

**Attendance code 16<sup>th</sup> January 2020 is**

# **Intelligent Data Analysis 2020**

**Prof. Martin Russell**

# Course structure 2020

- 11 x 2 hour lectures - Thursday, 11am – 1pm
  - Text retrieval (1) *Intro*
    - TF-IDF similarity, vectorization of documents
  - Maths revision 1: Vectors and linear algebra
  - Dimension reduction
    - Principal Components Analysis (PCA)
  - Visualization of high-dimensional data
    - PCA, Topographic maps, t-SNE
  - Clustering and vector quantization
  - Text retrieval (2)
    - Synonym relationships, Latent Semantic Analysis (LSA), Page Rank
  - Classification

# Assessment

- Assessment
  - 1.5 hour exam in May/June - answer 3 questions from 3
- Assignment for extended module

# Exercise sheets

- Weekly exercise sheets on Canvas
- Solutions will be published on Canvas
- Not part of formal assessment

# Course Canvas page & C code

- All materials will go on Canvas
- Canvas site for 2020 will contain:
  - Copies of all slides and C code
  - Details of assignment - for IDA (extended)
  - Weekly exercise sheets and solutions
  - Pointers to relevant websites
- C code
  - Simple ANSI C implementations of basic techniques from the course. Compile using the MS Visual Studio .NET command line **C compiler**.

# Office hours

- My office hours:
  - Tuesday 2.30-3.30pm
  - Thursday 2.30-3.30pm (until 13/2/2020 – week 5)
  - Friday 2.30-3.30pm (from 21/2/2020 – week 6)
- My office hours will be in my office in the **Gisbert Kapp building, GK-N429**

# Moore's Law and disk capacity

- Moore's Law
  - *Technology performance doubles and prices halve every 18 months*
- Implications for data storage
  - Applies to disk capacity
  - We have the potential to record and store online a big proportion of what we do.

# How much speech fits on 1TB?

- How much speech data can be stored on a 1TB disk?
- Assume:
  - 16kHz sampling rate (16,000) samples per second (Nyquist) *max  $N$  Hz  $\rightarrow 2N$  Hz sampling rate*
  - 16 bits per sample
- Then:
  - 1 second of speech requires 32,000 bytes
  - 1TB =  $3.125 \times 10^7$  s = 520,833 mins = 8,681hrs = 362 days



# Petabytes

- 1 petabyte of disk space costs
  - \$2,000,000 in 2003
  - \$25,000 in 2019 (based on Amazon, \$100 for 4TB!)
- 1 petabyte =  $10^{15}$  bytes
  - $10^6$  – (1MB),  $10^9$  – (1GB),  $10^{12}$  – (1TB)
  - $10^{15}$  – zillion
  - 1 zillion used to be synonymous with infinity – an unimaginably large number!

# A Petabyte is a lot of data...

- 1PB =
  - 20 million 4-drawer filing cabinets filled with text
  - 13.3 years of HD-TV video
- 1.5PB =
  - Combined size of the 10B photos on Facebook
- 20PB =
  - The amount of data processed by Google per day

(A Google search will find many similar examples)

# Accessing data – “aboutness”

- Why store these huge corpora?
- Because **information** in them is potentially **useful**
- But, how can we find the **relevant** items?
  - AV recording of a meeting contains more information than conventional minutes, but only useful with good search functions
- Need to know:
  - What each item in a corpus is **about**
  - Relationships between different corpus items
  - Relationships between ‘queries’ and corpus items
- Manual **indexing** impossible - deal with ‘raw’ data.
- Need to determine **automatically** what a text is **about**

query — corpus  
/ \

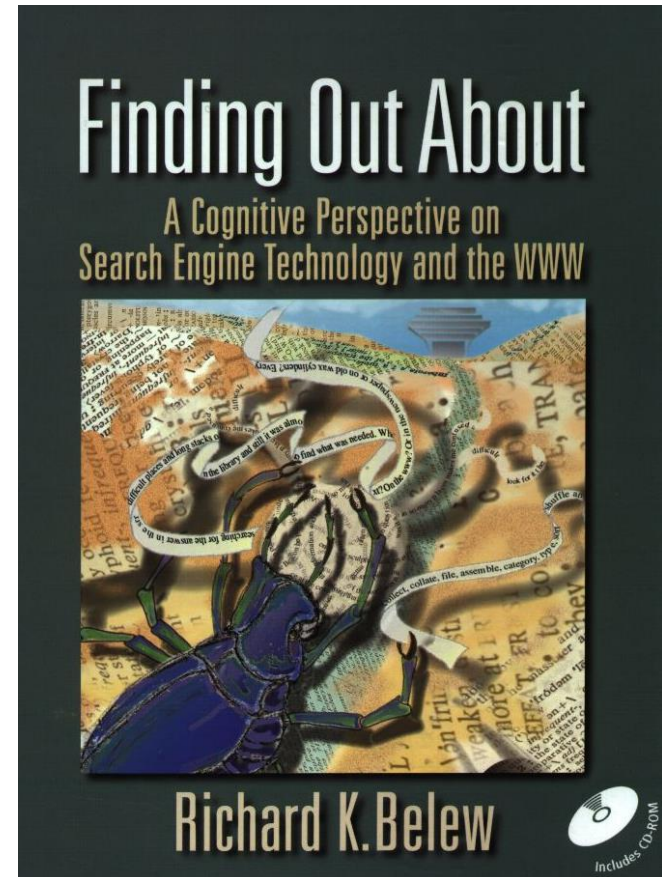
# The problem of “Aboutness”

- What is a text, audio signal, or image **about**?
- This is a problem in **semantics**
- This is exactly the type of problem which:
  - Humans are good at, but
  - Computer programmes are particularly bad at!
- For example – “is this image about dogs?”

# “Aboutness”

information  
retrieval

- Richard K Belew
- *Finding Out About: A Cognitive Perspective on Search Engine Technology and the WWW*
- Cambridge University Press, 2001
- Includes CD-ROM & website



# When is an image “about” dogs



# The problem of “aboutness”

- Intuitively, if we focus on text things should be more straightforward
- But even human interpretation of texts may be ambiguous...
- Simple example:

*I saw the man/on the hill/with the telescope*



# Text Understanding

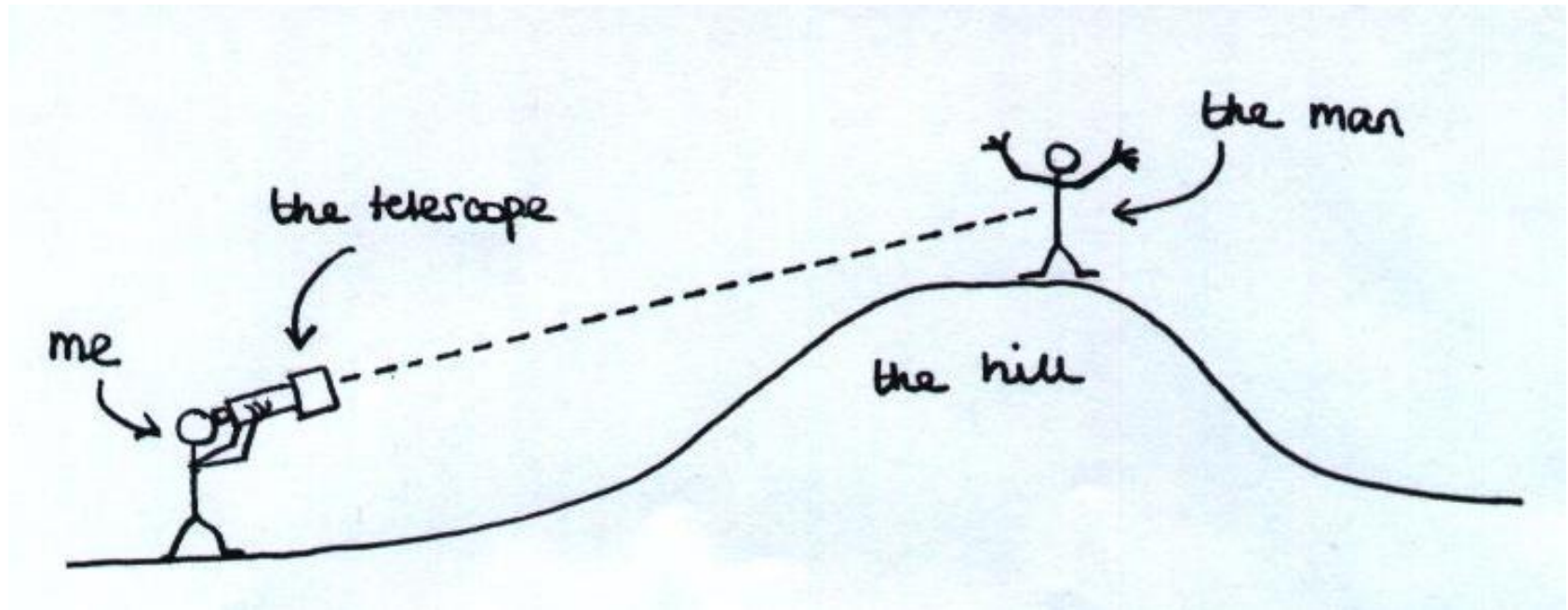
- How can a machine **understand** what this sentence is about?
- Traditionally this involves:
  - Finding the **grammatical** role and meaning of each word
  - Analyse* – **Parsing** the word sequence – applying a set of rules to identify the **structure** of the word sequence relative to a grammar
  - A **grammar** is a **model** that encodes all of the valid word sequences (sentences) in a language



# Text Understanding

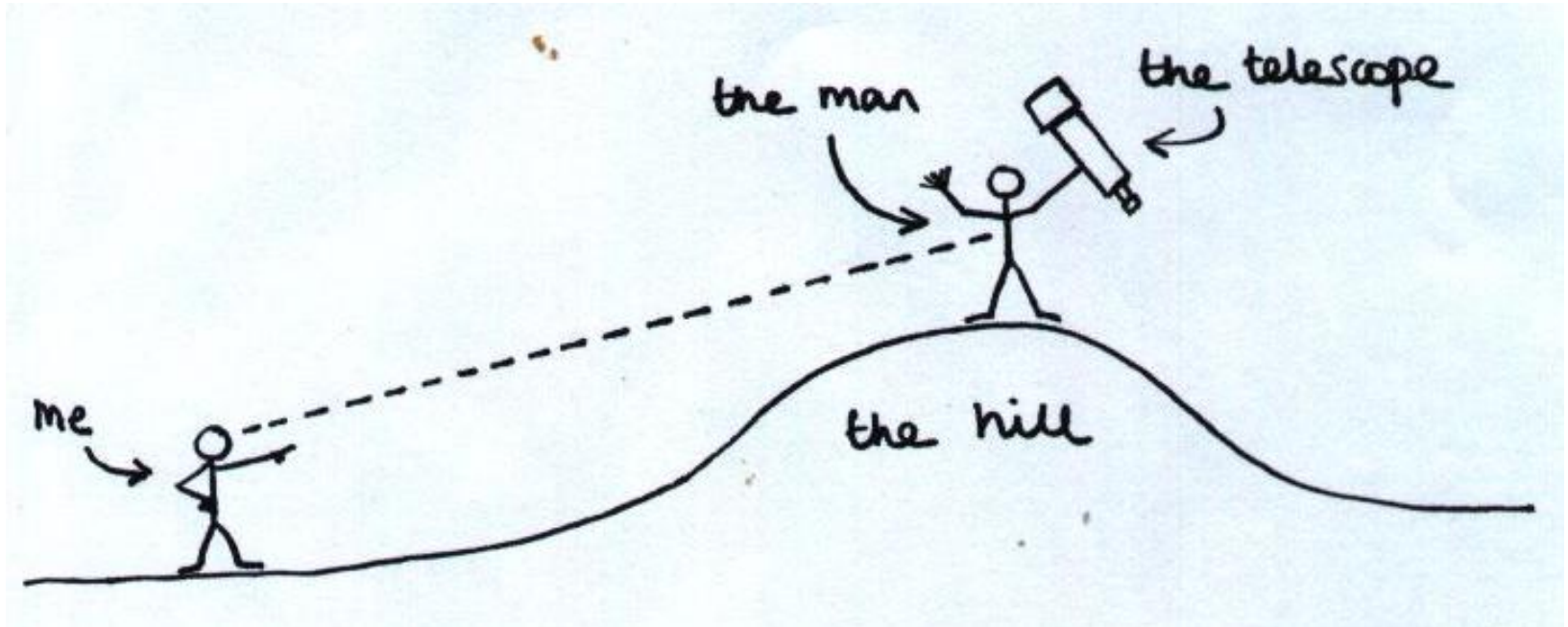
- Words have different meanings and grammatical roles (e.g. “lead” (verb or noun))
- A word sequence may have multiple interpretations relative to the grammar
- A grammatical word sequence may not occur in the given grammar (under generation)
- Conversely, an ungrammatical sentence may be in the grammar (over-generation)

*I saw the man on the hill with the telescope*



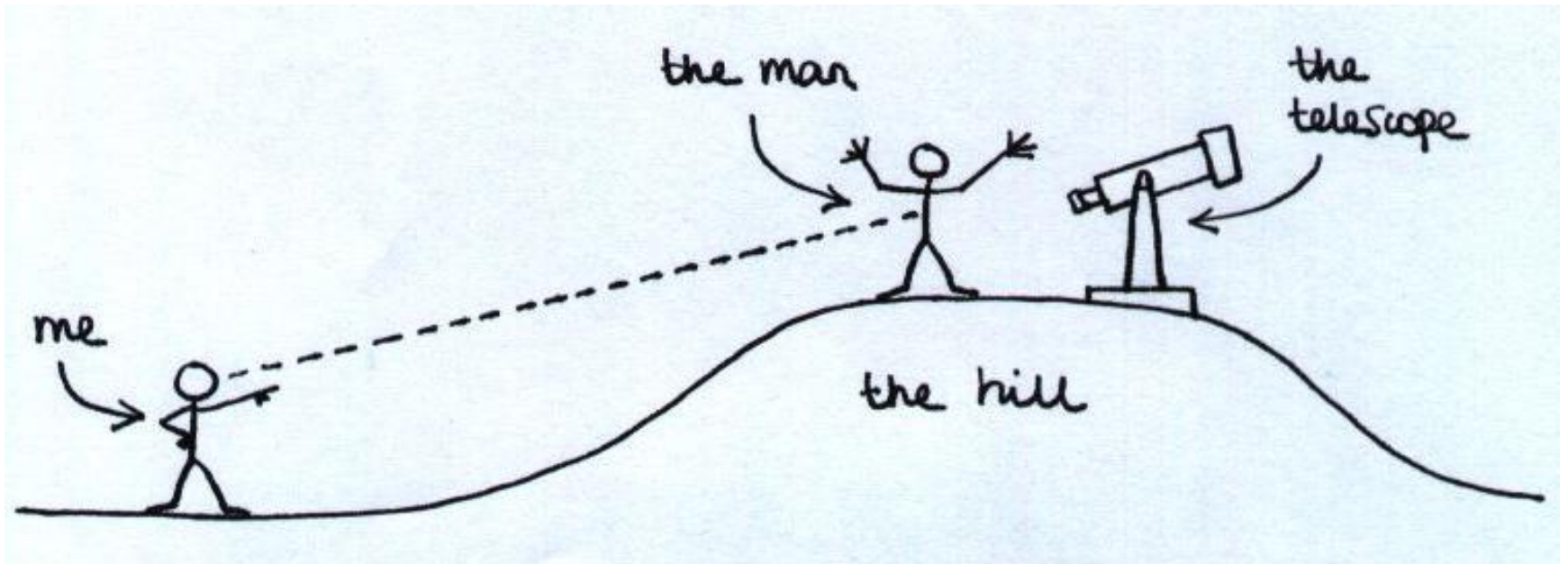
*I saw **the man on the hill** with the telescope*

*I saw the man on the hill with the telescope*



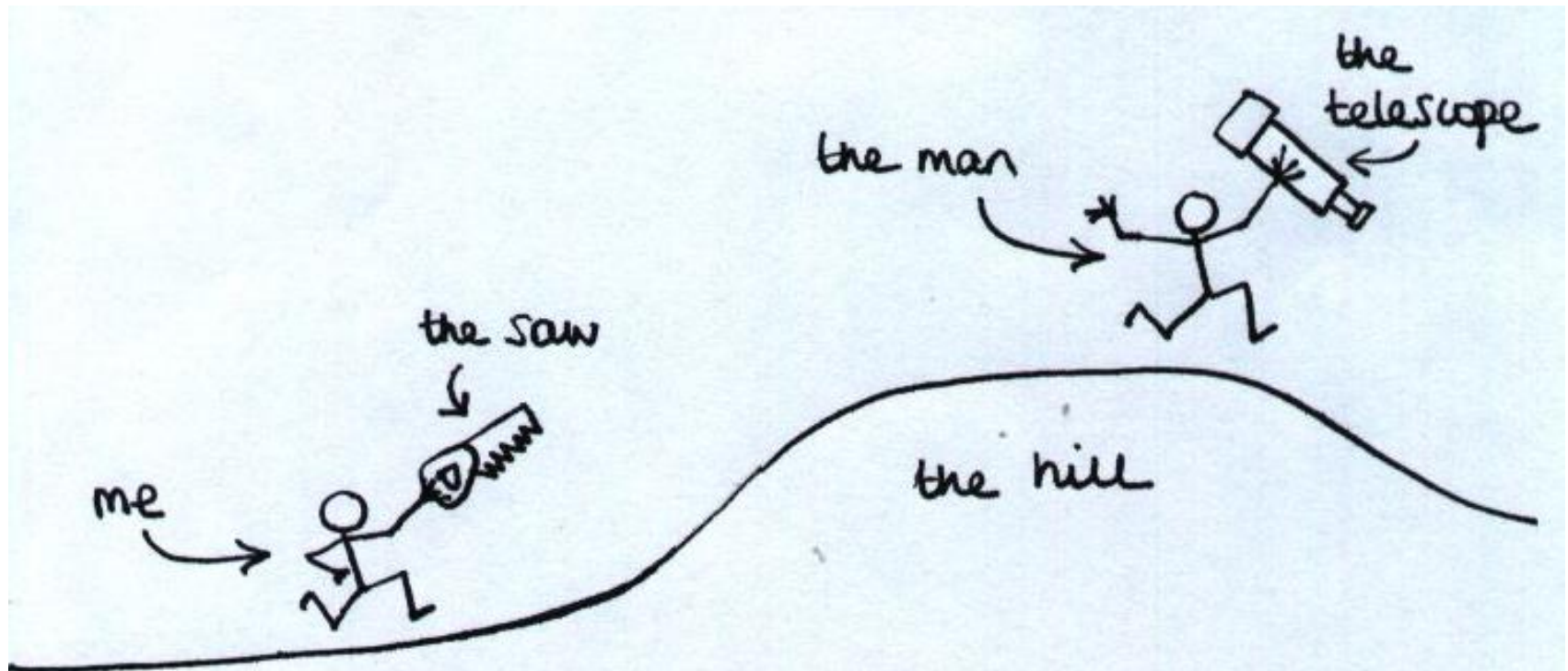
*I saw the man on the hill with the telescope*

*I saw the man on the hill with the telescope*



*I saw the man on the hill with the telescope*

*I saw the man on the hill with the telescope*



*I saw the man on the hill with the telescope*



# Analysis

- Example illustrates two different problems
  - Different grammatical parses of same word sequence

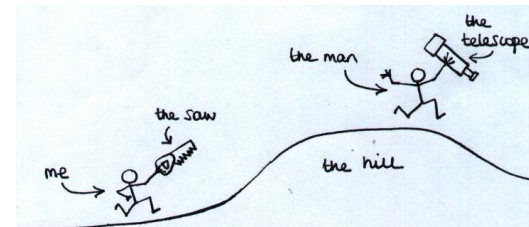
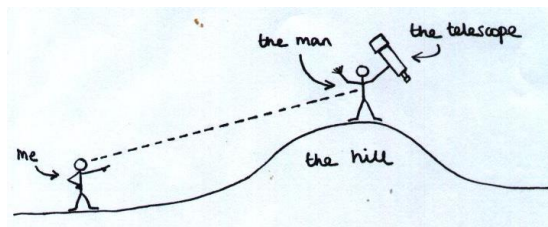
*I saw the man on the hill with the telescope*

vs

*I saw the man on the hill with the telescope*

- Identical parses but different interpretations of words

*I saw the man on the hill with the telescope*



- Move towards **Machine Learning**

# What is Data Mining?

- Mining
  - *Digging deep into the earth, to find hidden, valuable materials*
- Data Mining
  - Analysis of large data corpora: biomedical, acoustic, video, text,... to **discover structure, patterns and relationships**
  - Corpora too large for human inspection
  - Patterns and structure may be hidden

# Related “hot” topics

- “Big Data”
- Pattern recognition/processing
  - As a prerequisite for Data Mining (e.g. ASR for spoken data retrieval)
  - As a consequence of Data Mining
- Machine learning
  - (Deep) Neural Networks, “Deep Learning”
- Data Visualization
  - Dimension reduction



# Some example data

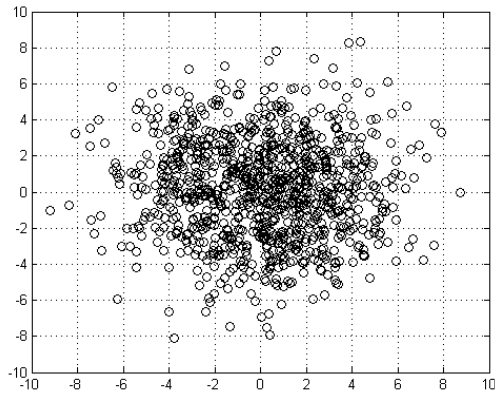


Fig 1: Single,  
spherical  
cluster  
centred at  
origin.

*correlation X*

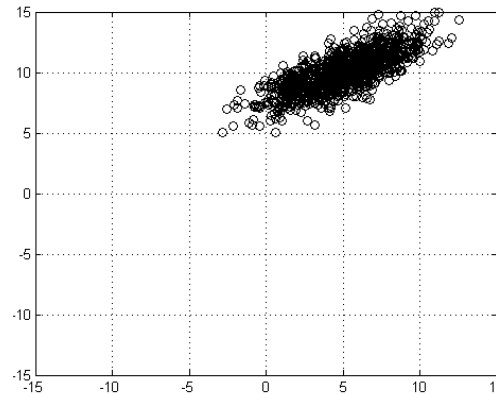


Fig 2: Single,  
arbitrary elliptical  
cluster

*correlation ✓  
diff. variance*

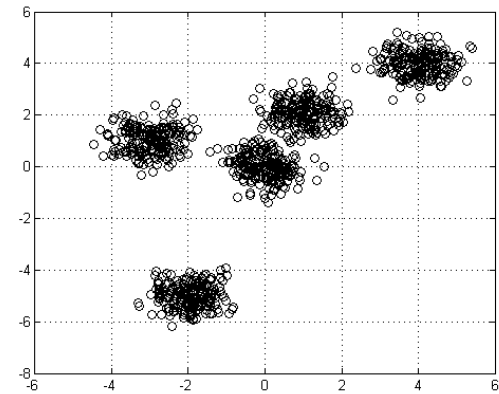


Fig 3: Multiple,  
arbitrary elliptical  
clusters

# What is Information Retrieval (IR)?

- Underlying principles of Search Engine technology
- Finding out About... [Belew]
- Retrieving Information from text sources
- Retrieving Information from other sources
  - Spoken Data Retrieval
  - Bio-informatics
- In IDA we will focus on **text retrieval**

# IR vs Database Retrieval

- IR is not 'database retrieval'
- Databases are characterised by:
  - Strong prior assumptions about
    - Salient properties of data
    - Format
    - Logical relations between data items
    - Likely user queries
  - Formal, restrictive query syntax
  - Need for dedicated maintenance to keep it up-to-date
  - Gives specific replies to specific queries



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<a href="#">London</a>	<b>£52</b>
<a href="#">New York</a>	<b>£49</b>
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<a href="#">Florida</a>	<b>£17</b>

# IR vs Database Retrieval

- IR (Finding Out About)
  - No prior assumptions about:
    - Salient properties of data
    - Format of data
    - Logical relations between data items
  - Less specific ‘natural language’ queries
  - Source information remains up-to-date
  - Much less focussed replies

# Relevant topics in mathematics

- Vectors and matrices
  - Data that we analyse is generally vector data
  - A single data point may comprise multiple measurements
  - Words or documents typically represented as vectors
  - A basic understanding of the mathematics of vectors (linear algebra) is crucial for intelligent data analysis and text retrieval
- Probability
- Next lecture – linear algebra revision

# Summary

- Introduction to course components
  - Background mathematics
  - Data visualization and data mining
  - Information retrieval
- Motivation
  - Availability of huge corpora of raw data
- Problems
  - Aboutness