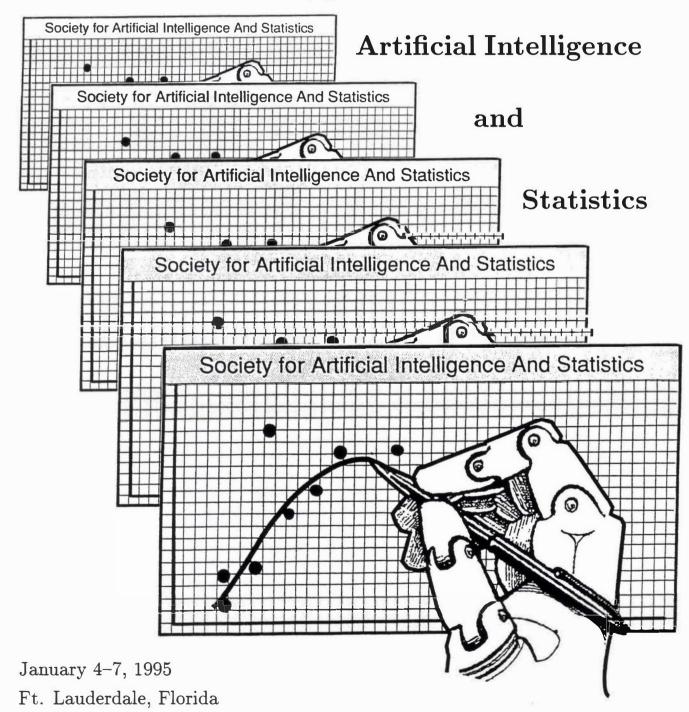
Preliminary Papers

of the

Fifth International Workshop

on



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PREFACE

Almost 10 years ago, in April 1985, Bill Gale and Daryl Pregibon organized the first Workshop on Artificial Intelligence and Statistics at Princeton, New Jersey. In his Preface to the edited volume Bill motivated the Workshop by noting that research in each field could benefit the other. The 15 papers presented at the first Workshop reflected a variety of interdisciplinary themes: task analysis of and expert advisory systems for statistical strategy, reasoning under uncertainty, supervised and unsupervised learning, heuristic search in pattern recognition, statistical tools for knowledge acquisition, and characterizing the behavior of 'intelligent' systems through statistical analysis. Papers in subsequent workshops have further developed some of these themes and/or branched into other areas at the interface of AI and statistics.

It is difficult to know the influence of the AI and Statistics Workshops on research in the two communities. Suffice it to say that I have used these Workshops to guide, gauge, and 'correct' my own work. I have heard several volunteer it as their favorite forum for discussion and debate, and in any case, it seems generally acknowledged as a workshop of very high quality. Finally, in contrast to 15 papers presented at the first Workshop in 1985, there are 62 papers, selected from a total of 111 submissions, being presented at the Fifth International Workshop on AI and Statistics in 1995. Apparently, many concur with Bill's motivation for the original Workshop.

There are many people that helped bring this Workshop to fruition. Hans Lenz, the Program Chair, was responsible for selecting the program committee, assigning papers, collecting reviews, and deciding the technical program. He also iterated with me in scheduling the presentations. Despite occasional logistical problems in the Nashville-Berlin link, it was a pleasure working with Hans. Hans and I thank those that provided submissions for review - unfortunately, not all quality submissions could be accommodated. We thank the program committee for reviews and advise on paper selection. Daryl Pregibon handled finances, local arrangements, and provided input on scheduling - if there was a General CoChair position, then the title would belong to Daryl. The Tutorial Chair, Prakash Shenoy, put together a wonderful tutorial program. Our tutorial presenters, David Aha, Trevor Hastie, Steffen Lauritzen, and Glenn Shafer deserve special thanks - the time investment necessary for a good tutorial is very high, and we appreciate their efforts. My wife, Pat Fisher, and Jing Lin, Julio Ortega, and Doug Talbert of Vanderbilt University helped put together the proceedings. Robyn Landers of the University of Waterloo has been of great service by maintaining the AI and Statistics electronic mailing list, over which considerable Workshop information has been promulgated.

Finally, I greatly appreciate the sage advice of past organizers, particularly their input on program committee selection, but on a variety of other issues as well. They include Peter Cheeseman, Bill DuMouchel, David Hand, Wayne Oldford, and Daryl Pregibon. I am optimistic that this year's meeting will follow in the tradition of high quality that they helped hone.

Doug Fisher, General Chair

ORGANIZATION

General Chair:

D. Fisher

Vanderbilt U., USA

Program Chair:

H. Lenz

Free U., Berlin, Germany

Program Committee Members:

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E. Roedel Humboldt U., Germany

G. Shafer Rutgers U., USA

P. Smyth JPL, USA

Tutorial Chair:

Local Arrangements:

P. Shenoy

U. Kansas, USA

D. Pregibon AT&T Bell Labs, USA

Sponsors:

Society for Artificial Intelligence and Statistics International Association for Statistical Computing

TUTORIALS

Wednesday, January 4

Machine Learning	9:00	AM to	12:15	PM
Statistical Methods for Inducing Models from Data Prof. Steffen Lauritzen, Aalborg University	9:00	AM to	12:15	PM
Probabilistic Models of Causality	2:00) PM t	o 5:15	PM
Statistical Models for Function Estimation and Classification Prof. Trevor Hastie, Stanford University	2:00) PM t	o 5:15	PM

SCHEDULE

Thursday, January 5

8:45	Opening Comments
9:00	A causal calculus for statistical research, with applications to observational
	and experimental studies - J. Pearl
9:30	Causal discovery from data in the presence of selection bias -G. F. Cooper 140
≥ 10:00	Likelihood-based causal inference – Q. Yao, D. Tritchler
10:30	Break
11:00	Missing data models as meta-data - R. G. Almond, J. Schimert
11:30	Learning in hybrid noise environments using statistical queries - S. E. Decatur 175
12:00	Lunch
1:30	Heuristic search for model structure – J. F. Elder
2:00	Tree structured interpretable regression – D. Lubinsky
2:30	Break
3:00	Poster Summaries
5:00	Break
6:00	Dinner
6:30	Poster Session 1 (ends at 8:10)
8:20	Poster Session 2 (ends at 10:00)

Friday, January 6

8:30	Learning Bayesian networks: Search methods and experimental results
	- D. M. Chickering, D. Geiger, D. Heckerman
9:00	Derivation DAGs for inferring interaction models - F. M. Malvestuto 354
9:30	Learning possibilistic networks from data – J. Gebhardt, R. Kruse
10:00	Break
10:30	Software for data analysis with graphical models - W. L. Buntine, H. S. Roy 76
11:00	Preliminary system design for an EDA assistant - R. St Amant, P. R. Cohen . 502
11:30	Framework for a generic knowledge discovery toolkit - P. Riddle, R. Fresnedo,
	D. Newman
12:00	Lunch
1:30	A generalization of the Tetrad representation theorem - G. Shafer, A. Kogan,
	P. Spirtes
2:00	Two algorithms for inducing structural equation models from data
	- P. R. Cohen, D. E. Gregory, L. Ballesteros, R. St Amant
2:30	Break
3:00	Panel/Debate (topic TBA)
8:00	Informal Discussions (on topics TBA)

Saturday, January 7

9:00	On test selection strategies for belief networks - D. Madigan, R. G. Almond 342
9:30	Using causal knowledge to learn more useful decision rules from data
	- L. A. Cox, Jr
10:00	Dynamic learning bias selection - C. J. Merz
10:30	Break
11:00	Viewpoint-based measurement of semantic similarity between words
	- K. Kasahara, K. Matsuzawa, T. Ishikawa, T. Kawaoka
11:30	Discovering morphemic suffixes: A case study in MDL induction
	- M. R. Brent, S. K. Murthy, A. Lundberg
12:00	Closing Comments
12:15	Break
12:45	Business Meeting

Poster Session 1 Thursday, January 5

A comparative evaluation of sequential feature selection algorithms - D. W. Aha, R. L. Bankert

Hypergraph grammars for knowledge-based model construction - R. G. Almond

Tools for empirically analyzing AI programs - S. D. Anderson, D. M. Hart, D. L. Westbrook, P. R. Cohen, A. Carlson

Decision-theoretic case-based reasoning - J. S. Breese, D. Heckerman

Tailoring rulesets to misclassification costs - J. Catlett

Analysis and Application of the Generalized Mean-Shift Process - Y. Cheng

A further study of pruning methods in decision tree induction - F. Esposito, D. Malerba, G. Semeraro

* Classifying new words for robust parsing - A. Franz

Omega-Stat: An environment for implementing intelligent modeling strategies - E. J. Harner,

H. C. Galfalvy

A decision-based view of causality - D. Heckerman, R. Shachter

Solving influence diagrams using Gibbs sampling - A. Jenzarli

Hierarchical clustering of composite objects with a variable number of components – A. Ketterlin, P. Gançarski, J. J. Korczak

Structure learning of Bayesian networks by hybrid genetic algorithms - P. Larrañaga, R. H. Murga, M. Poza, C. M. H. Kuijpers

An exact probability metric for decision tree splitting and stopping - J. K. Martin

Modeling life time data by neural networks - Y. B. Moon, H-J. Kim

Statistical preprocessing for decision tree induction - S. K. Murthy

Detecting complex dependencies in categorical data - T. Oates, D. Gregory, P. R. Cohen

Learning Bayesian networks using feature selection – G. M. Provan, M. Singh Combining statistics and AI in the optimization of semiconductors for solar cells – J. Risius, G. Seidelmann

Data representations in learning - G. Srikantan, S. N. Srihari

Learning from data by guiding the analyst: On the representation, use, and creation of visual statistical strategies - F. W. Young, D. J. Lubinsky

Poster Session 2 Thursday, January 5

Learning multiple relational rule-based models - K. Ali, C. Brunk, M. Pazzani

Picking the best expert from a sequence - R. Bergman, R. L. Rivest

Ploxoma: Testbed for uncertain inference - H. Blau

Comparing the prediction accuracy of statistical models and artificial neural networks in breast cancer – H. B. Burke, D. B. Rosen, P. H. Goodman

Predicting stock returns with genetic programming: Do the short-term nonlinear regularities exist? - S-H. Chen, C-H. Yeh

Truncated Gaussians as tolerance sets – F. Cozman, E. Krotkov

Textual data mining - S. J. Cunningham

Non-Linear dimensionality reduction: A comparative performance analysis – O. de Vel, S. Li, D. Coomans

Two applications of statistical modelling to natural language processing – W. DuMouchel, C. Friedman, G. Hripcsak, S. B. Johnson, P. D. Clayton

Which method learns most from the data? Methodological issues in the analysis of comparative studies – A. Feelders, W. Verkooijen

Abductive reasoning in Bayesian belief networks using a genetic algorithm - E. S. Gelsema

Finding dependencies in event streams using local search -A. E. Howe Robust linear discriminant trees -G. H. John

🔻 On graphical models for dynamic systems – A. Lekuona, B. Lacruz, P. Lasala

Propagation of Gaussian belief functions - L. Liu

MENTOR: A Bayesian model for prediction and intervention in mental retardation - S. Mani, M. Valtorta, S. McDermott

Evaluating and comparing classifiers: Complexity measures -J. K. Martin

Part-of-speech tagging from small data sets – E. Neufeld, G. Adams, H. Choy, R. Orthner, T. Philip, A. Tawfik

Searching for attribute dependencies in Bayesian classifiers - M. Pazzani

Representing and solving asymmetric decision problems using valuation networks - P. Shenoy

A hill-climbing approach to construct near optimal decision trees -X. Sun, Y. Qiu, L. A. Cox, Jr.