Ψ_i, Ψ	Rainfall bias in mean / portion of month i rainfall	
$\hat{\psi}_{h,k}$	Pooled estimate of dry portion, agg. h , month k	
ω	ratio of radius parameter ϕ to radius of target region r_t	
Ω	Generic parameter of Poisson distribution	

Acronyms

<i>ABARE</i>	Australian Bureau of Agriculture and Resource Economics
<i>ABS</i>	Australian Bureau of Statistics
<i>AMF</i>	Annual Maximum Frequency
<i>ARI</i>	Annual Recurrence Interval
<i>ARMA</i>	Auto-Regressive Moving-Average
<i>BLRP</i>	Bartlett-Lewis Rectangular Pulse
<i>BOM</i>	Bureau of Meteorology
<i>BTE</i>	Bureau of Transport Economics
<i>CV</i>	Coefficient of Variation
<i>ENSO</i>	El Niño Southern Oscillation
<i>FFT</i>	Fast Fourier Transform
<i>GDP</i>	Gross Domestic Product
<i>GRF</i>	Gaussian Random Field

continued on next page

Symbol	Quantity	Units
<i>IPO</i>	Interdecadal Pacific Oscillation	
<i>MTB</i>	Modified Turning Bands	
<i>NSRP</i>	Neyman-Scott Rectangular Pulse	
<i>MCS</i>	Monte Carlo Simulation	
<i>SNSRP</i>	Spatial Neyman-Scott Rectangular Pulse	
<i>SNSRP</i> ^(<i>p</i>)	The <i>p</i> th superimposed SNSRP process	
<i>SOI</i>	Southern Oscillation Index	

List of Symbols
