

#### e-Science & Text Mining a marriage made in heaven?

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# e-Science & Text Mining

## Motivations

- Add text mining services to e-Science projects
- Use e-Science technologies to build text mining services

## Approach

- Work with end user Scientists
- Architecture for distributed text mining based on Discovery Net



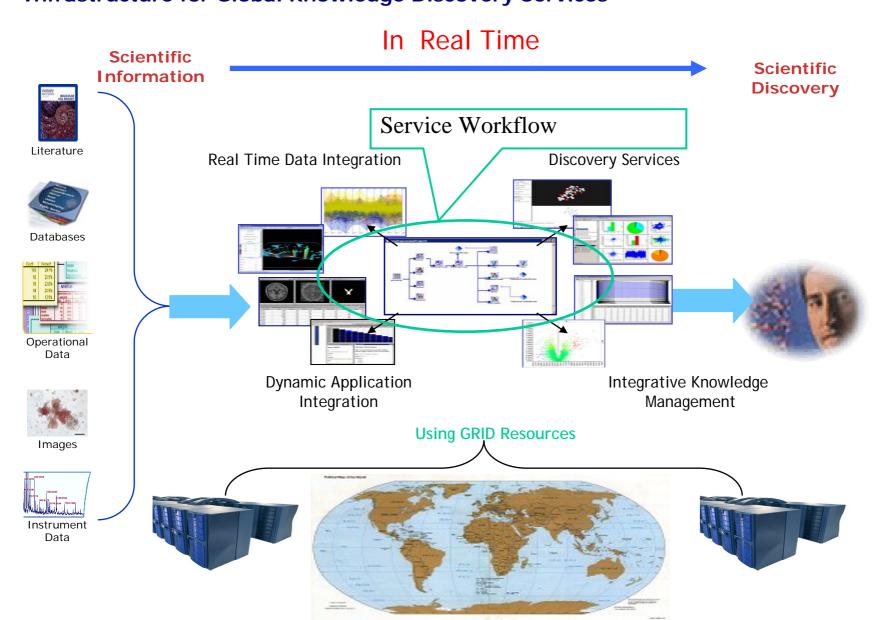
## e-Science & Text Mining Projects at Imperial College

- Discovery Net:
  - £2.2M EPSRC Pilot Project, ends in March 05
  - Service-based infrastructure
  - Meta data models for mixed data mining / text mining
- Real-time Text Mining:
  - £125K EPSRC Best Practice, start Oct 04
  - Collaboration with myGrid: Service interoperability
  - Automatic Annotation of Medline Documents with GO codes
- Integrative Biology in silico: Applications of Advanced Informatics to Systems Biology
  - £550K BBSRC BEPII Project, start Oct 04
  - Using Text Mining to interpret Discovery Results
  - Insulin Signalling



## Infrastructure for Global Knowledge Discovery Services

#### Discovery Net e-science infrastructure



## **Challenge:** Meeting Scientists' Requirements

#### **Scientist's Requirements:**

Unifying the World's Knowledge

 I want an easy a user-friendly tool that helps solve my scientific problem. There is a wealth of knowledge in the literature and I want to tap in. I heard e-Science text mining is great.

#### What Computer Scientists say:

DIS Discovery Net

- Information Retrieval: Google, PubMed, ....
- NLP: Entity/Information Extraction, Lexicons, Anaphora Resolution, Semantic Ambiguities,
- Data Mining: Machine Learning, Statistics, Classification, Clustering, etc
- **Grid Computing:** Large Data-sets, Distributed Data Sources, Compute-intensive Tasks, ---- Condor, Globus, OGSA, GT3, GT4, GT2010..
- **e-Science:** Workflow, Service Computing, Ontologies, Metadata, Information Integration, Semantic Grid, ...

#### **Scientists say:**

- ???, Can't I just have a better Google!!

#### **Challenge:**

- Fill Gap between what scientists want and what technology provides



#### Application Context: Biological Atlas of Insulin Resistance

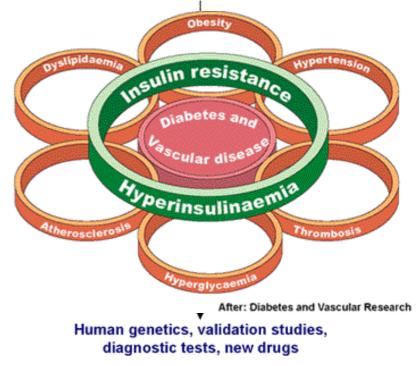
http://www.bair.org.uk/

Working in collaboration with scientists working on the £5.5M BAIR project funded by the Wellcome Trust.

#### The Biological Atlas of Insulin Resistance

The knowledge in the Atlas will lead to a new and fundamental understanding and classification of the causes of insulin resistance and the processes leading to its development. This information will be used as a platform for studies of the causes of insulin resistance in humans, and the basis for more rational and effective strategies for its prevention and treatment than are currently available.

We have adopted the term "atlas" because it emphasises that our aim is a description of the insulin-action "universe" that incorporates and integrates many different types of information, is flexible enough to cope with continuous updating and revision, and capable of serving as a discovery tool – for example, allowing insulin-resistant states of unknown aetiology to be "mapped" onto stable metabolic coordinates.



Genome variation and environment

Prof James Scott, FRS.

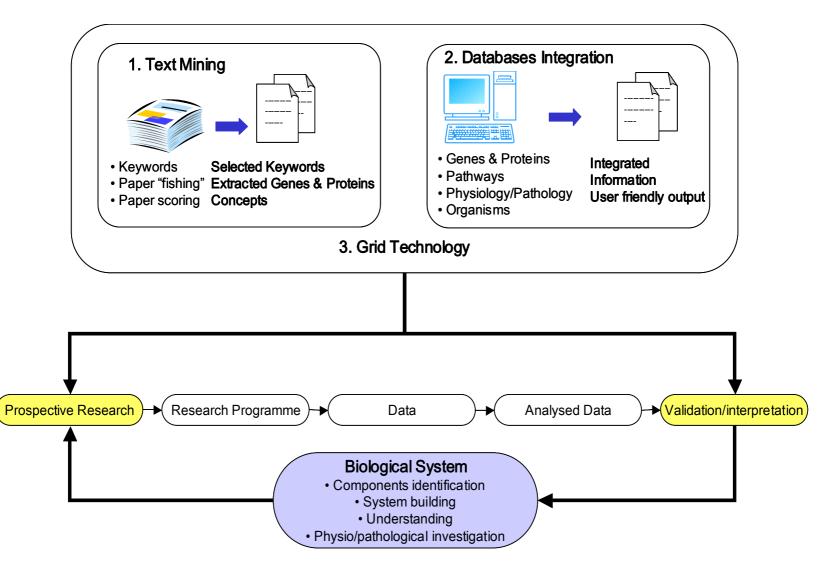


#### **Integrative Biology Example**

Unifying the World's Knowledge

#### What Scientists actually from Text Mining?

#### Example from Integrative Systems Biology Studies





#### **Summarising the Requirements**

#### What Scientists actually want from Text Mining?

Examples from Integrative Biology Studies

#### Leveraging Literature in Discovery Interpretation

- Automatic Result Annotation
  - (e.g. What is the difference/similarity between microarray results)

#### **Generating Prioritised Document Lists**

- Automatic Document Categorization
- Topic Maps
- Linking literature to available background knowledge

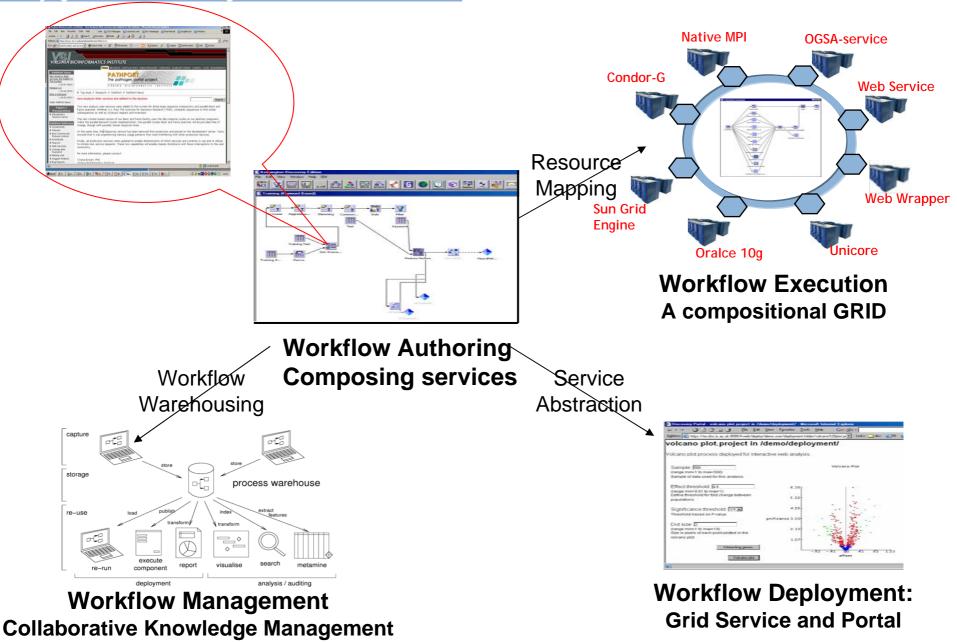
#### Generating New Knowledge

- Information Extraction/Database Curation
  - (e.g. protein-protein interactions)
- Summarising/Comparing IR Query Results
  - (e.g. compare result sets for disease in two stages)



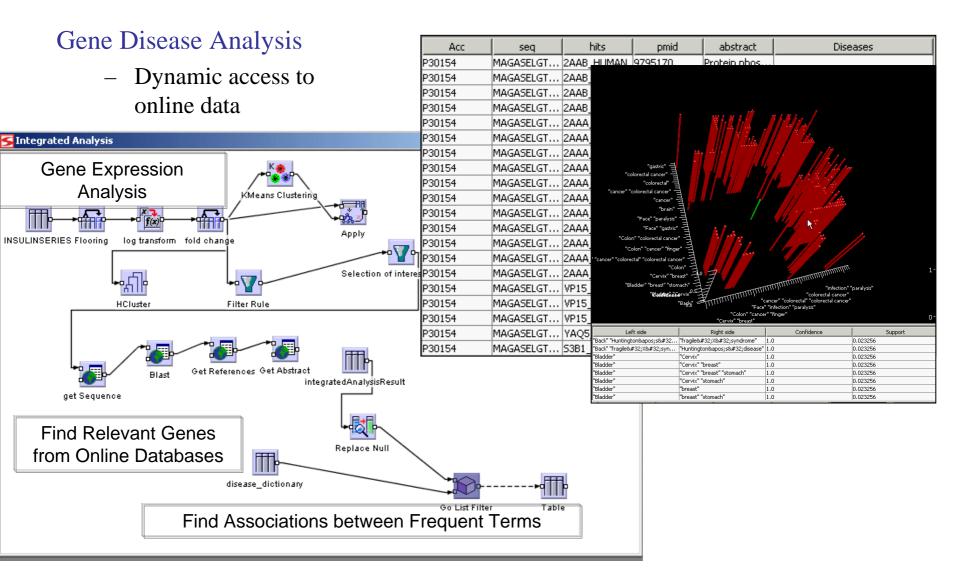
#### **Discovery Net Model**

Unifying the World's Knowledge





#### Adding text mining services to e-Science projects



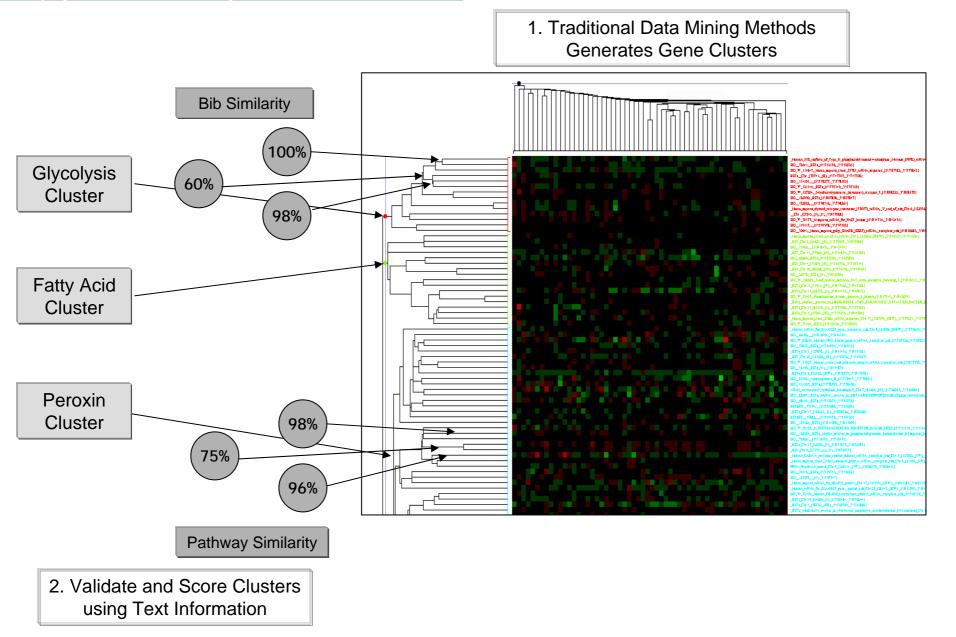
#### Unifying the World's Knowledge

# **Short Animation**

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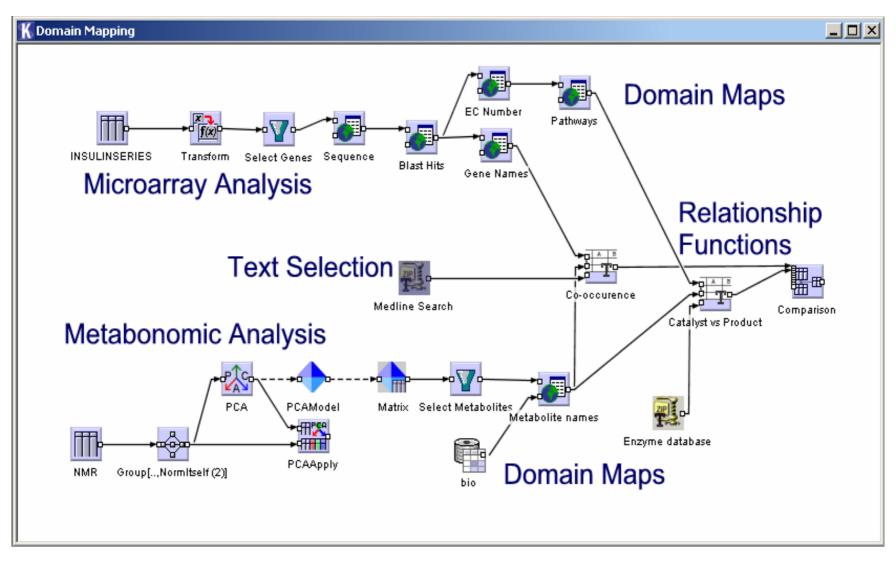
Unifying the World's Knowledge

#### Leveraging Literature in Discovery Interpretation: Gene Group Interpretation



Unifying the World's Knowledge

#### Leveraging Literature in Discovery Interpretation: Domain Mapping



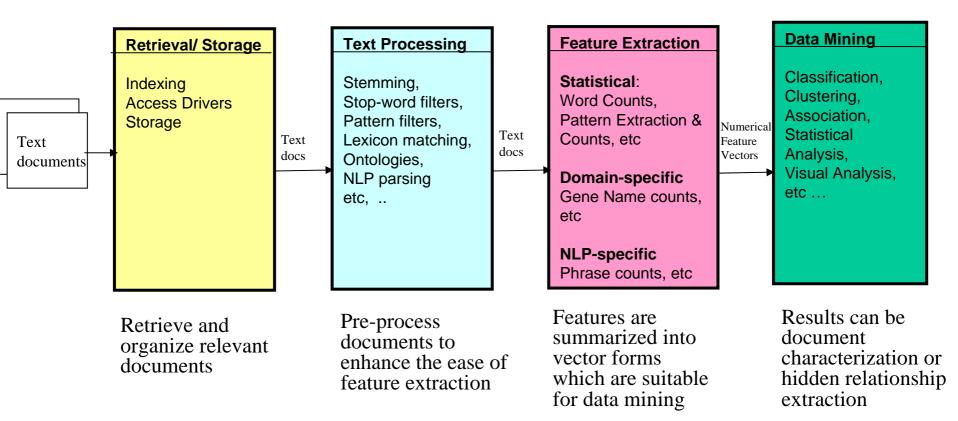
Using literature analysis for the interpretation of gene expression and metabonomics data



#### **Enabling Literature Analysis Workflows**

#### **Use e-Science technologies to build text mining services**

#### **Text Mining Pipelines**





Example 1: Develop a document classification system (KDD CUP 2002)

#### **Training Set:**

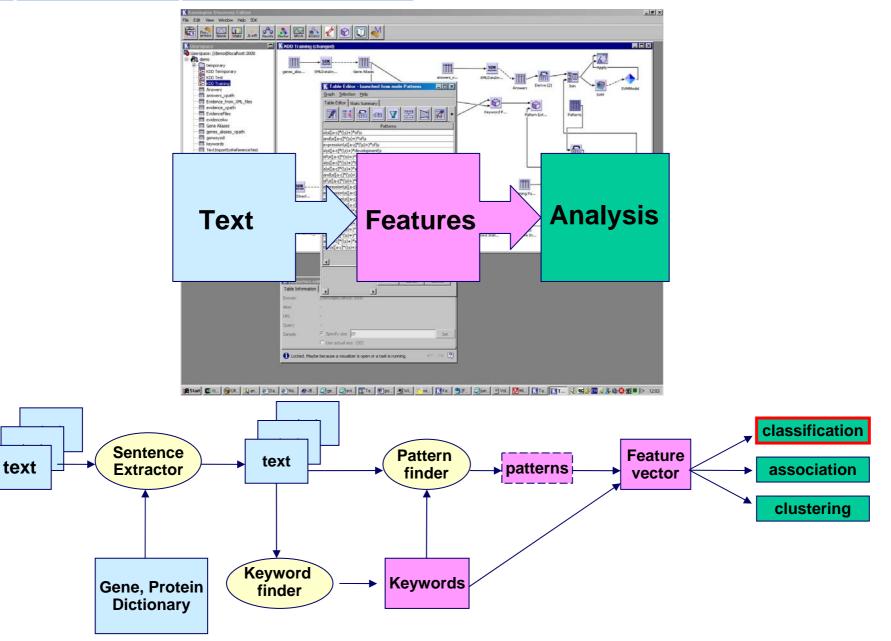
**850 papers on Fruit Fly** (each with the associated gene names identified in a list) were provided. Each paper is labelled as follows:

- 1. Whether it meets the curation criteria or not. (Y/N)
- 2. List of Genes names appearing in paper
- 3. For each Gene name paper a (RNA and/or Protein) label is provided for two types of gene expression products. (polypeptide, transcription, or both).

#### **<u>Task</u>: Identify relevant documents from a previously unseen set**

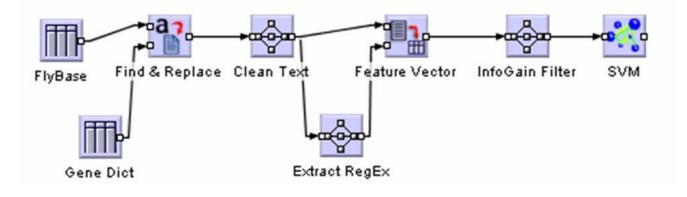
#### Unifying the World's Knowledge

# Application: Identifying Relevant Literature





#### **Text Mining Workflows**

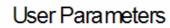


Predictive Accuracy of Relevance prediction, using Support Vector Machine classification

> Overall accuracy: 84.5% Precision 78.11% Recall 73.40%



# HPC implementation of text mining nodes



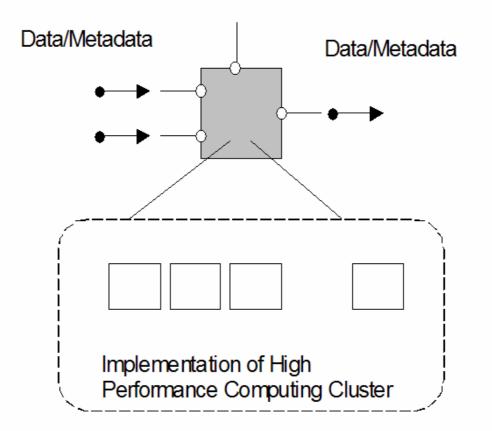


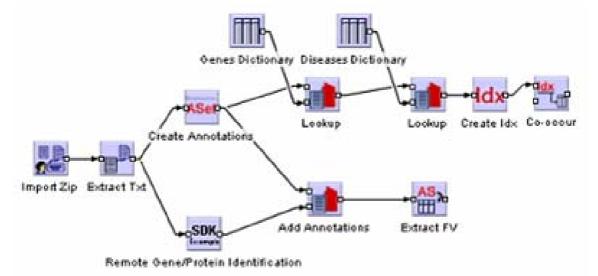
Figure 3: Web service interface / HPC implementation of a text mining component



#### **Text Meta Data Model**

Unifying the World's Knowledge

Service Interfaces only tell you how to invoke remote service but it is up to you to decide what information flows between services !



#### Tipster-based model for in document annotation

Text	Start	End	Annot. Type	Attributes
Insulin	1	7	token	pos:noun, stem:insulin
resistance	9	18	token	pos:noun, stem:resist
Insulin resistance	1	18	compound token	disease:insulin resistance
plays	20	24	token	pos:verb, stem:plai
major	26	30	token	pos:adj, stem:major
role	32	35	Token	pos:noun, stem:role

Loose model allows co-operating text mining services, allows multiple parsing.



# **Why Tipster**

#### Unifying the World's Knowledge

#### **Mining Over Annotated Data**

#### Identify & Extract Biological/Chemical Entities from Documents

- Genes, Diseases, Compounds
- User-defined entities

- ...

#### Query Entity Relationships

- Interactively Select Sentences/Patterns that satisfy annotation constraints
  - Base Entities: e.g. Disease and Gene
  - Entity Properties: e.g. Sentence, Verb, ..

# Extract New Knowledge about Relationships between Entities

- e.g. association rules between entities or their properties
- e.g. phosphorylastion -> TSC2

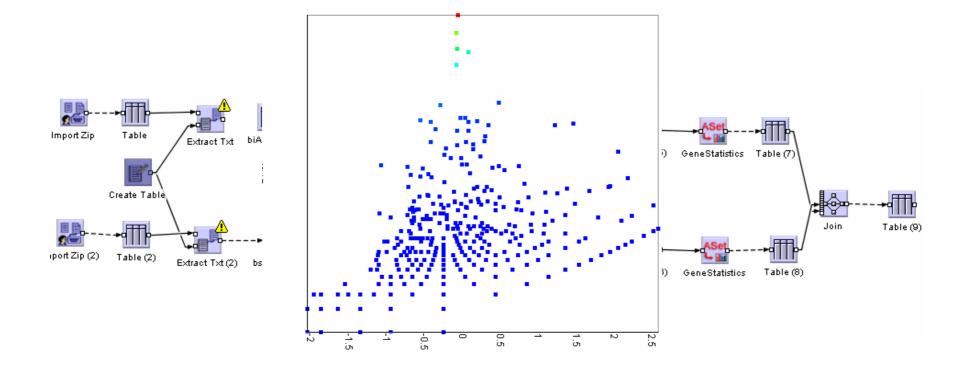
#### Use Annotations as Feature Vectors

- Classification
- Clustering

Query    Homolog of mammalian (&kgr,B kinases (IKKs), The zird5z phenotype and sequence suggest that the gene is specifically required for the activation of Relish, a zDrosophilaz NF-&kgr,B family member.      [zKey Words z Innate immunity, (&kgr,B kinase; zDrosophilaz; antimicrobial peptide; Relish; NF-&kgr,B]
for the activation of Relish, a zDrosophilaz NF-&kgr,B family member. [zKey Words:z Innate immunity, I&kgr,B kinase; zDrosophilaz; antimicrobial peptide; Relish; NF-&kgr,B]
[zKey Wordsz Innate immunity, I&kgr,B kinase; zDrosophilaz; antimicrobial peptide; Relish; NF-&kgr,B]
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Introduction
Introduction
[uarrow.gif] Top
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Laordil Introduction
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In both mammals and zDrosophilaz, microbial infection activates
Toll-like receptor (TLR) signaling pathways as a part of the innate
host defense response (for review, see Anderson 2000 [ref-arrow.gif]
), TLR-mediated signaling pathways are essential for appropriate
responses to bacterial infection. In addition, mouse TIr4 mediates
septic shock associated with infection by gram-negative bacteria
(Vogel 1992 [ref-arrow.gif] ; Poltorak et al. 1998 [ref-arrow.gif] ).
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The available data indicate that different microbial cell wall
components activate different Toll-like receptor signaling pathways.
which regulate distinct sets of target genes. In mammals, TLR4 is the
prime mediator of responses to bacterial lipopolysaccharide, while
6] TLR2 mediates responses to bacterial peptidoglycans (Poltorak et al.
1998 (ref-arrow.gif) ; Takeuchi et al. 1999 (ref-arrow.gif) ; for
review, see Beutler 2000 (ref-arrow.gif) ). The best-studied aspect of
the zDrosophilaz innate immune response is the rapid transcriptional
induction of antimicrobial peptide genes in response to infection
er [

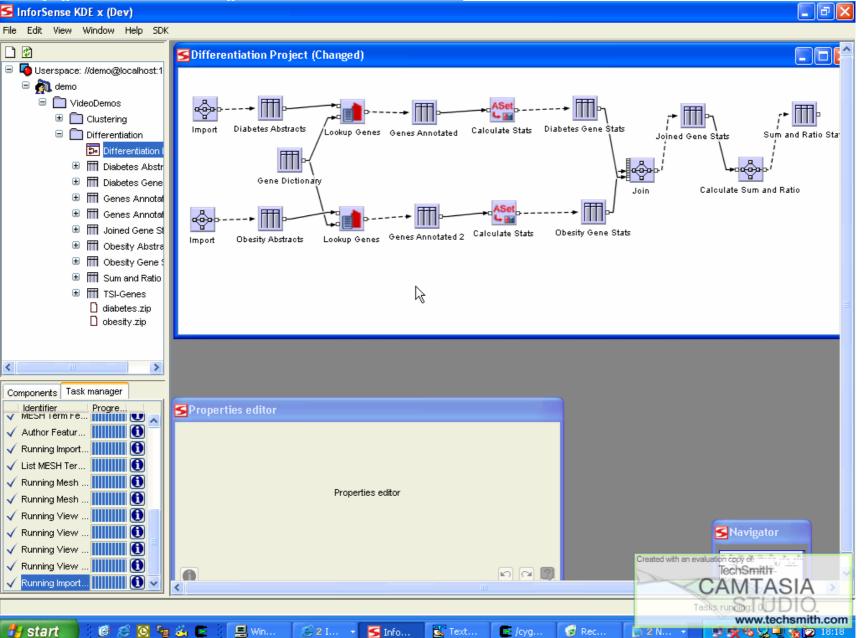


Application: Summarising difference between IR result sets



# **Short Animation**

Unifying the World's Knowledge





#### Discovery Net Text Mining Summary

#### Infrastructure

- Workflow Model for Service Programming
- Annotated Text Data Type
  - Holds both Text and Analysis Results (Annotations)
  - Tipster-based model
- Sparse Feature Vector Data Type
- Integration with DNet data mining tools and algorithms
- Interfaces to Lucene/Oracle Text
- Service Abstraction & Deployment

#### Components

- Import/Export Nodes (Word, PDF, Bibliographic Formats)
- Text Pre-processing Nodes (Clean, Replace, etc)
- Text Annotation/Analysis Nodes (Mark-up entities, N-grams, collocations, POS, etc)
- Indexing Nodes (Manage Large Data Sets)
- Feature Extraction & Mining Nodes (Stats, Compare, Classify, Cluster, etc)
- Viewer Nodes (Visually explore results)



#### e-Science & Text Mining

# e-Science & Text Mining: *Marriage made in heaven or hell?*

- For application developers/ text mining community
  - Workflows & Service Programming gaining momentum in life sciences
  - Open Up Your Text Mining Tools
  - Agree on Interfaces & a Document Annotation/Metadata Models
- For end user scientists
  - No two scientist have exactly the same needs
    - Rapid prototyping/application development through service integration
  - Useful only when details are hidden
    - Workflows deployed higher level services
    - Workflows accessed from friendly user interfaces



## Infrastructure for Global Knowledge Discovery Services

#### In Real Time **Scientific Scientific** Information **Discovery** Service Workflow Literature **Discovery Services Real Time Data Integration** 100 Databases Operational Data **Dynamic Application** Integrative Knowledge Integration Management **Using GRID Resources** Images Instrument Data

**Discovery Net e-science** 

infrastructure