Issue 1.6

Modelling time in health care systems

December 2004

Previous issues at http://www2.wmin.ac.uk/coiec/nosokinetics.htm

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As it's the last month of the year and the festive season will soon be upon us I thought I would give you a quiz. There's only one question and the answer is on page six.

What does it say? Where could it have come from? How was it made?

We have much to celebrate, thanks to you all we now reach over 500 people world wide.



Here's a clue.

Depends on how you see it.

A recent paper in the Australian Medical Journal concludes, using bed usage per thousand, that ageing of the population between 1993 and 2002 was not associated with an increase in the proportion of hospital beds used by older patients. Indeed, in relation to the population aged 75 and over bed utilisation declined by 10%.

Figure 1 paints a different picture of bed usage. The problem arises, because beds per thousand of the population masks the trends in total bed days or separations and does not address the issue of supply. These issues have important ramifications for policy decision-making.

Gray LC, Yeo MA et al (2004) MJA;181: 478-481





Data Source: Australian Institute of Health and Welfare

Copyright infringement: In the October Issue we inadvertently published a figure from Knowledge Discovery and Data Mining (1996), U. M. Fayyad, G. Pastetsky-Shapiro, P. Smyth and R. Uthurusamy (eds), page 10. Copyright © 1996 American Association for Artificial Intelligence. The web version has been altered to correct our mistake.

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Modelling long-term care placement: 86% success using a computational evolutionary heuristic <u>Ruxandra Gorunescu</u>, University of Craiova, Romania

Learning Classifier Systems is a sub-area of Symbolic Empirical Learning which considers the rules that underpin decision-making as IF-THEN rules. The aim is to create rules from a training data set, which make correct decisions in another data set. Given a set of training events, evolutionary computational heuristics are able to produce the classification rules. Expert opinion is an important starting step.

We applied a Genetic Chromodynamic approach to placement data from the Social Service Department of the London Borough of Merton. Starting rules: independent, continent, self-feeding to residential care, dependent, incontinent, needs feeding to nursing care. Initial computational results were encouraging and indicate the suitability of applying evolutionary algorithms to patient classification.

Prior Evolutionary Computing Approaches to Learning Classifier Systems		2000 Genetic Chromodynamics as a multi-modal engine (D. Dumitrescu, 2000)	
Michigan Approach 1978	Pittsburgh Approach 1980	Maintains multiple solutions in a multiple sub- populations environment.	
A chromosome is a rule. The entire population represents the rule set.	A chromosome is an entire rule set.	Forms and maintains subpopulations, of an arbitrary structure, that co-evolve and lead, at convergence, each to an optimum	
The last population gives the correct rule set.	The best chromosome gives the correct rule set.	Uses a variable sized population and a stepping- stone search mechanism in connection with a local interaction principle	
Standard evolutionary heuristics are doubled by a credit assignment system as a multi-modal engine.	Standard evolutionary heuristics may be used.	Crossover and mutation are exclusive operators when applied to the same chromosome. It introduces a new operator for the merging of very similar chromosomes.	

Overall, 86.4% were correctly placed: four were incorrectly placed and ten were not placed.

Table 1. Comparison of actual and modelled placement decisions						
Decision of Merton Social		Placement made by the Evolutionary Model				
Service Panel		Correct	Incorrect	Not placed		
Residential Home	(39)	35 (89.7%)	0	4 (10.2%)		
Nursing Home	(51)	43 (84.3%)	3 (5.9%)	5 (9.8%)		
Nursing Long Stay	(13)	11 (84.6%)	1 (7.7%)	1 (7.7%)		
Overall (103)		89 (86.4%)	4 (3.9%)	10 (9.7%)		

Possibly more of the same data would improve the results. Alternatively, and more likely, other facts concerning the clients or the hidden behaviour of the decision makers are needed to match the decision making of the panel. For some of the decisions made seem to break the panel guidelines and rules.

Reference: Dumitrescu, D., "Genetic Algorithms and Evolution Strategies", Blue Publishing House, Cluj - Napoca, 2000

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Using Ogive Plots to display occupancy statistics: Mark Mackay

In previous editions of Nosokinetic News, the issue of length of stay data being skewed has been raised. This article continues the theme of looking at how data may be displayed. On this occasion, one method of illustrating bed occupancy data will be described.

Ogive plots display cumulative frequency distributions. Figure 1. This example shows the number of beds occupied in an acute hospital on an average day and the number of beds occupied for x or more days. Using a discharge data set, and EXCELL 365 bed census days were used to create the average day.

This is the pattern of bed occupancy that we would find if we walked round the acute wards on one day collecting the dates of admission of all inpatients. Of course, dependent on the time of day the total beds occupied would change: more at midday, less at midnight.

Ogive plot: Pattern of hospital inpatient bed usage with 95% confidence intervals



Figure 2: The inclusion of confidence intervals shows the extent of variation around each point in the cumulative frequency distribution (note a different data set).



Figure 1: A cumulative frequency graph. Here the cumulative frequency represents the average number of beds occupied for x or more days. This is known as a "descending" or "more than ogive".

The addition of confidence intervals provides some degree of indication of the variability around the profile. NB the line was dropped to show the confidence intervals.

The advantage of using this type of illustrative technique is twofold. First, an appreciation of the number of beds required for patients with particular periods of stay can be gained. Secondly, given data sets, where the admission and discharge times are recorded, then these graphs can be constructed for particular times of day and can be used to profile occupancy at the busiest time of day (see Mackay and Gorunescu, 2001). Thus overcoming the problem with the average daily length of stay statistic, which provides information but no knowledge about bed occupancy at differing times during the day.

additional information:Kirk RE (1990). Statistics: An Introduction. Holt Rinehart and Winston, Inc. Florida er, H (1984). Statistics for business and economics. Scott Foreman and Company. United States of Iackay, M and Gorunescu, F (2001). Midnight Bed Census, Patient Length Of Stay And Bed Occupancy. Proceedings of the 10th International Symposium on Applied Stochastic Models and Data Analysis, Vol 2 pages

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USEFUL SITE: Data Warehouse on Trends in Health and Aging <u>www.cdc.gov/nchs/agingact.htm</u>

Silver, E. A. (2004). An overview of heuristic solution methods. JORS; 55: 936-956

Written for operational researchers and managers interested in the use of mathematical models to aid decision-making the paper focuses on the different ways that models can be used to find a solution. Each method discussed is referenced and guidelines are given..

Harper, P. R. and M. A. Pitt (2004). On the challenges of healthcare modelling and a proposed project life cycle for successful implementation. JORS; 55: 657-661.

The size, complexity, interactivity and continuous dynamism of healthcare represents a particular challenge for modellers who seek to develop workable abstractions of underlying systems. Organisational and strategic changes should not be considered in isolation, as changes in one part of the system can have adverse effect elsewhere. Based on their experience a 13-point project life cycle is proposed, beginning with a steering group and ending with promotion of the results. The paper ends with a plea for the greater dissemination of successful projects to a wider audience.

Good news

'MASHNET' - A new network for modelling and simulation in healthcare

The EPSRC has awarded a three year grant of £60K to fund 'Mashnet'. Starting in early 2005, Mashnet aims to bring together the three communities of health services, academic research and industry to establish a common framework of communication within and between these domains, to promote more successful research and implementation. For further information contact <u>Dr Martin Pitt</u>, decision analyst and health modeller, at the Peninsula Technology Assessment Group based in the new Peninsula Medical School in Exeter.

Primary Prevention in older people works

<u>A meta-analysis of 19 randomised trials</u> indicates that preventive primary care outreach interventions to older people reduces the risk of dying by 17% and increases the likelihood of living in the community by 23%. The trials are a mixed bag, but the findings support the concept that early intervention, when dependency changes, is beneficial. Yet, a surprising conclusion is that admissions to hospital do not increase. This is contrary to

Flying against the hype; disease management, quality costs

"Better management, improves quality and save costs". Going against the stream, a study of disease management (1996-2002) of four diseases in Kaiser Permanete showed that improving quality increased the costs. The authors conclude that the causal pathway -from improved care to reduced morbidity to cost savings - has not produced sufficient savings to offset the rising costs

Fireman, B., J. Bartlett, et al. (2004). "Can disease management reduce health care costs?" Health Affairs 23(69): 63-75.

Forthcoming conferences: also see http://www2.wmin.ac.uk/coiec/nosokinetics.htm

IFORS Hawaii? July 11-15, 2005: Website

We have been invited to contribute a health stream session: contact Peter Millard

Young OR, Bath? April 4th to 6th, 2005:

For further details or abstract submission please contact Adele Marshall or submit at Website

25th Applied Statistics in Ireland Conference, Enniskillen, 18th-20th May 2005. Abstracts due 31st March 2005. Contact Adele Marshall

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British Nosokinetics Group

On 4th November 2004, at Westminster University, the British Nosokinetics Group was founded. Chairman; Prof Sally McClean; Secretary, Dr Adele Marshall; Conferences, Dr Elia ElDarzi; President; Prof Peter Millard.



Epilogue: Thank you for your support during 1994. The Nosokinetics Group wishes you all the best for a Happy and successful 2005. For earlier editions <u>http://www2.wmin.ac.uk/coiec/nosokinetics.htm</u> The next newsletter will come out in February. <u>Editor: Prof Peter H Millard</u>

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