

42nd Gregynog Statistical Conference Programme

(2006)

The talks will take place in Seminar Room 1 (2nd Floor, far end).

Friday	16.00	<i>Tea</i>	
21 April	17.30	Prof. Wally Gilks	Leeds University
		<i>Introduction to Bioinformatics</i>	
	19.00	<i>Dinner</i>	
Saturday	08.00	<i>Breakfast</i>	
22 April	09.30	Dr. Christine Currie	Southampton University
		<i>Balancing bias and variance in the optimisation of simulation models</i>	
	11.00	<i>Coffee</i>	
	11.30	Prof. Wally Gilks	Leeds University
		<i>Fusing microarray data</i>	
	13.00	<i>Lunch</i>	
		<i>Afternoon free</i>	
	16.00	<i>Tea</i>	
	17.00	Prof. Mike Kenward	London School of Hygiene and Tropical Medicine
		<i>Some practical applications, and issues, with multiple imputation</i>	
	18.30	<i>Dinner</i>	
	19.45	Prof. Chris Glasbey	Biomathematics & Statistics Scotland, Edinburgh
		<i>Image restoration, segmentation and warping using generalizations of dynamic programming</i>	
Sunday	08.00	<i>Breakfast</i>	
23 April	09.30	Prof. Charles Taylor	Leeds University
		<i>Kernel methods in statistical learning</i>	
	11.00	<i>Coffee</i>	
	11.30	Prof. Wally Gilks	Leeds University
		<i>A statistical approach to distance-matrix phylogenetics</i>	
	13.00	<i>Lunch and finish</i>	

Speakers

Dr. Christine Currie	Southampton
Prof. Wally Gilks	Leeds
Prof. Chris Glasbey	Biomathematics & Statistics Scotland, Edinburgh
Prof. Mike Kenward	London School of Hygiene and Tropical Medicine
Prof. Charles Taylor	Leeds

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Prof Neville Davies

Southampton

Prof Russell Cheng

Swansea

Alan Meyer	Owen Bodger	See Ju Chua
Alan Watkins	Hannah Finselbach	

Warwick

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Mohand Feddag	Theodore Papamarkou	Demetris Lamnisos
John Fenlon	Beatriz Penaloza	Christopher Howitt
Masayuki Henmi	Katherine Boyd	Claudia Lozada-Can
Prof Jane Hutton	Maria Costa	Michalis Kolossiatis
Jen Marsh	Peter Kimani	Patrick Ho
Prof Mark Steel		Miland Joshi
Wenjuan Zhang		

Abstracts

Three sessions on Bioinformatics
Professor Wally Gilks
Leeds University

1. Introduction to Bioinformatics

The field of bioinformatics has grown up around the huge and rapidly expanding genomic and related databases, including the human genome sequence. These open access databases and the many servers providing tools to interact with them present fresh challenges to statisticians. I will present a very brief overview of the field, and indicate where statisticians might try to get involved.

2. Fusing microarray data

Microarrays can be used to measure the activity of tens or hundreds of thousands of genes simultaneously.

The data generated are typically very noisy and microarray experiments may be poorly reproducible.

Ideally, therefore, experiments should be replicated in several ways and in several laboratories before scientific conclusions are drawn. To make the most of such data, we propose a method, based on multivariate regression, for microarray data fusion. We apply the method to data arising from cell cycle experiments in yeast.

3. A statistical approach to distance-matrix phylogenetics

Phylogenetics is the study of the evolutionary relationships between different species, typically represented in the form of a tree. Some quite sophisticated statistical methods have been employed to estimate phylogenetic trees on the basis of DNA sequences from each species. However, for very large phylogenies, such methods are too slow. More computationally tractable methods are based on a distance matrix constructed from the sequence data.

However, such methods have weak statistical underpinnings, in particular with regard to handling uncertainty.

We propose an agglomerative regression-based method for distance-matrix phylogenetics, having a sound statistical foundation, but still retaining the computational efficiency of simpler methods.

Balancing bias and variance in the optimisation of simulation models
Dr. Christine Currie
Southampton University

We consider the problem of identifying the optimal point of an objective in simulation experiments where the objective is measured with error. We describe some simple simulation experimental designs that emphasize the statistical aspects of the process. When the objective can be represented by a Taylor series near the optimum, we show that the best rate of convergence of the mean square error is when the variance and bias components balance each other. More specifically, when the objective can be approximated by a quadratic with a cubic bias, then the fastest decline in the mean square error achievable is $n^{-2/3}$. Some elementary theory as well as numerical examples will be presented. Comparisons between the method described here and the established algorithm of Stochastic Approximation will also be made.

Image restoration, segmentation and warping using generalisations of dynamic programming

Chris Glasbey

Biomathematics and Statistics Scotland

(Pdf of talk available from <http://www.bioss.ac.uk/~chris>)

Dynamic programming (DP) is a fast, elegant method for finding the global solution to a class of optimisation problems. For example, it can be used to find maximum a posteriori (MAP) estimators of boundaries, to automatically segment 2-D medical images into anatomical regions (Glasbey and Young, 2002). A variant, dynamic time warping, can also be used to align pairs of tracks in 1-D electrophoresis gels. However, for many image problems, including 3-D segmentation, 2-D warping and image restoration, DP is not possible.

We consider three generalisations of DP for image restoration, segmentation and warping. The first approach is a greedy algorithm first proposed by Leung et al (2004), termed iterated dynamic programming (IDP), where DP is used to recursively solve each of a sequence of lower-dimensional problems in turn, to find a local optimum. A second algorithm replaces DP by a more computationally intensive Forward-Backwards Gibbs Sampler (Scott, 2002), and uses a simulated annealing cooling schedule to guarantee the optimal solution. The final approach is an empirical, stochastic optimiser, which is implemented by adding noise to IDP. Results are compared with existing pixel-by-pixel methods of iterated conditional modes (ICM) and simulated annealing, and illustrated using data from synthetic aperture radar (SAR), 3-D X-ray computed tomography and 1-D and 2-D electrophoresis gels.

Glasbey, C.A. and Young, M.J. (2002). Maximum a posteriori estimation of image boundaries by dynamic programming. *Applied Statistics*, 51, 209-221.

Leung, C., Appleton, B. and Sun, C. (2004). Fast stereo matching by Iterated Dynamic Programming and quadtree subregioning. *British Machine Vision Conference* (Eds. A Hoppe, S Barman and T Ellis) 1, 97-106.

Scott, S.L. (2002). Bayesian methods for Hidden Markov Models: recursive computing in the 21st century. *Journal of the American Statistical Association*, 97, 337-351.