# Data Mining and Cyber Threat Analysis – Five Trends

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# Trend 1. Alert Management Systems

Focus in deployment is shifting from models to alerts, from data mining systems to alert management systems.

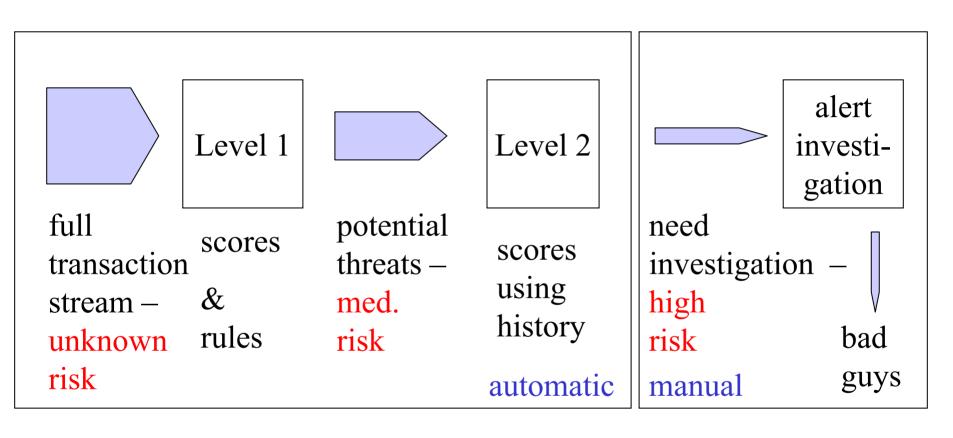
# What is an Alert Management System?

- ☐ An Alert Management System (AMS) is a real time system which maintains *profiles* about individuals, threats, or other entities and in real time processes *events* and returns alerts about profiles and their risks.
- ☐ Examples: credit card fraud detection, threat assessment systems, intrusion detection systems, homeland defense, etc.

# What are the Five Critical Functions of an AMS?

- 1. Scoring compute risk scores for transactions, profiles, targets, etc.
- 2. Linking social network analysis of targets
- 3. Matching against watch lists, e.g. OFAC
- 4. Checking regulations & policies
- 5. Routing analysts have finite capacity

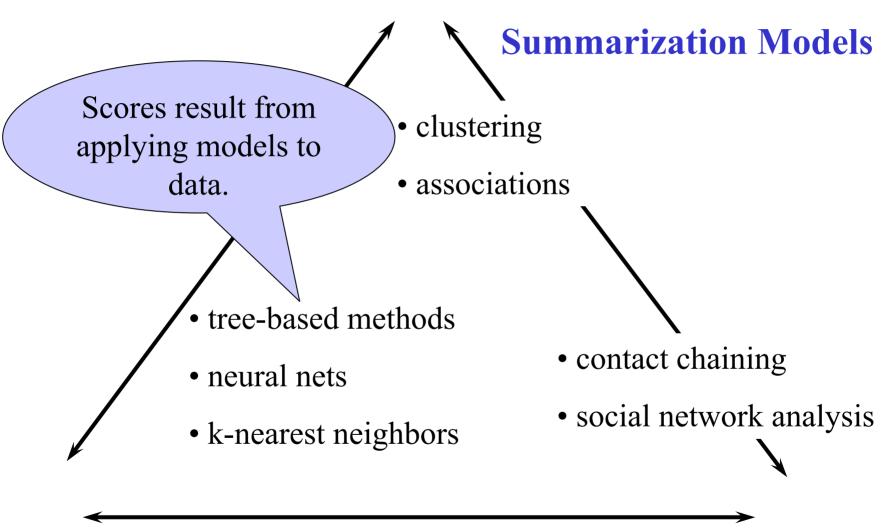
### Alert Data Flow



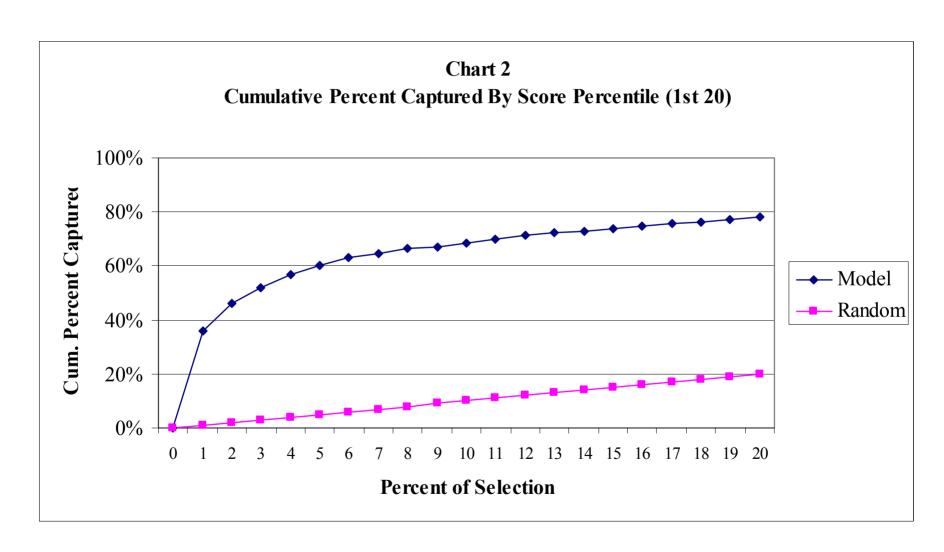
### 1a. What is Scoring?

Mining data in motion—assigning scores to data in real time.

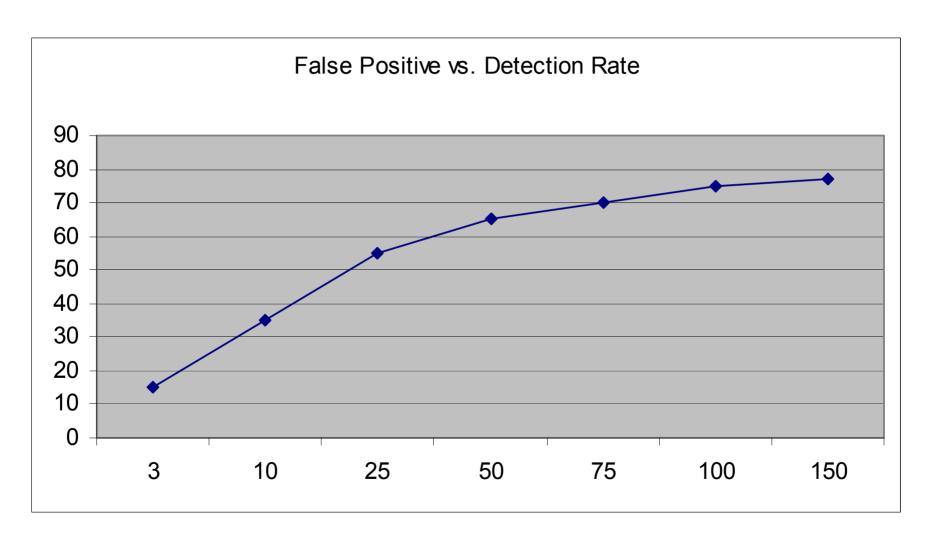
### Data Mining/Statistical Models



### **Detection Rates**



### False Positive



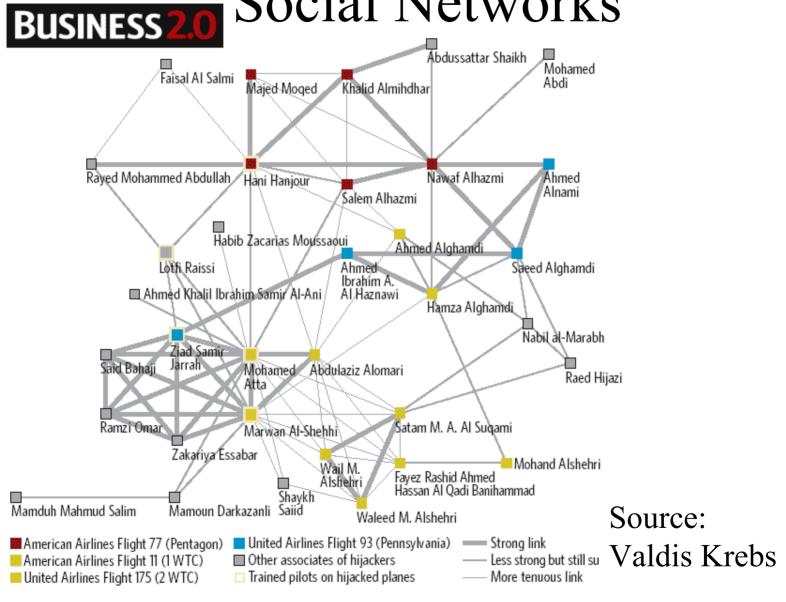
# Comparing Different Models

	Trees	Neural Networks	Rules
Accuracy	Yes	Yes	No
Easy to Maintain	Yes	No – hard to retrain	Yes – small No – large
Easy to interpret	Yes	No	Yes
Scalable to large data	Yes	No	No

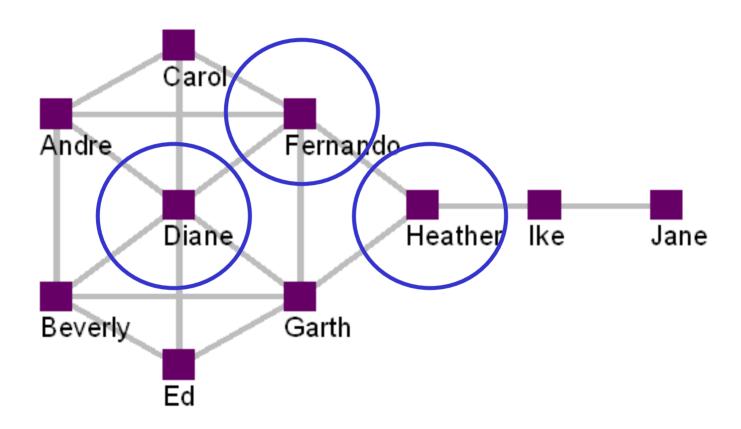
### 1b. What is Linking?

Mining data at rest — bad guys tend to hang out with other bad guys.

Social Networks



### Social Networks



degree, betweeness, & closeness

### 1c. What is Matching?

Living with watch lists and other lists of good and bad guys

### Matching



Wall Street Journal May 6, 2002.

OFAC Entry: ADEN, Abdirisak, Skaftingebacken 8, Spanga 163 67, Sweden; DOB 01 Jun 68 (individual) [SDGT]

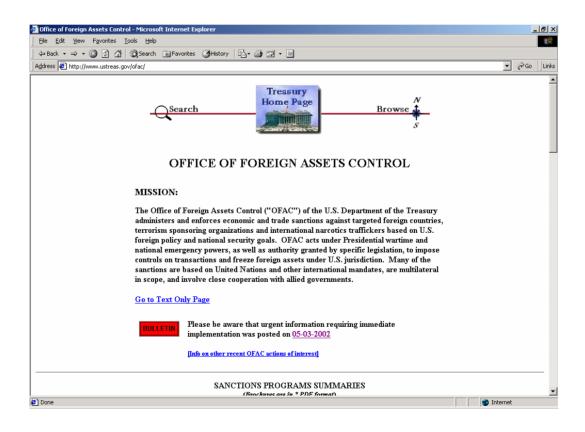
# Similarity Search Is an Important Component of a Matching System

AHMED, Ahmed (a.k.a. ALI, Ahmed Mohammed Hamed; a.k.a. ABDUREHMAN, Ahmed Mohammed; a.k.a. ABU FATIMA; a.k.a. ABU ISLAM; a.k.a. ABU KHADIIJAH; a.k.a. AHMED HAMED; a.k.a. Ahmed The Egyptian; a.k.a AL-MASRI, Ahmad; a.k.a. AL-SURÎR, Abu Islam; a.k.a. ALI, Ahmed Mohammed; a.k.a. ALI, Hamed; a.k.a. HEMED, Ahmed; a.k.a. SHIEB, Ahmed; a.k.a. SHUAIB), Afghanistan; DOB 1965; POB Egypt; citizen Egypt (individual) [SDGT]

# 1d. What is Checking?

How to stop worrying and learn to live with regulations.

# There will be more and more regulations about what data can be used and how...



### 1e. What is Routing?

0.1% of 30,000 transactions/second = 30/second at 10 minutes/investigation vs. 100 analysts and 8 hours per day.

### Routing



☐ Routing is about getting the right information to the right person at the right time.

# Trend 2. Real Time Data Mining

Exploiting Events and Profiles.

### What is an Event?



login



email



message



scan



phone calls



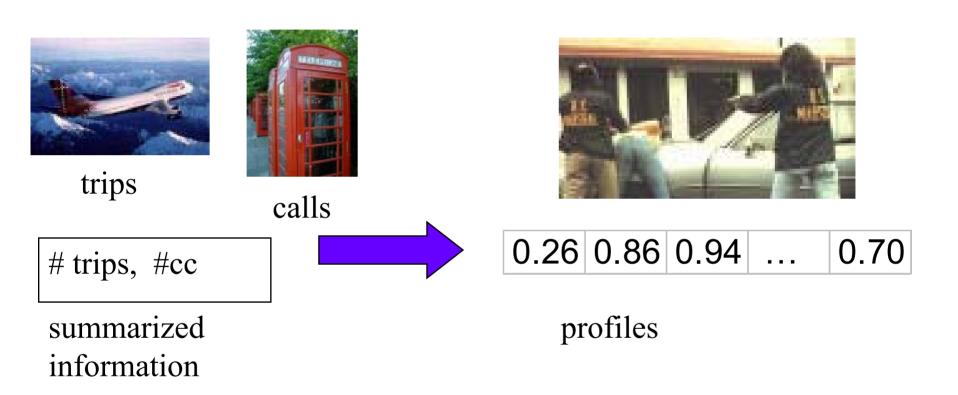
credit card transaction



cell phone call

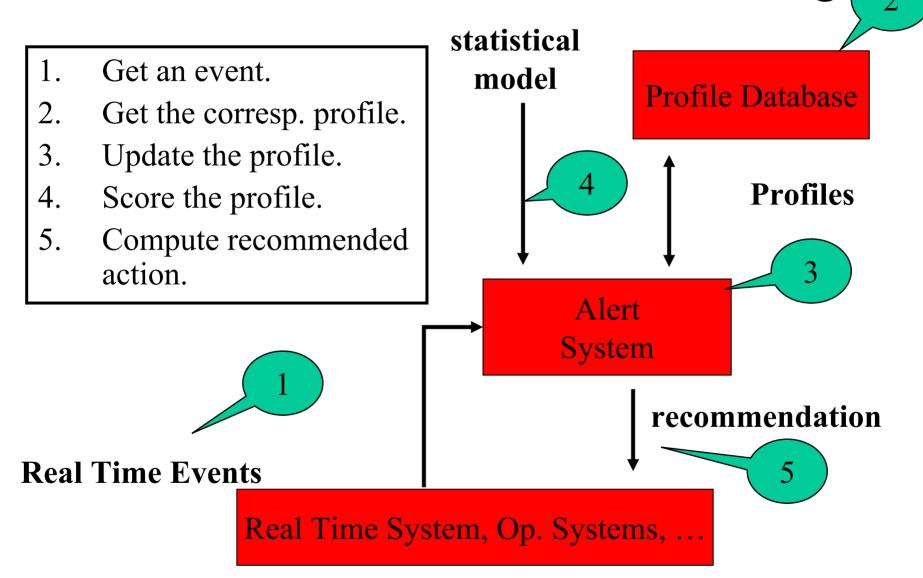
☐ An event is real time information about an entity, eg. person, place, event, threat, opportunity, etc.

### What is a Profile?



☐ A **profile** is the summarized data and attributes about an entity.

### What is Event Driven Data Mining?



### Trend 3.

In the past, we have built models and scored data at rest. In the future, more and more data will be streaming at faster and faster rates.

#### **Premises**

- □ Some data sets will be accessible via OC 12, GigE, 2.5 GigE, 10 GigE, etc. wide area networks Photonic networks.
- ☐ There will be data mining services for high performance networks, *as well as for* commodity networks.
- ☐ Many applications will trade accuracy for speed in order to keep up with line speed
- ☐ Call these Photonic Data Services (PDS)

# The Data Stack – Replacing Apps over Operating Systems

6. Data Mining Applications

5a. Storage Services

5b. Data Web Services

5b. Data Grid Services

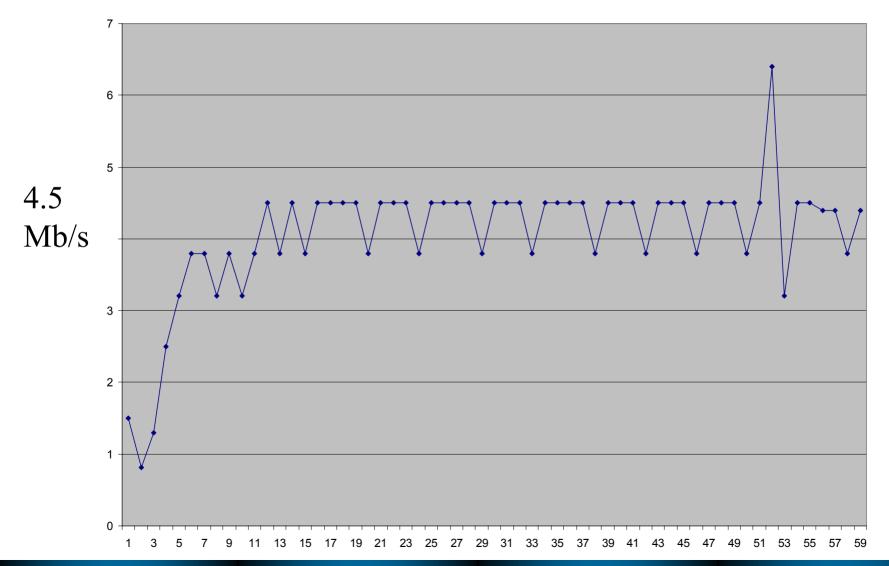
4. Transport – TCP, UDP, Reliable UDP

3. IP

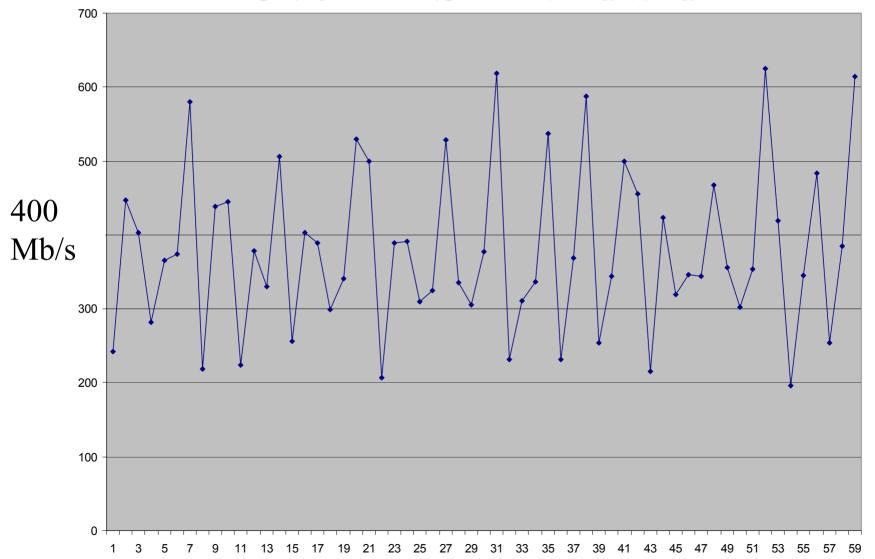
2. Photonic Path Services

1. Physical

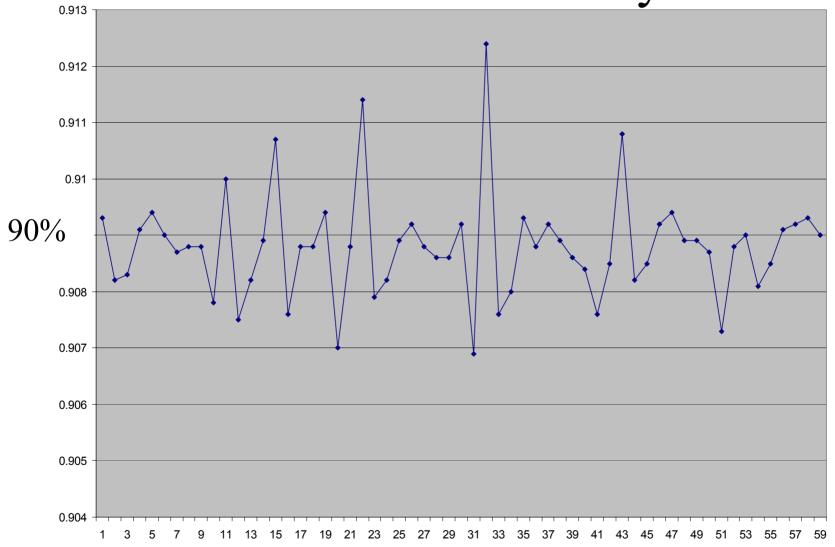
# TCP Data Transport Chicago to Amsterdam over 622 Mb/s Link



### Best Effort Distributed Merge Over PDS - Bandwidth



# Best Effort Distributed Merge over PDS - Accuracy



# Trend 4. Data Webs for Data Exploration

We have developed some good data mining algorithms & systems, we need better algorithms and systems for data *exploration*.

### Paradigm Today

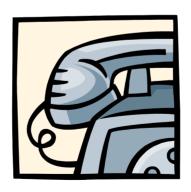
- 1. Learn about a new data source from a friend or over the web
- 2. FTP the data, federal express, or courier the data
- 3. A consultant or contractor spends a 1-3 months and then tells you whether or not to start a project to build a centralized data warehouse

### Emerging Paradigm: Data Webs

- ☐ Designed for quick overlays by key
- ☐ Designed for quick correlations
- ☐ If interesting, perform traditional exploratory data analysis, statistical modeling, etc.
- ☐ Reduce number of clicks to correlate two data sets



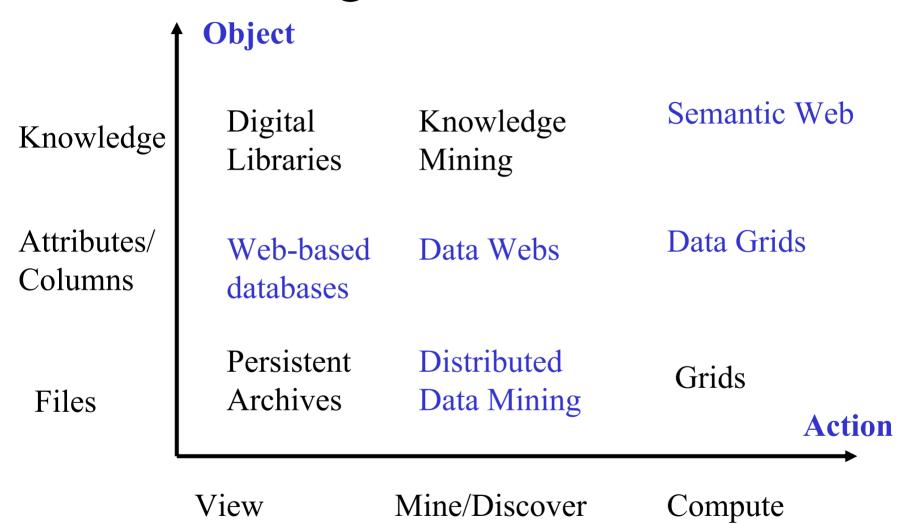
names



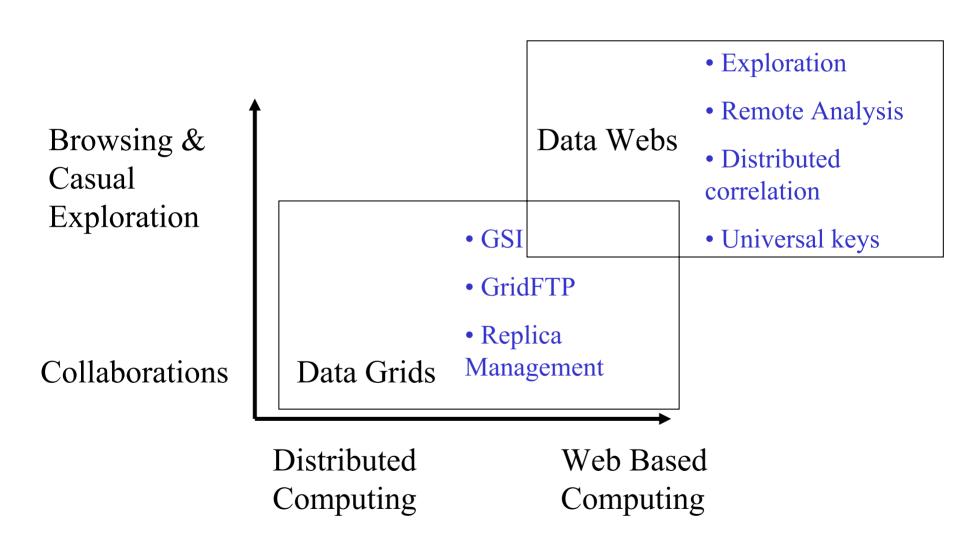
phone numbers



### Technologies for Global Data



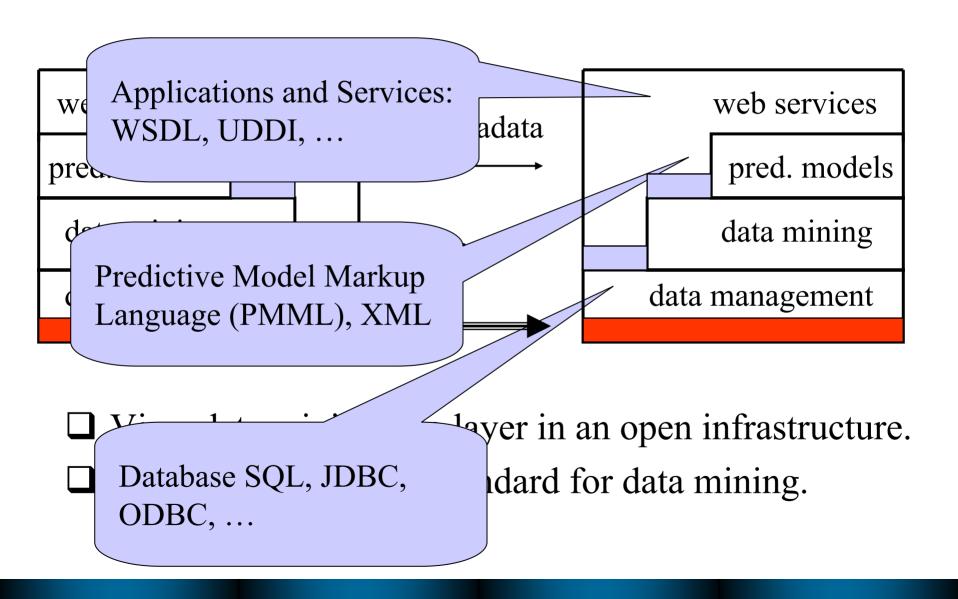
#### Data Grids vs. Data Webs



### Trend 5.

Standards are maturing.

# Maturing Standards



### Data Mining Group

- □ Products shipping with PMML Version 1.1
- □PMML Working Group Full Members
  - IBM, Magnify, Microsoft, MineIt, NCR, Oracle,
     Salford Systems, SAS, SPSS, xChange, University
     of Illinois at Chicago (over 20 vendors)
- □PMML Working Group Supporting Members
  - Angoss, Insightful, KXEN, Microsoft, SGI ...
- ☐Part of Source Forge

# Problems with Current Techniques

- ☐ Models are deployed in proprietary formats
- ☐ Models are application dependent
- ☐ Models are system dependent
- ☐ Models are architecture dependant
- ☐ Time required to integrate models with other applications can be long.

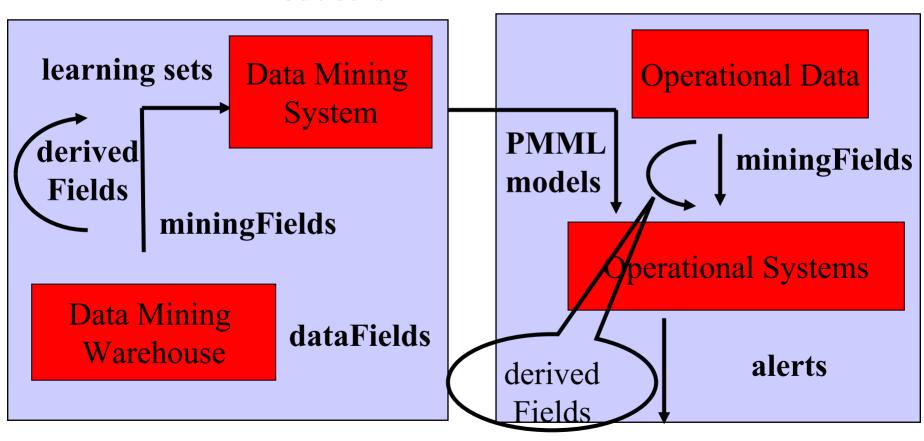
# Predictive Model Markup Language (PMML)

- ☐ Based on XML
- ☐ Benefits of PMML
  - Open standard for Data Mining & Statistical Models
  - Not concerned with the process of creating a model
  - Provides independence from application, platform, and operating system
  - Simplifies use of data mining models by other applications (consumers of data mining models)

### PMML Producers, Consumers, & Data Flow

#### **PMML Producers**

#### **PMML Consumers**



### Closely Related Standards

OMG CWM DM

Object model for representing data mining metadata: models, model results (UML/DTD/XML)

SQL-like interface for data mining operations (OLE DB/SQL) OLE DB

DMG PMML

Representation of data mining models for intervendor exchange (DTD/XML) JSR-073

SQL/MM Pt. 6 DM

**JDMAPI** 

SQL objects for defining, creating, and applying data mining models, and obtaining their results (SQL)

Java API for defining, creating, and applying data mining models, and obtaining their results (Java)

# Summary: Cyber Threat Analysis

- 1. Deployment is more about *alert management* than which algorithm.
- 2. Events and Profiles enable event driven applications.
- 3. There is a fundamental need to design algorithms for *high bandwidth* data streams, at 1 Gb/s and higher.
- 4. The best way to improve a model is to join new from a new source. Data web and *data exploration* systems are designed to make this easier.
- 5. Standards for data mining are maturing.

### For More Information

#### Robert Grossman

grossman at uic.edu or rlg at opendata.biz www.lac.uic.edu, www.opendata.biz, www.rgrossman.com,

#### Standards

www.dmg.org (PMML, DWTP, etc.)

#### Data Webs

www.dataspaceweb.net or info at ac.uic.edu

#### **Testbed**

Terra Wide Data Mining Testbed (TWDM) Terabyte Challenge Testbed www.ncdm.uic.edu