

# Data Mining: Introduction

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Introduction to Data Mining, 2<sup>nd</sup> Edition  
Chapter I

# What is Data Mining?

It is the use of **efficient** techniques for the analysis of **very large collections of data** and the **extraction** of useful and possibly unexpected patterns in data (**hidden knowledge**).

# Large-scale Data is Everywhere!

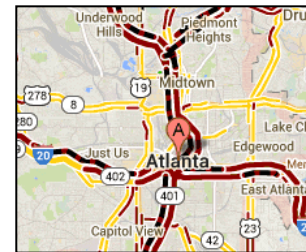
- **Enormous data growth in both commercial and scientific databases**
  - due to advances in data generation and collection technologies
- **New mantra**
  - Gather whatever data you can whenever and wherever possible
- **Expectations**
  - Gathered data will have value either for the purpose collected or for a purpose not envisioned.



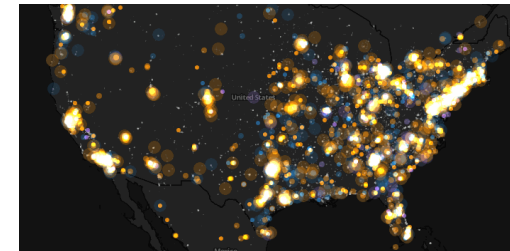
*Cyber Security*



*E-Commerce*



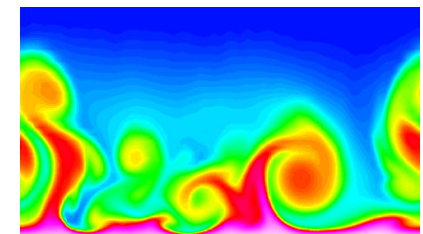
*Traffic Patterns*



*Social Networking: Twitter*



*Sensor Networks*



*Computational Simulations*

# Why Data Mining? Commercial Viewpoint

- **Lots of data is being collected and warehoused**

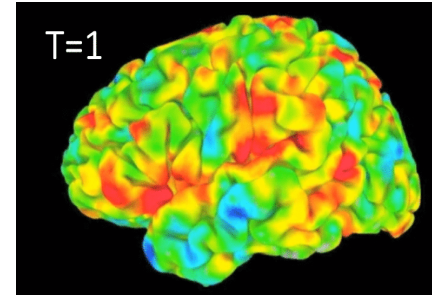
- Web data
  - Yahoo has Peta Bytes of web data
  - Facebook has billions of active users
- purchases at department/grocery stores, e-commerce
  - Amazon handles millions of visits/day
- Bank/Credit Card transactions



- **Computers have become cheaper and more powerful**
- **Competitive Pressure is Strong**
  - Provide better, customized services for an edge (e.g. in Customer Relationship Management)

# Why Data Mining? Scientific Viewpoint

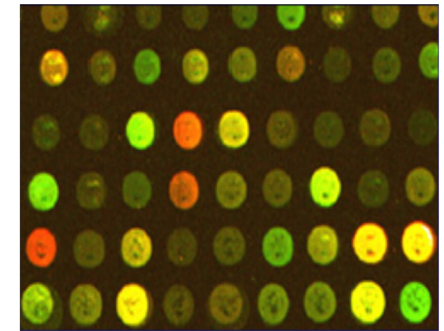
- **Data collected and stored at enormous speeds**
  - remote sensors on a satellite
    - NASA EOSDIS archives over petabytes of earth science data / year
  - telescopes scanning the skies
    - Sky survey data
  - High-throughput biological data
  - scientific simulations
    - terabytes of data generated in a few hours
- **Data mining helps scientists**
  - in automated analysis of massive datasets
  - In hypothesis formation



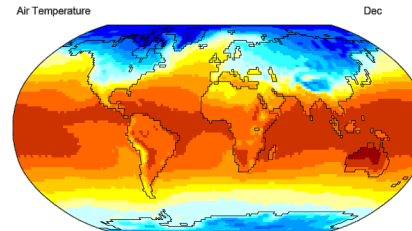
fMRI Data from Brain



Sky Survey Data



Gene Expression Data



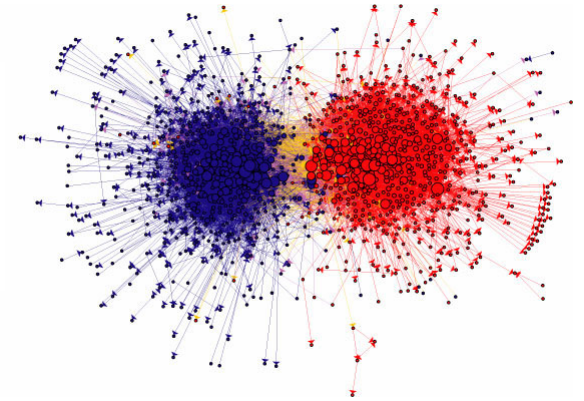
Surface Temperature of Earth

# Big data proxies of social life

Shopping patterns & lifestyle



RELATIONSHIPS & SOCIAL TIES

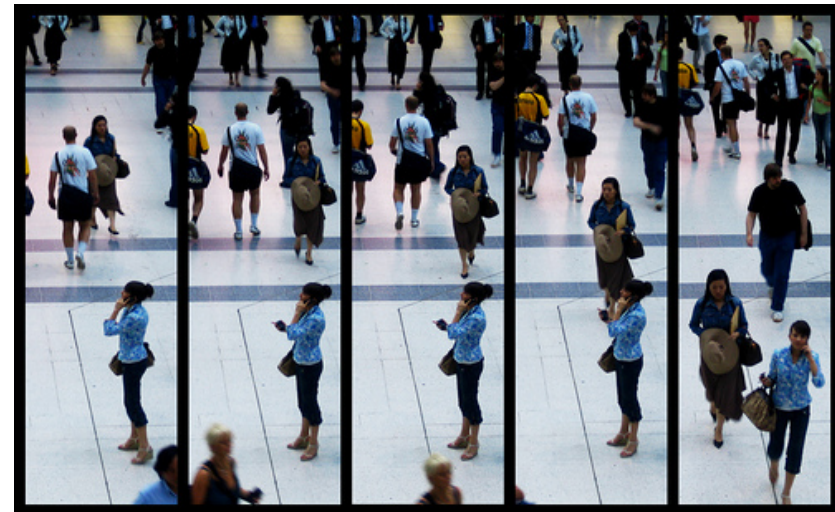


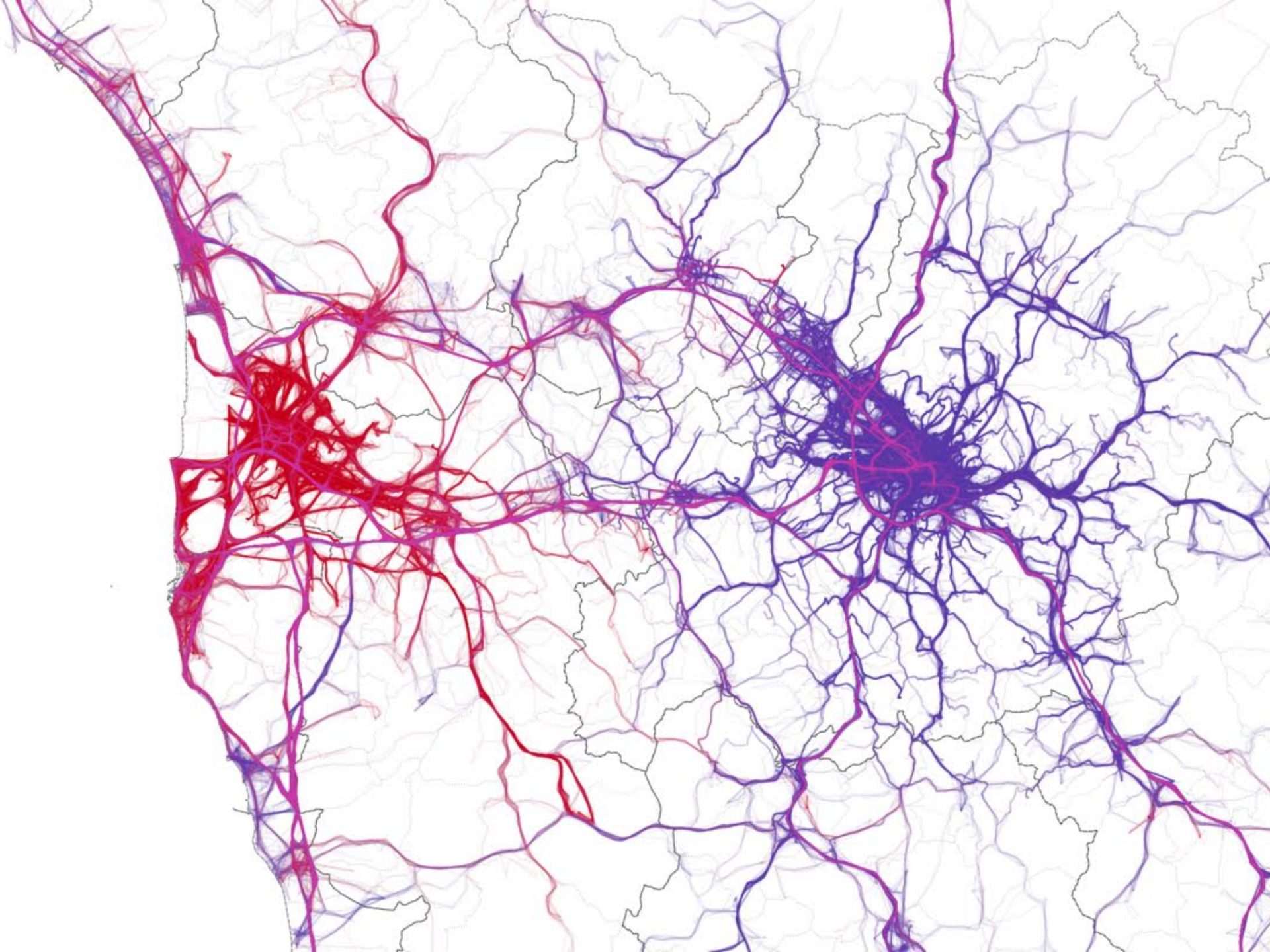
DESIRES, OPINIONS, SENTIMENTS

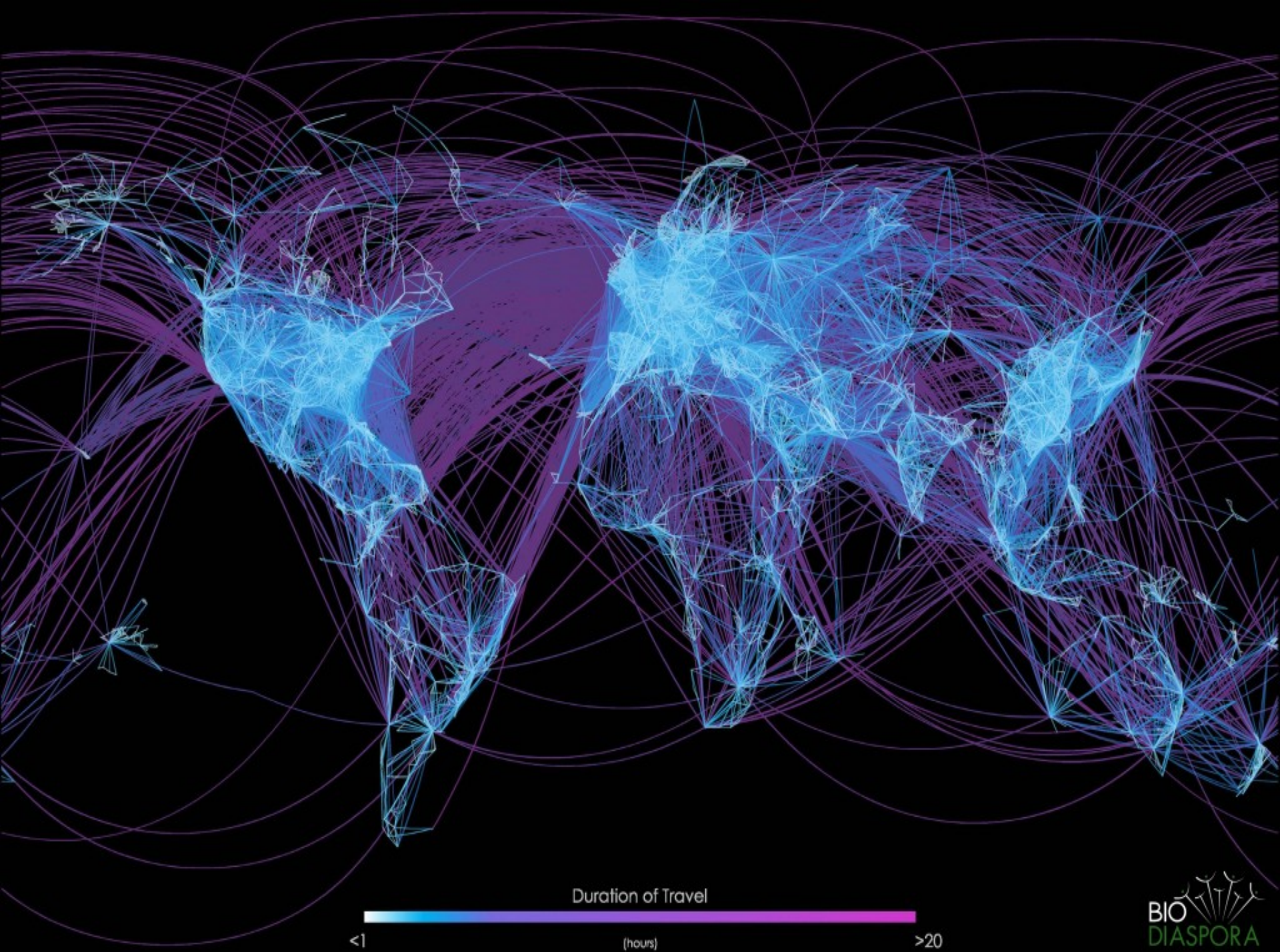


WIKIPEDIA  
*The Free Encyclopedia*

MOVEMENTS





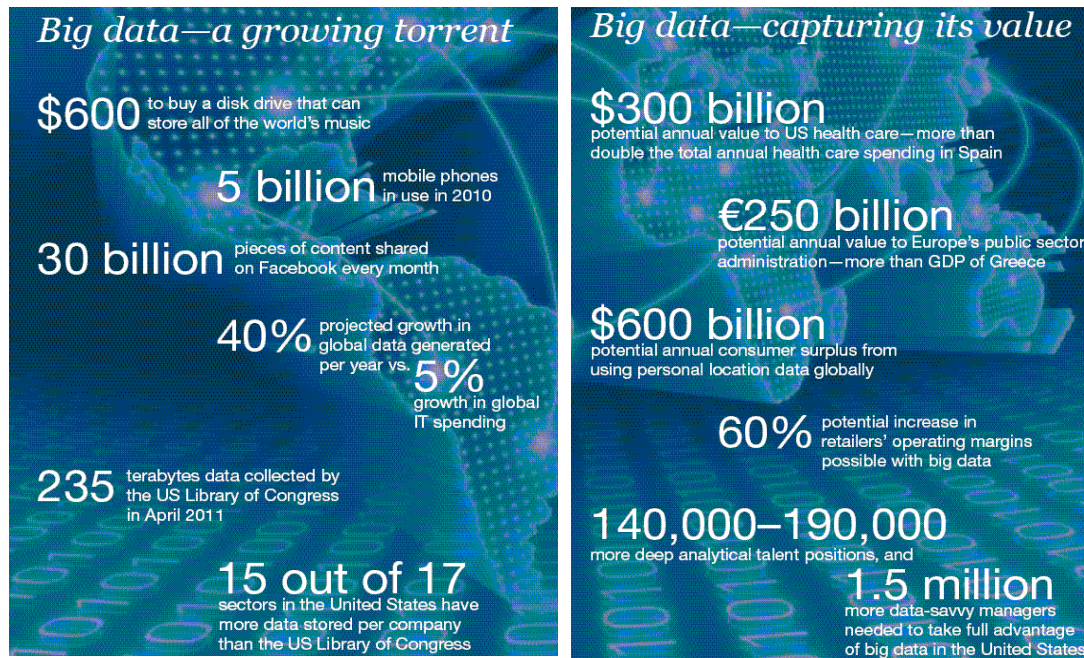




# Great opportunities to improve productivity in all walks of life

McKinsey Global Institute

## Big data: The next frontier for innovation, competition, and productivity

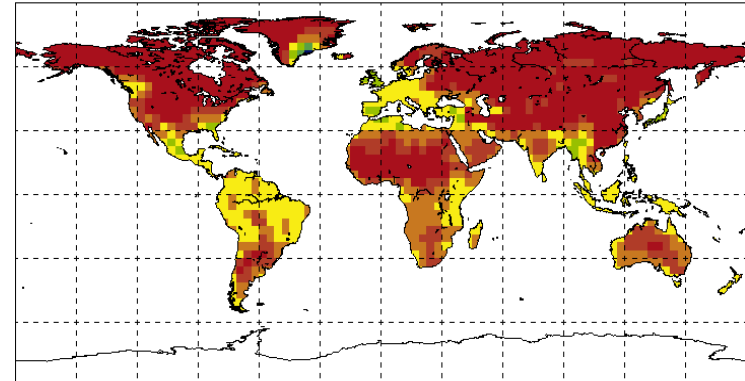


# Great Opportunities to Solve Society's Major Problems



**Improving health care and reducing costs**

CCCma/A2a January to January Mean Temperature (degrees C) 2080s relative to 1961-90



**Predicting the impact of climate change**



**Finding alternative/ green energy sources**

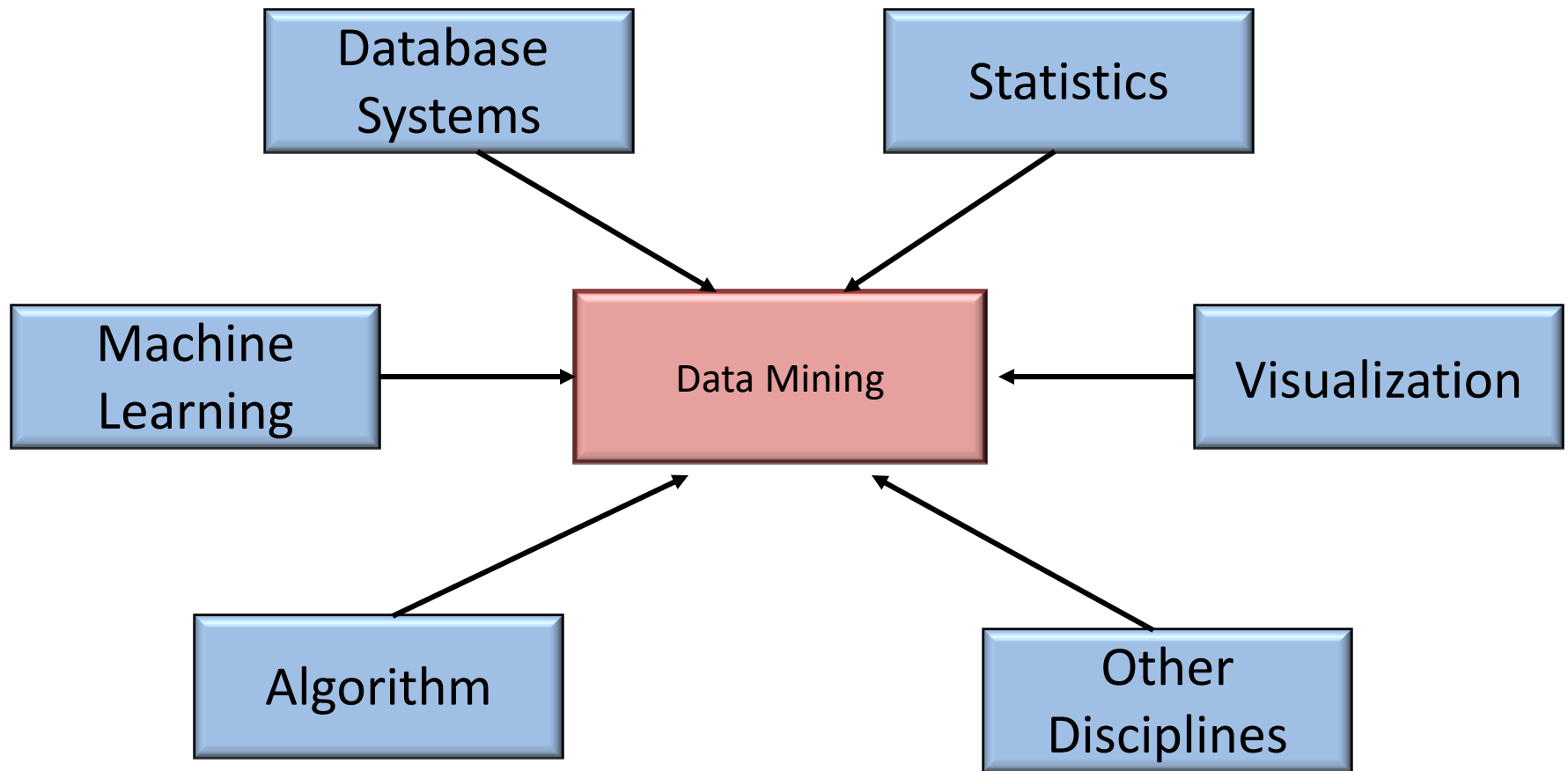


**Reducing hunger and poverty by increasing agriculture production**

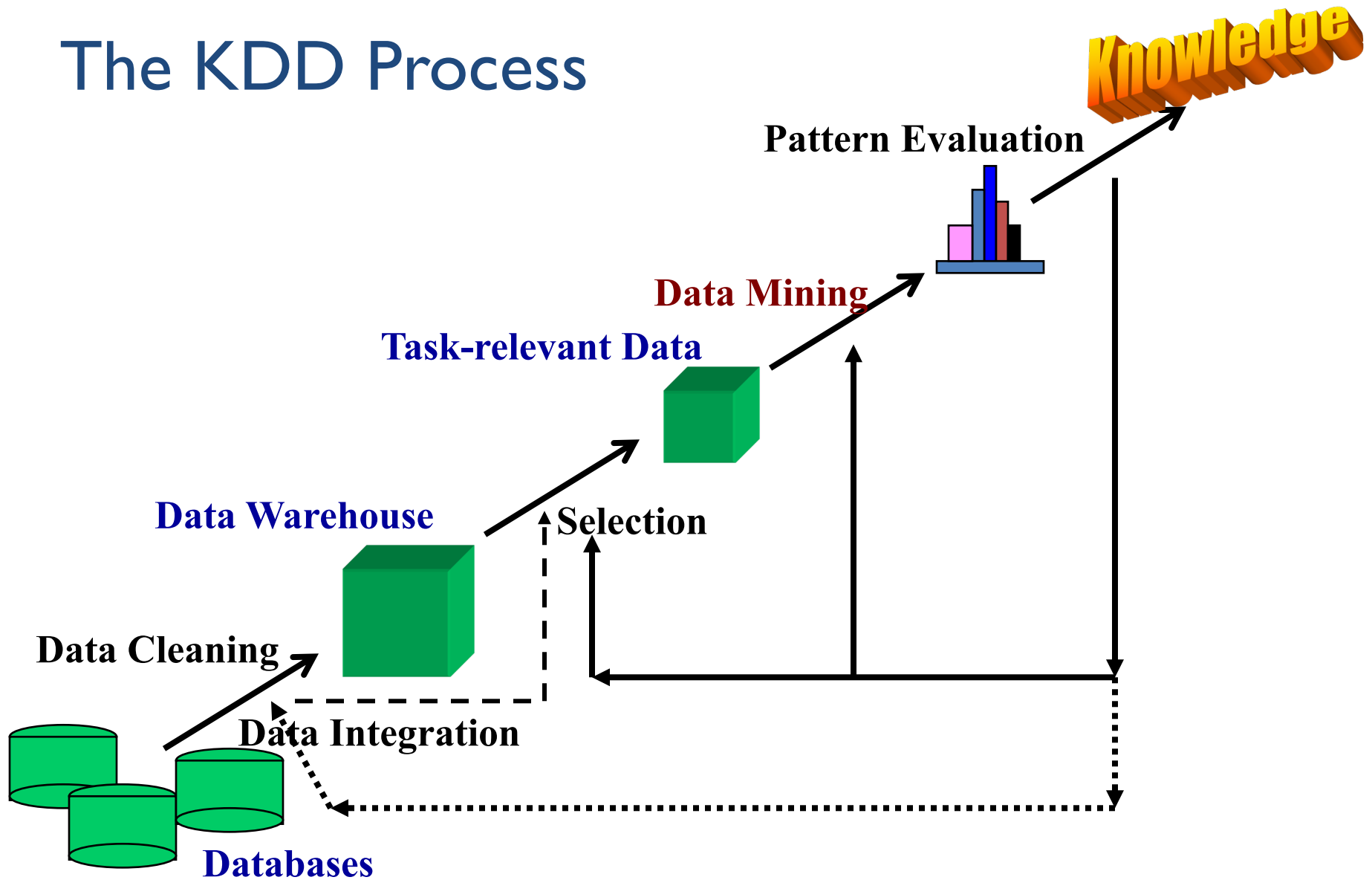
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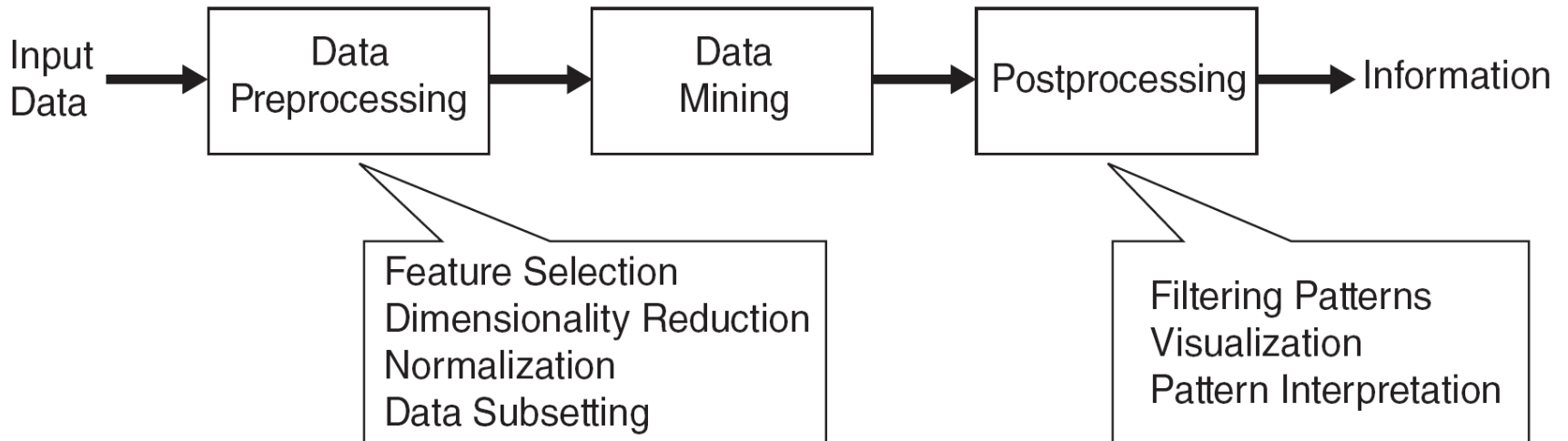
# Data Mining: Confluence of Multiple Disciplines



# The KDD Process



# What is Data Mining?



# Primary & Secondary Data

## Primary Data

- **Original data** that has been collected for a specific purpose
- Primary data is **not altered by humans**

## Secondary Data

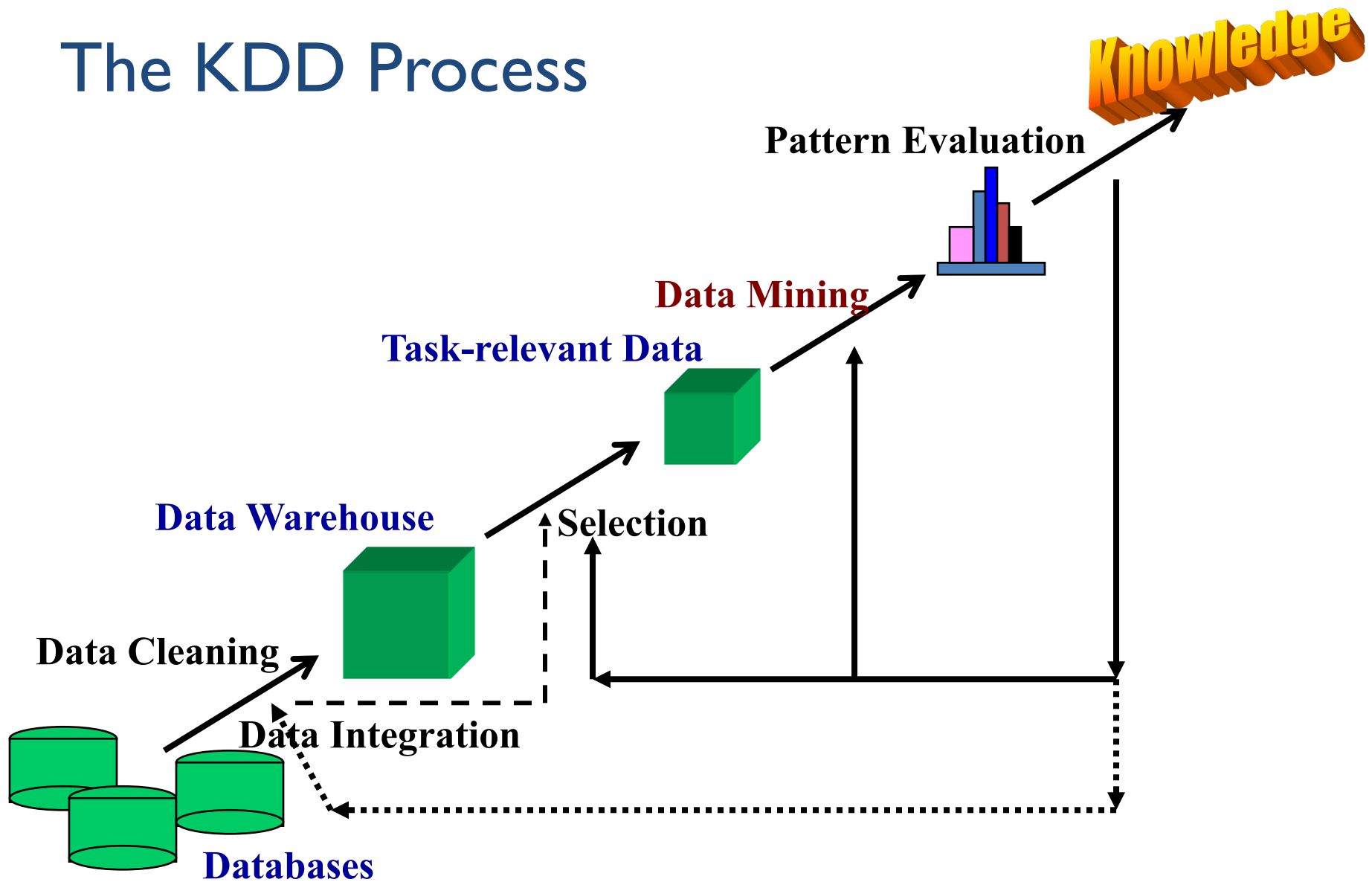
- **Data that** has been already collected and **made available for other purposes**
- Secondary data may be obtained **from many sources**



# Variety of Data Sources

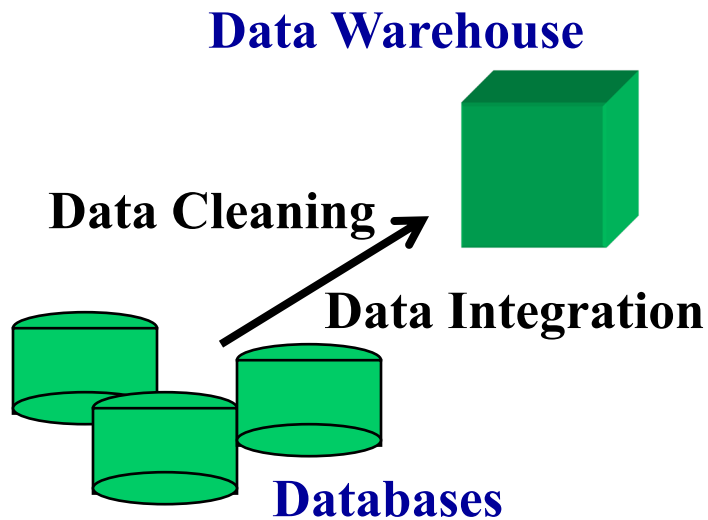


# The KDD Process





# Data Integration and Preparation

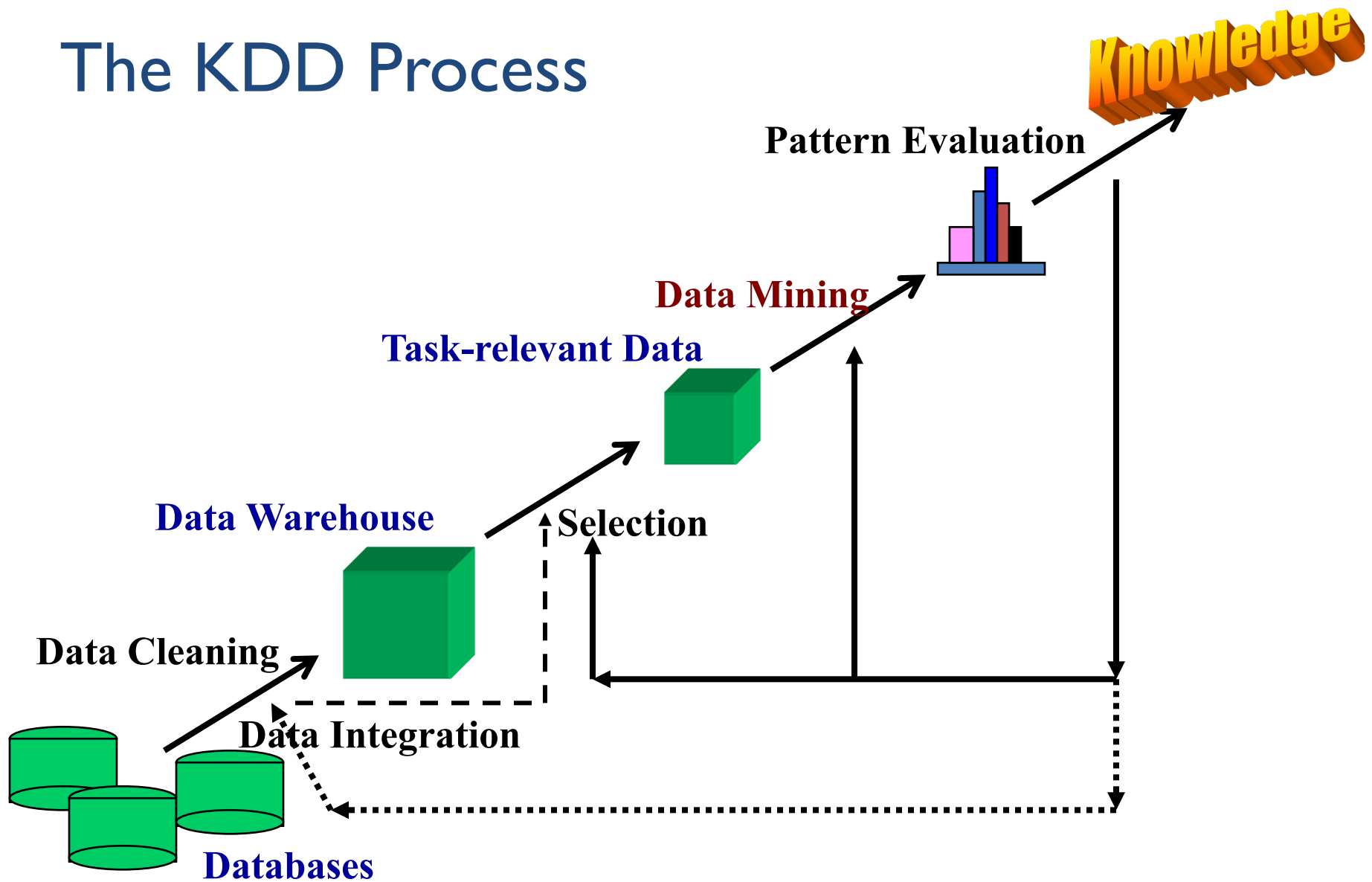


**Data Integration** involves the process of data understanding, data cleaning, merging data coming from multiple sources and transforming them to load them into a **Data Warehouse**

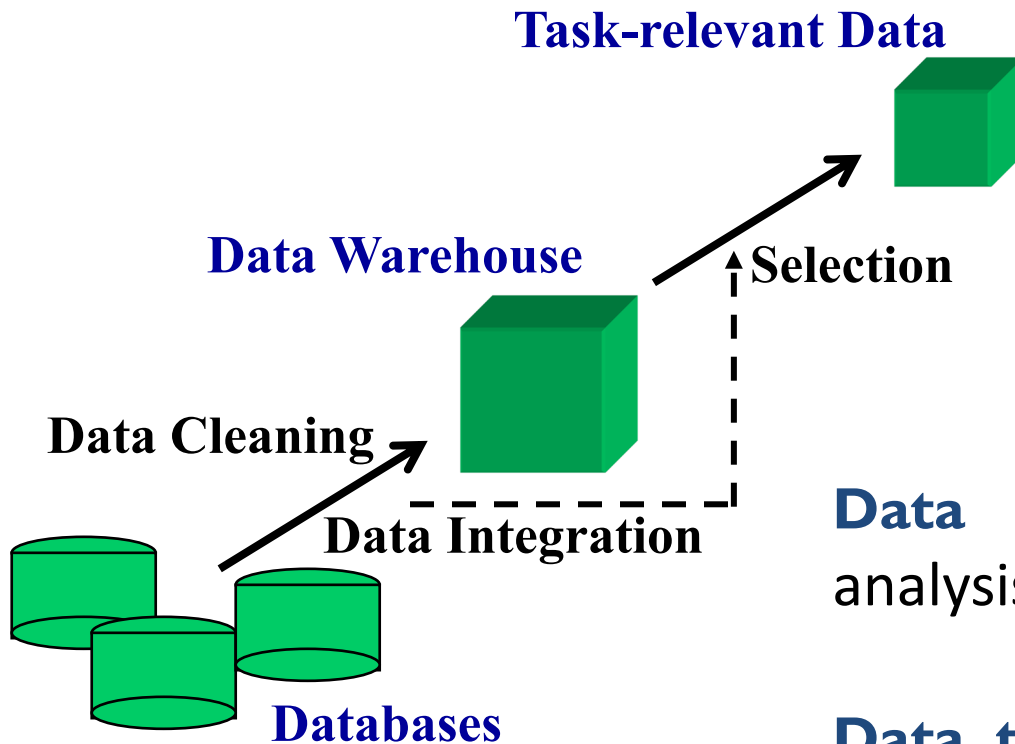
**Data Warehouse** is a database targeted to answer **specific business questions**

Developing a data analytics project requires the  
**BUSINESS UNDERSTANDING**

# The KDD Process



# Data Selection and Transformation



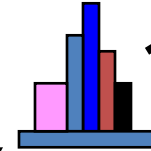
**Data Selection:** Data relevant to analysis tasks are retrieved from data

**Data transformation:** Transform data into appropriate form for mining (summary, aggregation, etc.)

# The KDD Process

**Knowledge**

**Pattern Evaluation**



**Data Mining**

**Task-relevant Data**



**Data Warehouse**

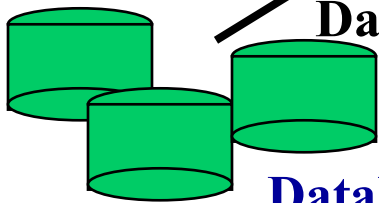
**Selection**



**Data Cleaning**

**Data Integration**

**Databases**



**Pattern Evaluation:** Identify truly interesting patterns

**Knowledge representation:** Use visualization and knowledge representation tools to present the mined data to the user

# Data Mining Tasks

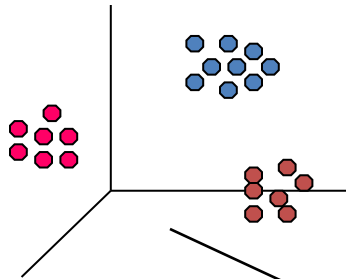
- Prediction Methods
  - Use some variables to predict unknown or future values of other variables.
- Description Methods
  - Find human-interpretable patterns that describe the data.

# Data Mining Tasks

- Prediction Methods
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From [Fayyad, et.al.] Advances in Knowledge Discovery and Data Mining, 1996

# Data Mining Tasks ...



Clustering

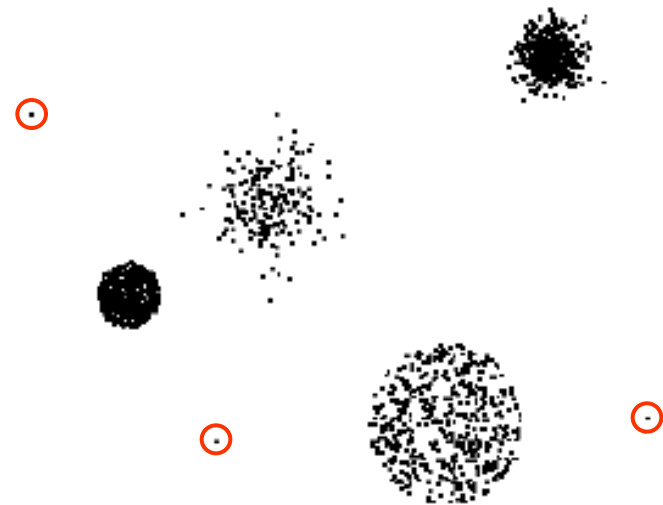
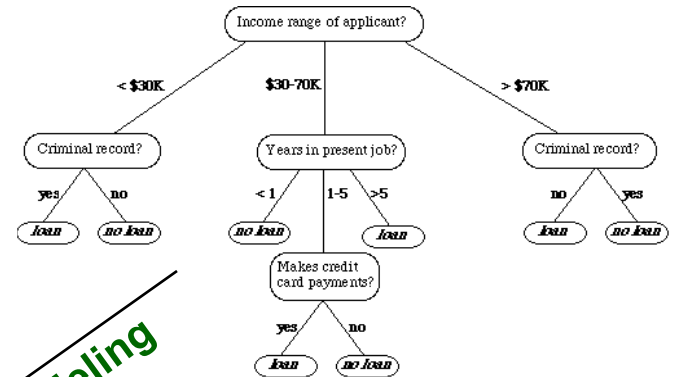
## Data

Tid	Refund	Marital Status	Taxable Income	Cheat
1	Yes	Single	125K	No
2	No	Married	100K	No
3	No	Single	70K	No
4	Yes	Married	120K	No
5	No	Divorced	95K	Yes
6	No	Married	60K	No
7	Yes	Divorced	220K	No
8	No	Single	85K	Yes
9	No	Married	75K	No
10	No	Single	90K	Yes
11	No	Married	60K	No
12	Yes	Divorced	220K	No
13	No	Single	85K	Yes
14	No	Married	75K	No
15	No	Single	90K	Yes

Association Rules

Predictive Modeling

Anomaly Detection



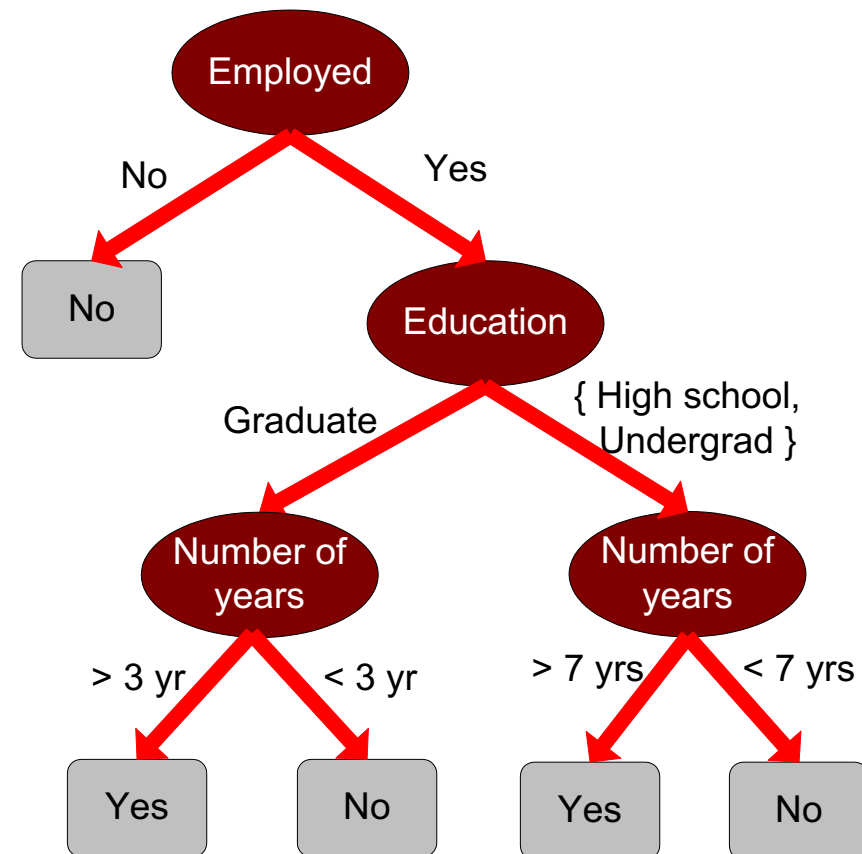
# Predictive Modeling: Classification

- Find a model for class attribute as a function of the values of other attributes

Class

<i>Tid</i>	Employed	Level of Education	# years at present address	Credit Worthy
1	Yes	Graduate	5	Yes
2	Yes	High School	2	No
3	No	Undergrad	1	No
4	Yes	High School	10	Yes
...	...	...	...	...

Model for predicting credit worthiness



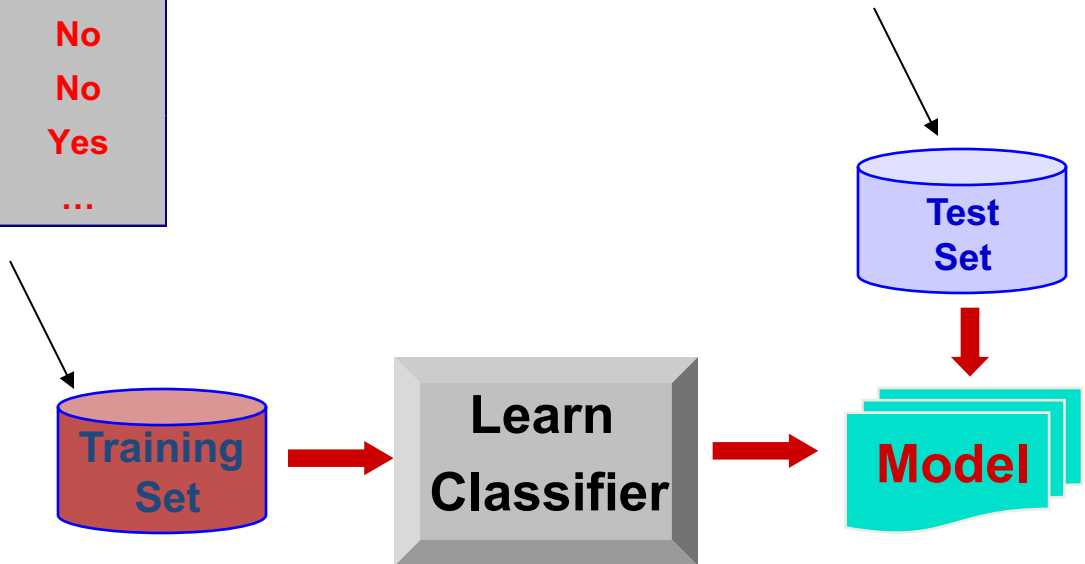


# Classification Example

categorical      categorical      quantitative      class

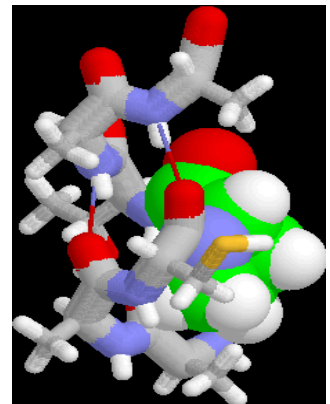
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1	Yes	Graduate	5	Yes
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3	No	Undergrad	1	No
4	Yes	High School	10	Yes
...	...	...	...	...

<i>Tid</i>	Employed	Level of Education	# years at present address	Credit Worthy
1	Yes	Undergrad	7	?
2	No	Graduate	3	?
3	Yes	High School	2	?
...	...	...	...	...

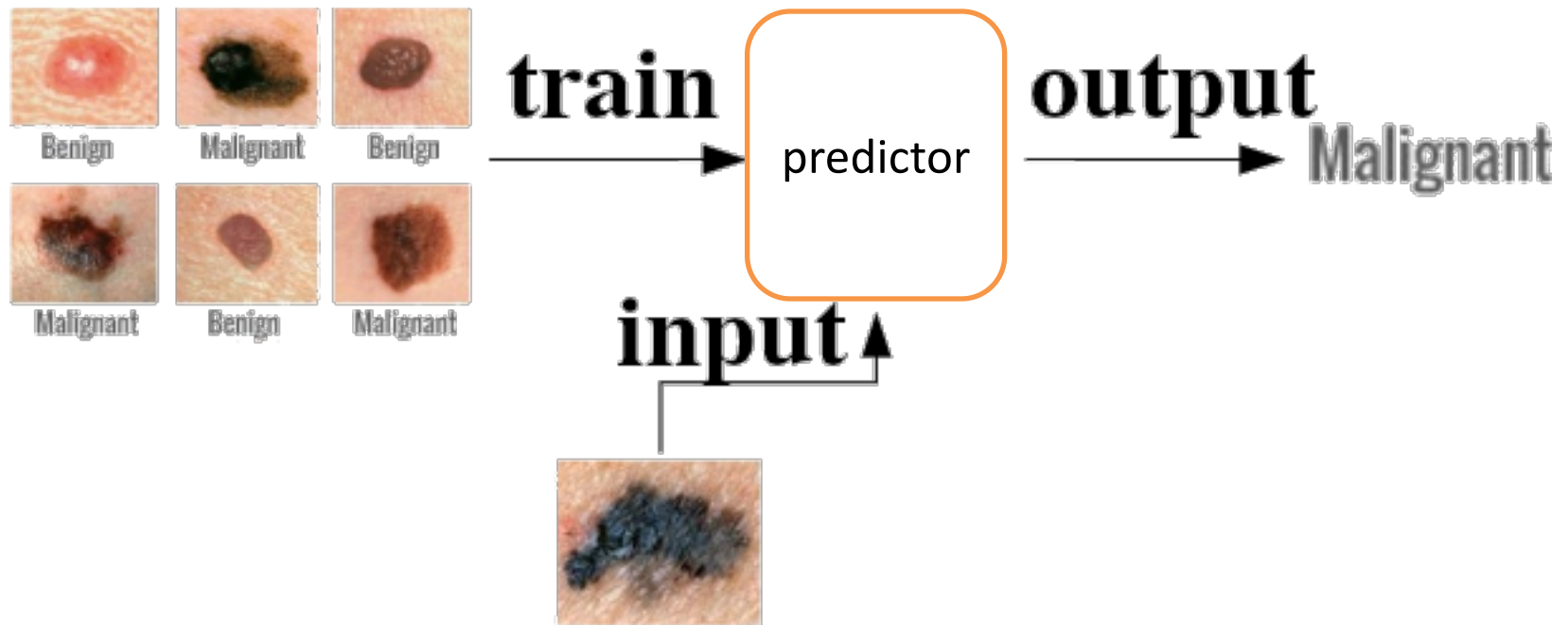


# Examples of Classification Task

- Classifying credit card transactions as legitimate or fraudulent
- Classifying land covers (water bodies, urban areas, forests, etc.) using satellite data
- Categorizing news stories as finance, weather, entertainment, sports, etc
- Identifying intruders in the cyberspace
- Predicting tumor cells as benign or malignant
- Classifying secondary structures of protein as alpha-helix, beta-sheet, or random coil



# AI = Machine Learning + Big Data



# Classification: Application I

## Fraud Detection

- **Goal:** Predict fraudulent cases in credit card transactions.
- **Approach:**
  - Use credit card transactions and the information on its account-holder as attributes.
    - When does a customer buy, what does he buy, how often he pays on time, etc
  - Label past transactions as fraud or fair transactions. This forms the class attribute.
  - Learn a model for the class of the transactions.
  - Use this model to detect fraud by observing credit card transactions on an account.

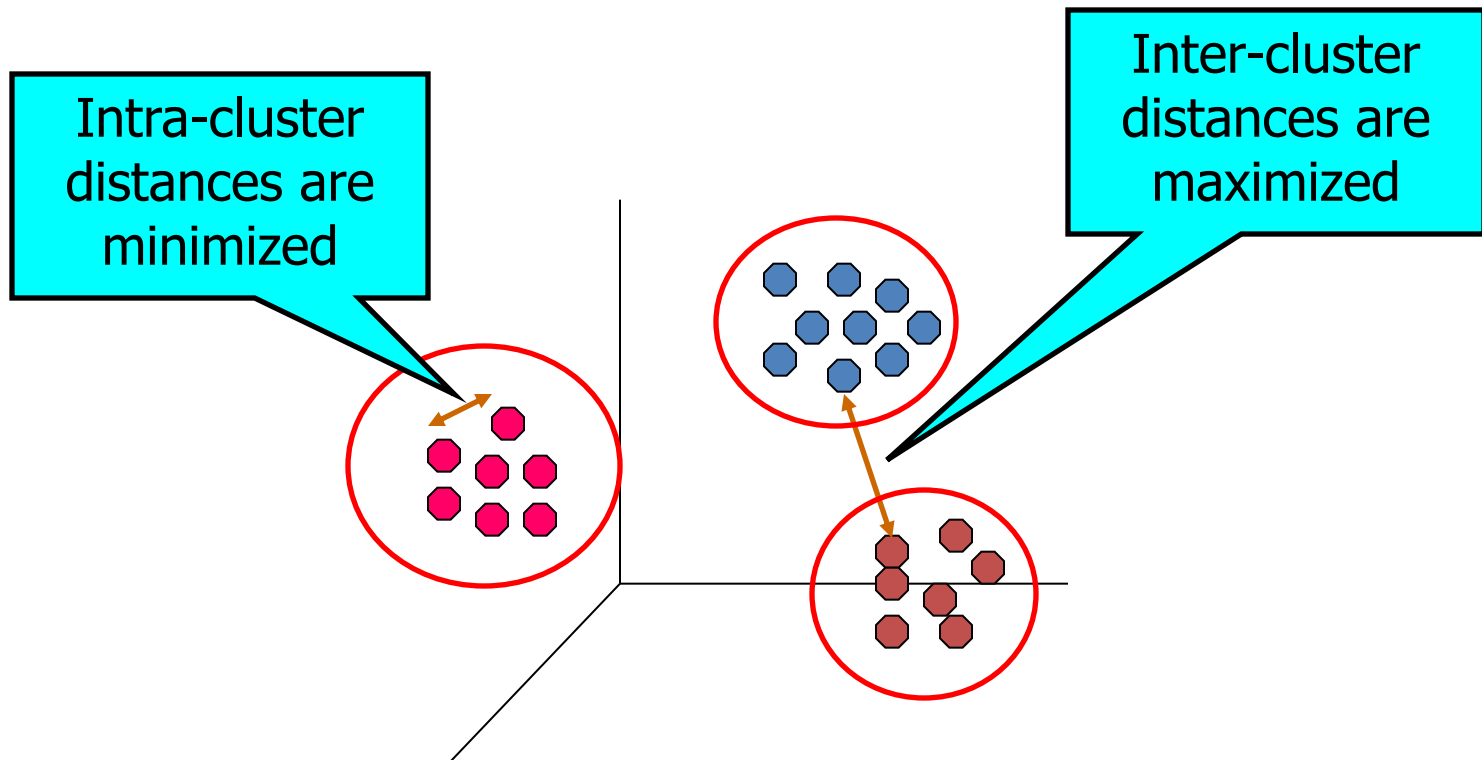
# Classification: Application 2

## Churn prediction for telephone customers

- **Goal:** To predict whether a customer is likely to be lost to a competitor.
- **Approach:**
  - Use detailed record of transactions with each of the past and present customers, to find attributes.
    - How often the customer calls, where he calls, what time-of-the day he calls most, his financial status, marital status, etc.
  - Label the customers as loyal or disloyal.
  - Find a model for loyalty.

# Clustering

- Finding groups of objects such that the objects in a group will be similar (or related) to one another and different from (or unrelated to) the objects in other groups



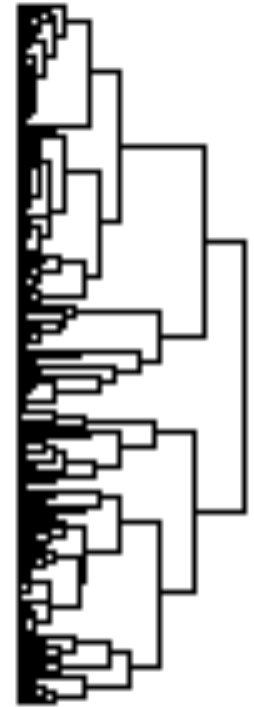
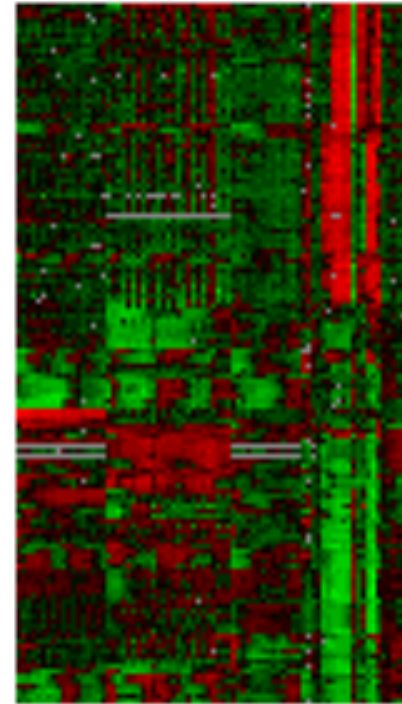
# Applications of Cluster Analysis

- **Understanding**

- Custom profiling for targeted marketing
- Group related documents for browsing
- Group genes and proteins that have similar functionality
- Group stocks with similar price fluctuations

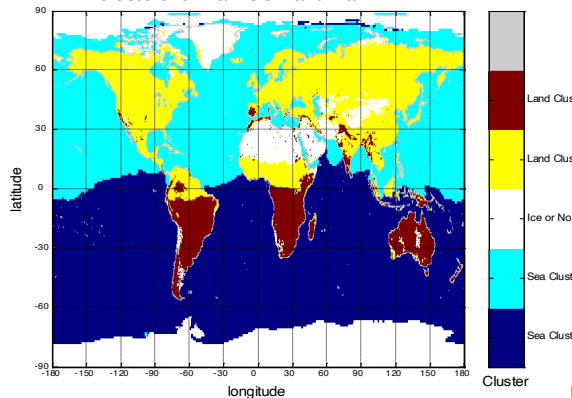
- **Summarization**

- Reduce the size of large data sets



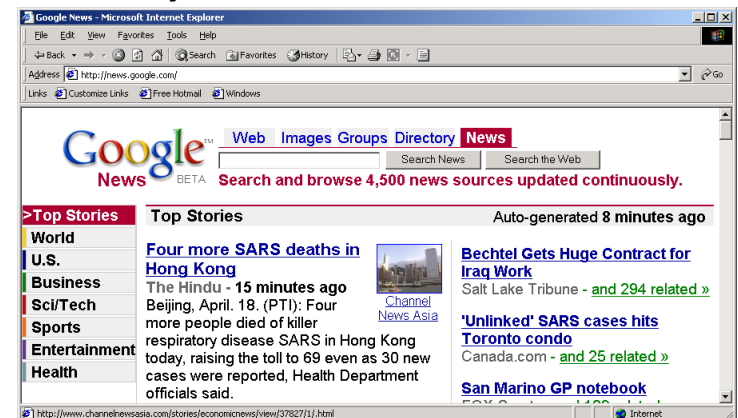
Courtesy: Michael Eisen

Clusters for Raw SST and Raw NPP



**Use of K-means to partition Sea Surface Temperature (SST) and Net Primary Production (NPP) into clusters that reflect the Northern and Southern Hemispheres.**

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# Clustering: Application I

## Market Segmentation:

- **Goal:** subdivide a market into distinct subsets of customers where any subset may conceivably be selected as a market target to be reached with a distinct marketing mix.
- **Approach:**
  - Collect different attributes of customers based on their geographical and lifestyle related information.
  - Find clusters of **similar customers**.
  - Measure the clustering quality by observing buying patterns of customers in same cluster vs. those from different clusters.



# A Behavior Based Segmentation

Using unsupervised clustering segmentation for a grocery chain which would like better product assortment for its high profitable customers

## Potential Inputs

### Value

- Basket Size
- Visit Frequency

### Basket

- Spend by category
- Type of category
- Brand spend (i.e. private label)

### Promotions

- % bought on targeted promotion
- % bought from flyer

### Time

- Time of day
- Day of week

### Location

- Store format
- Area population density

Clustering  
approach



## Deal Seeking Mom

### Key Differentiators



- Full store shop
- High avg. basket size / # trips



- High % purchased on promotion
- Rewards seeker



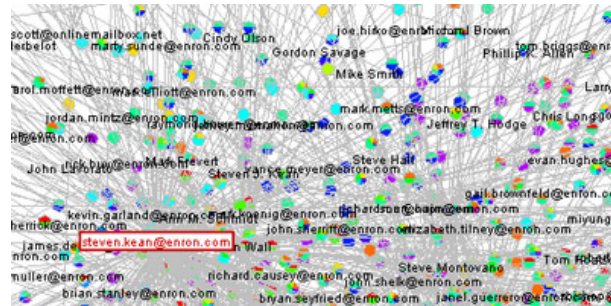
- High spend categories
  - Fresh produce
  - Organic food
  - Multipack juice, snack

# Clustering: Application 2

## Document Clustering:

- **Goal:** To find groups of documents that are similar to each other based on the important terms appearing in them.
- **Approach:** To identify frequently occurring terms in each document. Form a similarity measure based on the frequencies of different terms. Use it to cluster.

Enron email dataset



# Association Rule Discovery: Definition

- Given a set of records each of which contain some number of items from a given collection
  - Produce dependency rules which will predict occurrence of an item based on occurrences of other items.

<i>TID</i>	<i>Items</i>
1	Bread, Coke, Milk
2	Beer, Bread
3	Beer, Coke, Diaper, Milk
4	Beer, Bread, Diaper, Milk
5	Coke, Diaper, Milk

Rules Discovered:

{Milk} --> {Coke}

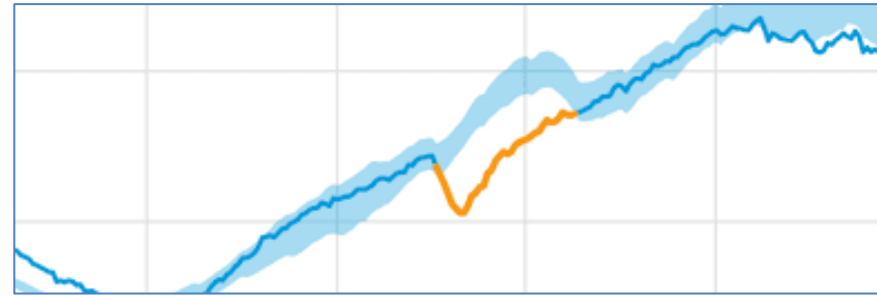
{Diaper, Milk} --> {Beer}

# Association Analysis: Applications

- **Market-basket analysis**
  - Rules are used for sales promotion, shelf management, and inventory management
- **Telecommunication alarm diagnosis**
  - Rules are used to find combination of alarms that occur together frequently in the same time period
- **Medical Informatics**
  - Rules are used to find combination of patient symptoms and test results associated with certain diseases

# Deviation/Anomaly/Change Detection

- Detect significant deviations from normal behavior
- Applications:
  - Credit Card Fraud Detection
  - Network Intrusion Detection
  - Identify anomalous behavior from sensor networks for monitoring and surveillance.
  - Detecting changes in the global forest cover.



# Motivating Challenges

Traditional techniques may be unsuitable due to some challenges:

- Scalability
- High Dimensionality
- Heterogeneous and Complex Data
- Data Ownership and Distribution
- Non-traditional Analysis