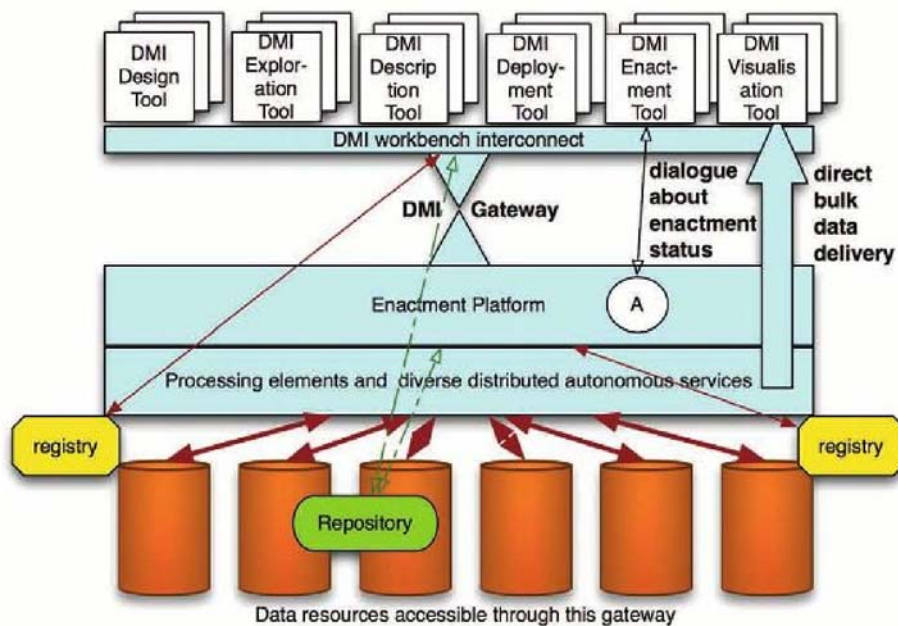


# Burrowing into the bran tub

*Felix Grant assesses statistical information mining techniques*



Schematic illustration of the ADMIRE data management Integration model

Intelligence organs of a fairly small, but technologically proficient, country recently managed to clone the email and document bases of a fairly large, but unsuspecting, company with an active and diverse research operation. Not the research results bases

themselves; just the email and document archives. The immediate benefits of the theft were obvious, mundane, and irrelevant here. More interesting from a scientific computing viewpoint (if no more morally or legally defensible) were the spinoffs from applying

statistical information mining techniques to the combined database contents.

As a direct result of the operation other companies, seen as friendly to the small country's interests, received repackaged and anonymised material suggesting productive new lines of enquiry. Patents resulting from these have already been filed; others are in the pipeline. The large company that was the target of the operation, significantly, remains unaware of the opportunities that exist within components of its own activities, which have never been brought together.

Or so, at least, I am told. I have no way to verify the truth of this story, of course. But the young man who tells it to me, sitting in the café at the National Gallery, his hands continually roving across the keyboard and digitiser pad of a notebook as he talks, has never sold me a lemon yet. And, true or not, it illustrates a truth: that much knowledge is locked away in information stores assembled for one set of reasons and never reexamined in other ways.

Less melodramatically, and less dubiously, information openly published on the internet forms a huge field within which to prospect potential information seams – 'The low user entry barrier of the web has resulted in massive amounts of unstructured and weakly structured data referring to objects, concepts, user interests and communities,' to quote the Digital Enterprise Research Institute at

## Keep taking the tablets

David Smith, pharmaceutical solutions architect for SAS UK, comments: 'Most data mining techniques can be applied to pharmaceutical data, and have great potential for answering a wide variety of questions across all parts of the business.'

'Association analysis, for example, can help enterprises to understand which drug combinations are associated with adverse events, and which associations are strong

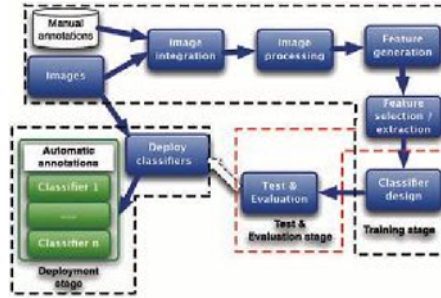
enough to be investigated and addressed. Segmentation techniques can help identify groups of physicians who are potential investigators in a new trial.

'Variable selection can be used to answer questions, such as which genetic markers determine whether a patient will respond well to a treatment, or which process parameters predict quality control failure of a production batch.'

'Mining of linguistic information helps to gather insight into pharmaceutical methods and techniques. For example, text mining allows pharmacists to identify groups of adverse events associated with a particular drug, a so-called syndrome effect. Content categorisation can highlight potential drug safety concerns suggested by internet tweets and blog entries, or point to abstracts that researchers should prioritise.'

Galway. As SAS's David Smith points out (see 'Keep taking the tablets'), tweets and blog entries can contain pointers to early identification of potentially vital phenomena. This is an aspect of what is known in the industry as pharmacovigilance, which 'can be defined as a set of practices aiming at the detection, understanding and assessment of risks related to the use of drugs in a population, and the prevention of consequential adverse effects [or] in a narrower sense ... postmarket surveillance'.<sup>[1]</sup>

A leader in making such text searching accessible to smaller, nontraditional users and demonstrating the value of placing intelligent defaults in their hands, is the Data Miner Recipes (DMR) tool in Statsoft's Statistica. It provides a clear cut, step-by-step path through a data mining project from initial cleaning and preparation of the data through to building and evaluating a model. It doesn't quite pass my '10 year old test' (in which I ask a child to try and use a software tool for a practical purpose), but that is mainly a cognitive difficulty in grasping the ideas involved. Over the past month, on the other hand, I have seen a dozen 14-year-olds embrace my copy with enthusiasm and derive meaningful results from experimental data in a few mouse clicks. Once the data file and variables are selected, the project can carry the user through to finished models without much interference – though any degree of sophisticated control is possible at every stage. There are other approaches, but



The data mining framework used by Han et al to automate annotation of gene expressions<sup>[8]</sup>

the DMR wizard removes most of the barriers to initial adoption and familiarity.

Extracting connections from text bases is, of course, neither the only way to skin a data set nor separate from other approaches. I've recently been watching a team of farmers without formal statistical training use DMR to interrogate a combination of numeric and text data. Seeking exploitable patterns in the behaviour and influence of natural pollination vectors, they join a wide range of observational and administrative records using time as a common key and then let DMR do the rest.

Agriculture is a rich recipient of the benefits accruing from information mining approaches. A quick dip into the literature on trait selection over the past year showed them to be behind six of the first seven results: adipocytes<sup>[2]</sup> and growth<sup>[3]</sup> in meat stock, milk production<sup>[4]</sup>, neuroendocrine correlation in poultry<sup>[5]</sup>, protein interactions in yeast<sup>[6]</sup>, and crop breeding<sup>[7]</sup>. This is not surprising, since

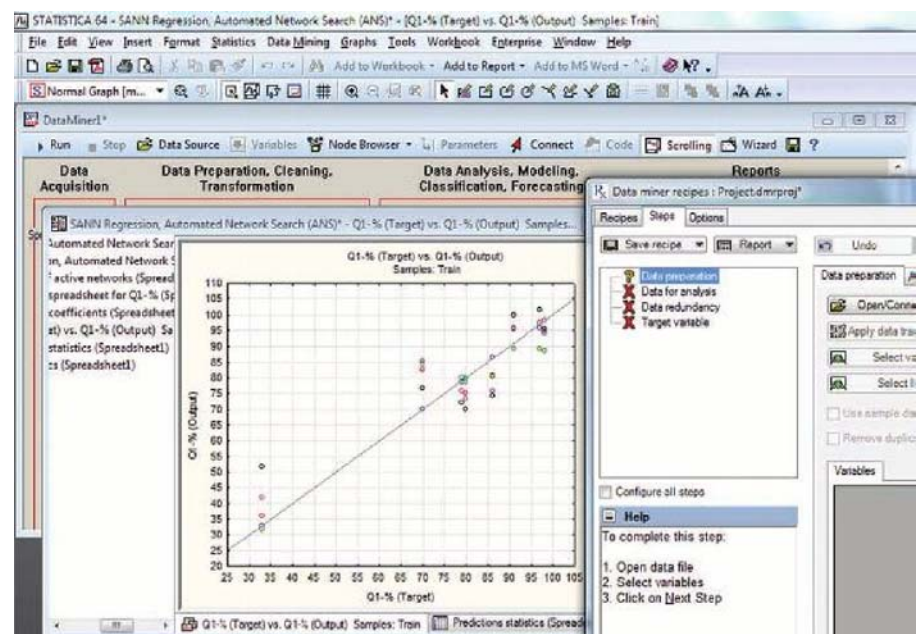
information mining is one of the stable of new methods that accompany the explosive blossoming of genetics. As the editor-in-chief of the Federation of American Societies for Experimental Biology's *FASEB Journal* observed a few months ago (see 'Science in the information age'), information mining rides a wave of new approaches which, in aggregate, represent a radical shift away from traditional hypothesis-based science. Or, to borrow a Dutch colleague's picturesque metaphor, 'We no longer ask Sinterklaas and Zwarte Piet for a specific named present at the beginning of the research season and wait patiently to see whether we get it at the end; we go and burrow through their lucky dip bran tub to see what they've got.'

This is one of those areas where traditional hypothesis-based approaches may never uncover a linkage, because there is nothing to suggest the hypothesis in the first place. Or, alternatively, one function of information mining can be seen as an enhancement of intuition, dramatically accelerating the rate at which hypotheses can be generated.

While the benefits of information mining are real and valuable at the small user end of the research spectrum, they grow exponentially with scale and are significant building blocks for science and for national or regional economies. At that scale, methods development has to be explored outside the user or provider framework. In Europe, the EPCC (Edinburgh Parallel Computing >

## Science in the information age

Gerald Weissmann, editor-in-chief, *FASEB Journal*, says: 'But the 'omic revolution has not just given us new facts: it has changed the way biologists think. Pioneers of 'omics and systems biology claim that they have either overturned traditional, hypothesis-driven research completely, or at the very least found an alternate way to do science. The novel techniques of microarray analysis, of "connectivity" or "molecular interaction" mapping, of kinetic simulation of cell processes have been made possible by information technologies that owe as much to Oracle and SAP as to Krebs and Chargaff. As discovery research becomes replaced by information-mining, data no longer lead to hypothesis, but make hypothesis unnecessary.'<sup>[12]</sup>



Wizard assisted text mining in Statistica



**Crossing main authors of the 'Hipparcos' collaboration with topical key words. From Egret et al, *Information mining in astronomical literature with TETRALOGIE* <sup>[11]</sup>**

- Centre) manages the ADMIRE (Advanced Data Mining and Integration Research for Europe) project, which 'aims to deliver a consistent and easy-to-use technology for extracting ... meaningful information by data mining combinations of data from multiple heterogeneous and distributed resources... which will give users and developers the power to cope with complexity and heterogeneity of services, data and processes'.

ADMIRE emphasises the need to accommodate the increasing size of information stores, sophistication of extraction requirements, and complexity of the resulting systems, responding with a unified conceptual structure based on integration of separate expertise types in relation to a defined component library structure. The outputs are 'a framework, an architecture and a set of use cases that illustrate how they can be used to

improve DMI [data management integration]'. Alongside conceptual development and implementation, it has applications on which its methods can be demonstrated in practice as well as a growing number of studies that call on its capabilities.

Sticking, for the moment, with my genetics thread, ADMIRE-supported projects include automated gene annotation (see 'Labelling the building blocks') through 'a new extensible data mining framework that integrates both the images in the file systems and annotation databases and combines image processing with statistical pattern recognition techniques to automatically identify gene expressions in images'<sup>[8]</sup>. A different example, picked from the list of accepted papers on the ADMIRE website, just because it appeals to me, is 'An ant-colony-optimisation based approach for determination of parameter significance of scientific workflows'<sup>[9]</sup>.

Ant colonies segue me neatly away from life sciences to social network analysis (a topic upon which I focused in the last issue), where once again massive data sets contain numerous yet to be discovered relations. Those relations may be anything from individual two-node edges to whole complex networks, or a blend of the two in the form of association between one or more unsuspected individual nodes and a network or networks or intermediate structures, such as cliques.

This is one of those areas, mentioned above, where hypotheses emerge where they otherwise might not, because information mining throws up associational links and

## Labelling the building blocks

Liangxiu Han and others<sup>[8]</sup> (conclusions of an ADMIRE supported study) state: '...we have developed a new data mining framework to facilitate the automatic annotation of gene expression patterns of mouse embryos. There are several important features of our framework: (1) the combination of statistical pattern recognition with image processing techniques can help to reduce the cost for processing large amounts of data and improve the efficiency. We have adopted the image processing method to standardise and denoise images. Wavelet transform and Fisher Ratio techniques have been chosen for feature generation and feature extraction. The classifiers are constructed using LDA. (2) For enhancing the extensibility

of our framework, we formulate our multi-class problem into a two-class problem and design our classifiers with a binary status: 'yes' or 'no'. One classifier only identifies one anatomical component. Classifiers for each gene expression are independent on each other. If new anatomical component need be annotated, we do not have to train previous classifiers again. The classifiers can be assembled and deployed into the system based on user requirements. (3) We have evaluated our proposed framework by using images with multi-gene expression patterns and the preliminary result shows our framework works well for the automatic annotation of gene expression patterns of mouse embryos.'

## Keeping a grip on the bigger picture

'Algorithms that work on complex networks with hundreds to thousands of vertices often disintegrate on real networks with millions (or more) of vertices. For example, betweenness centrality is not robust to noisy data (biased sampling of the actual network, missing friendship edges, etc.) [They require] niche computing systems that can offer irregular and random access to large global address spaces. ... the newest breed of supercomputers have hardware set up not just for speed, but also to better tackle large networks of seemingly random data. ... Applications include anywhere complex webs of information can be found: from internet security and power grid stability to complex biological networks.'

David A Bader, Georgia Institute of technology<sup>[13]</sup>.

matrices of links not obvious to unaided human intellectual perception.

The results can be dramatic. A single node that is discovered to have links with two (or more) discrete networks, for instance, becomes a bridge and instantly, by definition, converts those networks into (potentially highly complex) cliques of a single larger network. Discovery of the subtle clues that a particular node acts as a subliminal bellwether can shift weightings in the whole network of which that node is a part.

Professor David Bader, from the Georgia Institute of Technology (see 'Keeping a grip on the bigger picture') and working with the Pacific North West Laboratory (PNNL) under a Center for Adaptive Supercomputing Software for Multithreaded Architectures (CASS-MT) sponsorship umbrella, is one of many who are developing HPC software for supercomputer handling of this kind of problem. Across a number of publications he documents his group's development tools for analysing massive streaming data sets on Cray machines, currently the XMT which 'has the unique ability to process massive volumes of irregular, data-intensive applications, and is ideal for graphs-data that connects to other data in varying ways'<sup>[10]</sup>.

## References and sources

For a full list of the sources and references cited in this article, please visit [www.scientific-computing.com/features/referencesaug10.php](http://www.scientific-computing.com/features/referencesaug10.php)