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# Network Event Detection with Entropy Measures

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# Abstract

Information measures may be used to estimate the amount of information emitted by discrete information sources. Network streams are an example for such discrete information sources. This thesis investigates the use of information measures for the detection of events in network streams.

Starting with the fundamental entropy and complexity measures proposed by Shannon and Kolmogorov, it reviews a range of candidate information measures for network event detection, including algorithms from the Lempel-Ziv family and a relative newcomer, the T-entropy. Using network trace data from the University of Auckland, the thesis demonstrates experimentally that these measures are in principle suitable for the detection of a wide range of network events.

Several key parameters influence the detectability of network events with information measures. These include the amount of data considered in each traffic sample and the choice of observables. Among others, a study of the entropy behaviour of individual observables in event and non-event scenarios investigates the optimisation of these parameters.

The thesis also examines the impact of some of the detected events on different information measures. This motivates a discussion on the sensitivity of various measures.

A set of experiments demonstrating multi-dimensional network event classification with multiple observables and multiple information measures concludes the thesis.

## Preface and Acknowledgments

This Ph.D. thesis applies T-entropy, a relatively new information measure, for network event detection for the first time. If the author is correct, it is also the first Ph.D. thesis to review the expected entropies of common network protocol fields.

Now looking back this project, I remember many occasions on which I could have gotten sidetracked. I would like to express my gratitude to colleagues and friends for their support during the sometimes quite challenging periods of this project.

My academic supervisor Ulrich Speidel made himself available at short notice for discussions and was always ready with good suggestions when I ran into difficulties. On countless occasions Ulrich helped me to ship around difficulties by providing new insights and ideas. Ulrich's support was not limited to the project work itself: He also helped me come to terms with turbulent times in my personal life, the loss of my father being the most difficult.

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With English being my second language, this thesis depended heavily on the watchful eyes of my proof-readers, Caroline and Peter Seddon, as well as Michael Taylor to whom I feel deeply indebted.

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