

# Statistical Methods for Clusters of Extreme Values

Christopher A. T. Ferro, B.Sc. M.Sc.

Submitted for the degree of Doctor of Philosophy

at Lancaster University, September 2003.

I declare that the work presented in this thesis is my own,  
except where stated otherwise.

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## Abstract

Extreme values in sequences of independent random variables tend to occur in isolation; for sequences of serially dependent variables, extremes can occur in clusters. Extreme-value theory is well developed for stationary processes and provides mathematical characterisations of the clustering of extremes. Such characterisations are useful models for the extremal behaviour of physical processes: consider storms, floods and droughts for example. Statistical applications harness extreme-value theory to make inferences about the extremes of a process based on a finite sample of data. This thesis addresses several topics in the analysis of clusters of extreme values from both univariate and multivariate processes. Key developments are the following: a theoretically justified scheme for identifying clusters in a sample; estimators for the extremal index that do not require clusters to be identified; a semi-parametric estimator for multivariate extreme-value densities; estimators for cluster summaries that exploit the asymptotic structure of clusters; and a method for modelling clusters in multivariate processes.

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