The High ROI of Data Mining for Innovative Organizations

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Case Study Lessons Outline

- The Hype Cycle vs. Real Results
- 3 Major ways Data Mining helps:
 - Eliminate the bad:
 - Examples from IRS, Hewlett-Packard, Capital One
 - Discover the good:
 - Ex: Pfizer, Westwind Foundation
 - Streamline / Automate:
 - Ex: Lumidigm, Peregrine, SSA, Anheuser-Busch
- Lessons Learned: features of successful projects

Book written *for* practitioners, *by* practitioners

Published by Elsevier's Academic Press in May 2009

Details and reviews at Amazon: www.tinyurl.com/bookERI

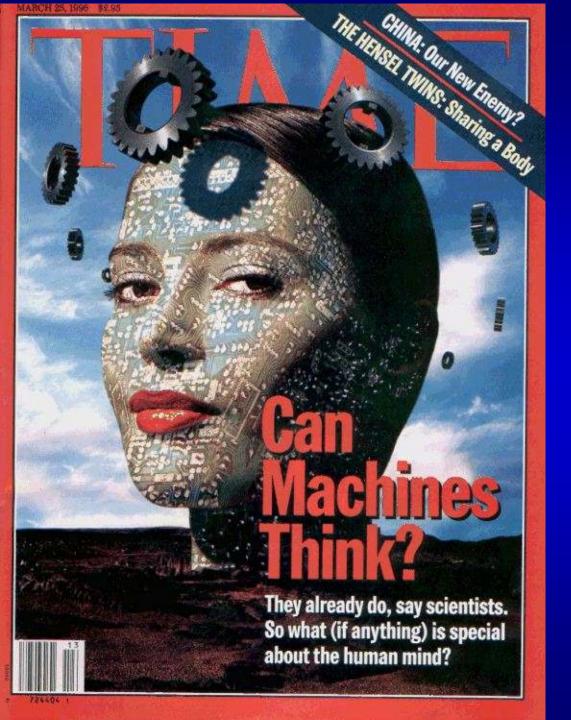
Won 2009 PROSE Award for Mathematics

ROBERT NISBET JOHN ELDER GARY MINER

HANDBOOK OF Statistical Analysis & Data Mining Applications



QuickTime™ and a decompressor are needed to see this picture



"Of course machines can think. After all, humans are just machines made of meat."

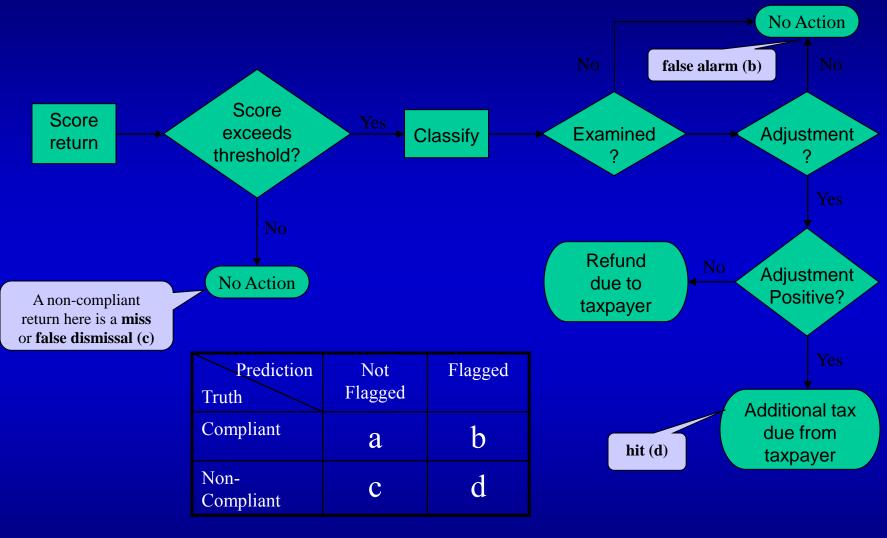
- MIT CS professor

Human and computer strengths are more complementary than alike.

3 Major Ways Data Mining Helps

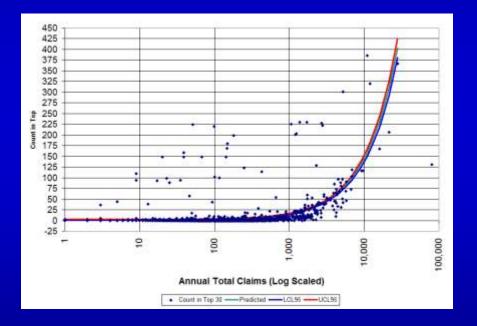
- Eliminate the Bad
 - Score credit risk, Detect fraud, Find outliers
- Highlight the Good
 - Time market trades, Discover new drugs, Uncover hidden value
- Streamline or Semi-Automate Decisions
 - Intuit products of interest, Assess status, Verify identity, Be aware of "what if" scenarios, Speed service

Case 1) IRS Fraud Detection



QuickTime™ and a decompressor are needed to see this picture. detection rate = d/(c+d)hit:scan = 1 : (b+d)/d workload = b+d

Case 2) Hewlett-Packard: Service Fraud Detection



- Tips indicated fraud exists
- Goal: Learn from known cases to find unknown
- Automate current process, build model on known, score all data, investigate top
- Recovered \$20M in 9 months
- Awards + promotions + growth Became profit center

Case 3) Lumidigm: Bio-Metrics (id not by what you have or know, but what you are)



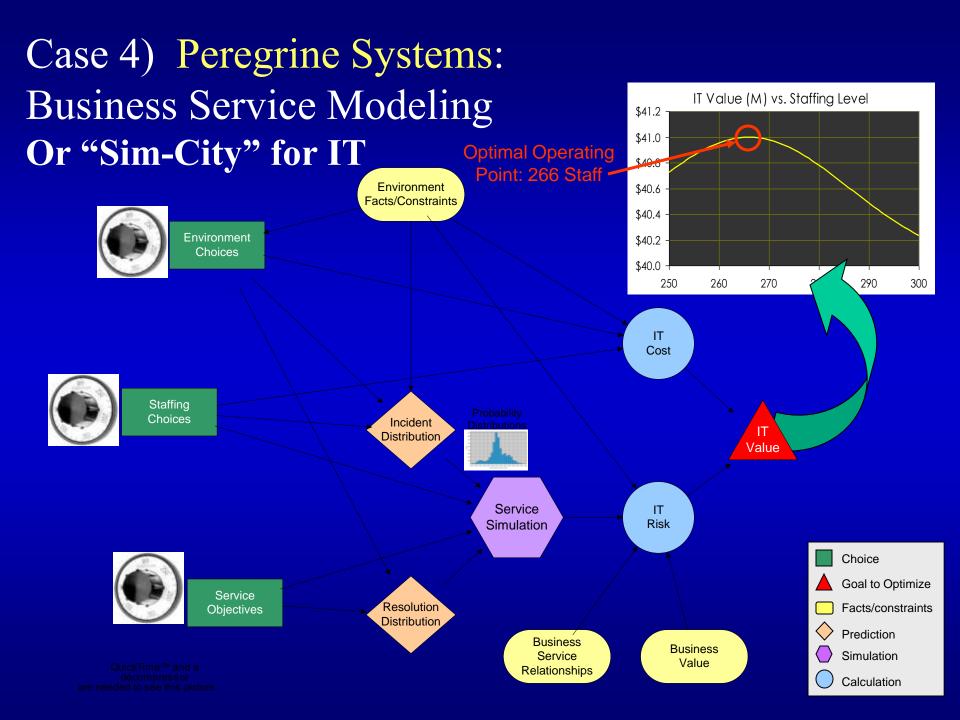
- Original Goal: Diagnose disease
- Data: infra-red skin reflections at multiple frequencies
- Problem: hard to remove person specifics
- Idea: Turn problem into gain: new bio-metric
- Q: *Who* are you?
- Use to validate ID for safety or convenience

Solution in search of the right Application



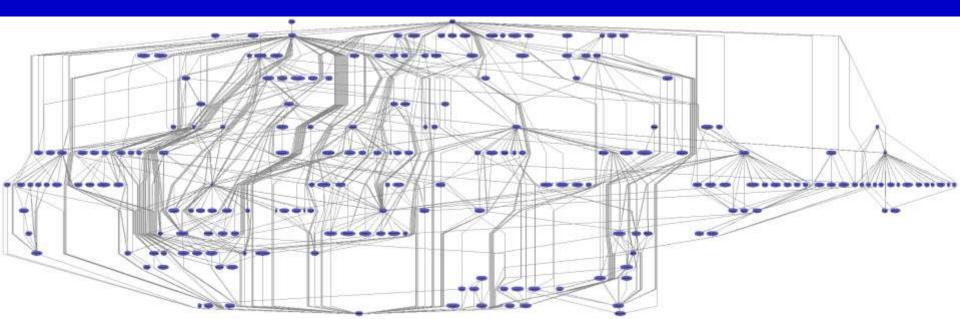


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Goal: Simulation & Optimization

- Build tool to manage IT Service (help tickets)
- New product critical to company (& worked!)
- Analytic breakthrough: keep uncertainty to end
- Success: purchased by HP; increased HP sales



Case 5) Social Security Administration Disability Approval

• Pain: Approval process is long, bureaucratic

Up to 2 Years With Textals and a second seco

1/2 of appeals overturn original decision

- Goal: Fast-track "easy" cases
- Challenge: Free-text on disability application
- **Result**: 20% of Approvals possible immediately and with greater consistency

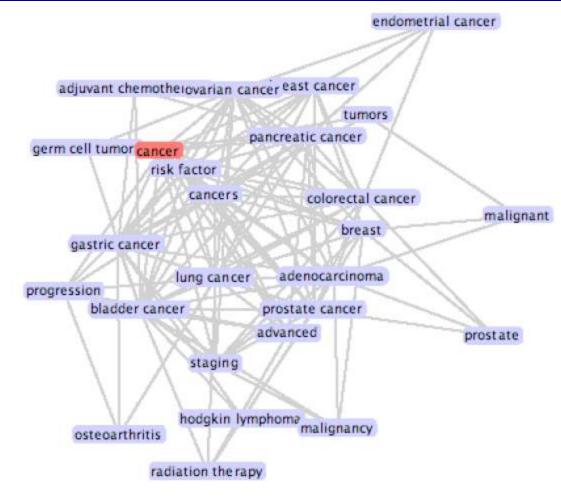
Text is Messy and Complex

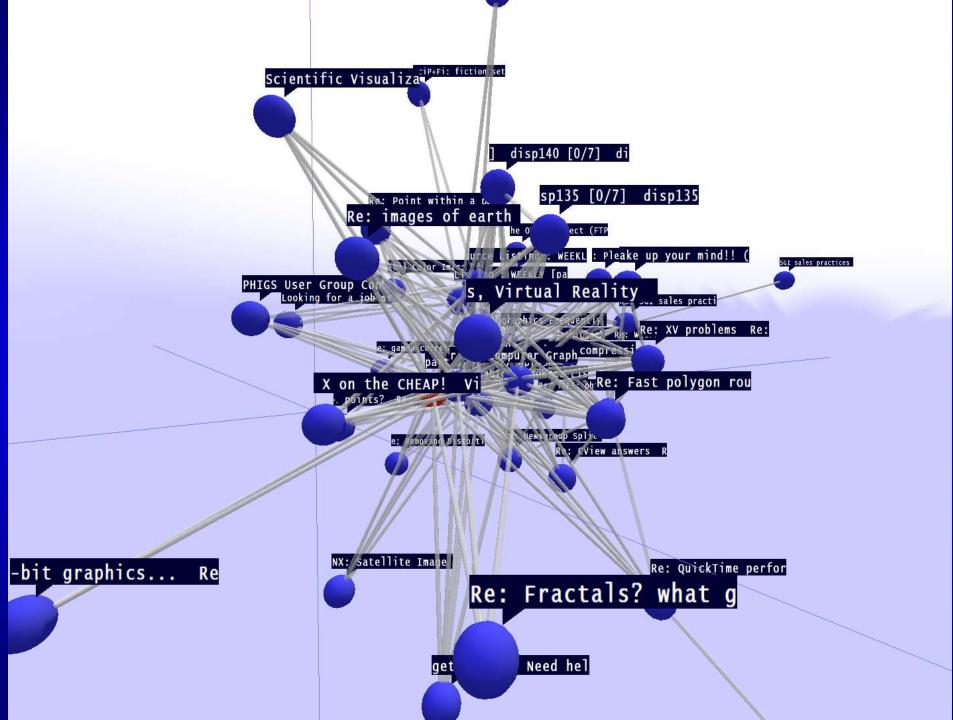
- Multi-Word Phrases (Concepts or Lemmas)
- Stemming (ex: Learn = Learning, Learned, Learns...)
- Synonyms (ex: ALS = Lou Gherig's Disease)
- Misspellings

Ex: 51 phrases found for "Learning Disability"

learning disability, learning deisability, learning disability, learning disabilities, learning disability, learni

ERI new technology: Discover web of word relations

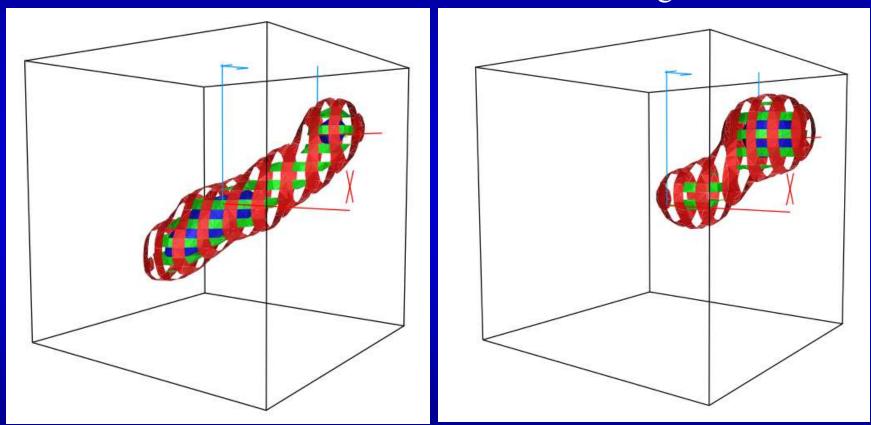




Case 6) Pharmacia & Upjohn (Pfizer) New Drug Effective?

Placebo

Drug



Density surfaces enclose ascending quartiles of data

QuickTime™ and a decompressor are needed to see this picture.

Bottom Line

- IRS fraud detection
- HP consumer fraud detection
- Lumidigm biometric identification
- Peregrine optimization
- SSA text mining
- Pharmacia & Upjohn (Pfizer) drug efficacy

Lessons Learned: Necessary Ingredients for Analytic Project Success

- Gain Expected: either:
 - Leverageable an incremental improvement will matter, OR
 - "Low-hanging fruit" nobody's yet (dared) attack the problem
- Interdisciplinary Team: experts needed in business area, statistics, algorithms, and databases
- Data Vigilance: capture and maintain the accumulating information stream
- Time: learning occurs over multiple cycles
- Business Champion essential!

... then Data Mining can add extraordinary value



John F. Elder IV Chief Scientist, Elder Research, Inc.

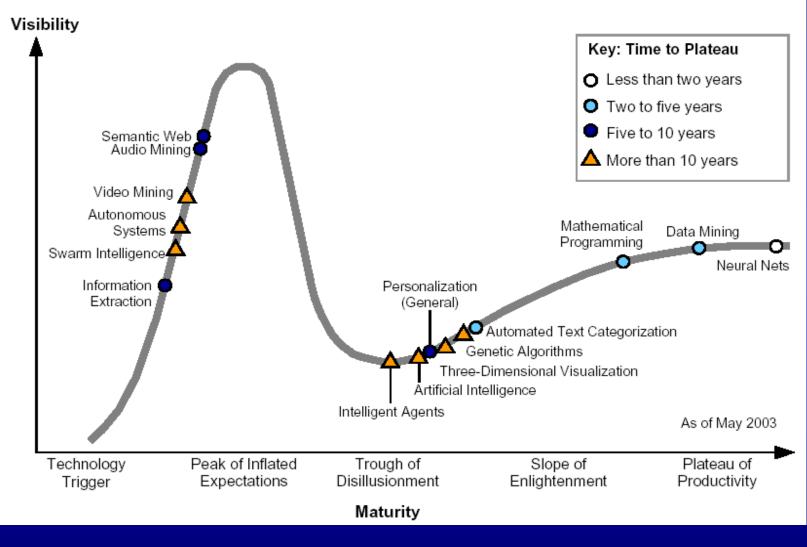
DR. JOHN ELDER HEADS A DATA MINING CONSULTING TEAM WITH OFFICES IN CHARLOTTESVILLE, VIRGINIA, WASHINGTON DC, MOUNTAIN VIEW, CALIFORNIA AND MANHASSET NEW YORK (*WWW.DATAMININGLAB.COM*). FOUNDED IN 1995, ELDER RESEARCH, INC. FOCUSES ON FEDERAL, COMMERCIAL, INVESTMENT, AND SECURITY APPLICATIONS OF ADVANCED ANALYTICS, INCLUDING TEXT MINING, STOCK SELECTION, IMAGE RECOGNITION, BIOMETRICS, PROCESS OPTIMIZATION, CROSS-SELLING, DRUG EFFICACY, CREDIT SCORING, RISK MANAGEMENT, AND FRAUD DETECTION.

JOHN OBTAINED A BS AND MEE IN ELECTRICAL ENGINEERING FROM RICE UNIVERSITY, AND A PHD IN SYSTEMS ENGINEERING FROM THE UNIVERSITY OF VIRGINIA, WHERE HE'S AN ADJUNCT PROFESSOR TEACHING OPTIMIZATION OR DATA MINING. PRIOR TO 15 YEARS AT ERI, HE SPENT 5 YEARS IN AEROSPACE DEFENSE CONSULTING, 4 HEADING RESEARCH AT AN INVESTMENT MANAGEMENT FIRM, AND 2 IN RICE'S *COMPUTATIONAL & APPLIED MATHEMATICS* DEPARTMENT.

DR. ELDER HAS AUTHORED INNOVATIVE DATA MINING TOOLS, IS A FREQUENT KEYNOTE SPEAKER, AND WAS CO-CHAIR OF THE 2009 *KNOWLEDGE DISCOVERY AND DATA MINING* CONFERENCE, IN PARIS. JOHN'S COURSES ON ANALYSIS TECHNIQUES - TAUGHT AT DOZENS OF UNIVERSITIES, COMPANIES, AND GOVERNMENT LABS - ARE NOTED FOR THEIR CLARITY AND EFFECTIVENESS. DR. ELDER WAS HONORED TO SERVE FOR 5 YEARS ON A PANEL APPOINTED BY THE PRESIDENT TO GUIDE TECHNOLOGY FOR NATIONAL SECURITY. HIS BOOK ON DATA MINING, WITH BOB NISBET AND GARY MINER, WON THE 2009 PROSE AWARD FOR MATHEMATICS. A BOOK ON ENSEMBLES WITH GIOVANNI SENI WAS PUBLISHED IN 2010.

QuickTime™ and a decompressor are needed to see this picture. JOHN IS A FOLLOWER OF CHRIST AND THE PROUD FATHER OF 5.

Data Mining and the Hype Cycle



Case 7) HSBC Cross-sell / Up-sell



- Q: What product will interest a customer next?
- Reason: Better target marketing campaigns
- New: Turn contact (a cost) into gain
- Data: transaction history

Case 7) Key Technology: Custom Association Discovery Tool: *QuiltMaker* HSBC Quilt Maker V2.1

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Visualize and Quantify next-most-interesting profitable product

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Ready

Case 8) Anheuser-Busch Image Recognition



6-ca	n an 6-bo	ot 6-b	oot 6-b	ot 6-bo	ot	6-bot	6-bot	6-bot
6-bot	6-bot	6-bot	6-can	12-ca	n	12-0	can	
6-bot	6-bot	6-bot	6-can 6-can	12-ca		12-can		

- Q: How do A-B products appear in the store?
- Reason: Discover and implement configurations that work
- Currently: Takes 4 hours to record configurations
- Goal: automate
- Data: images

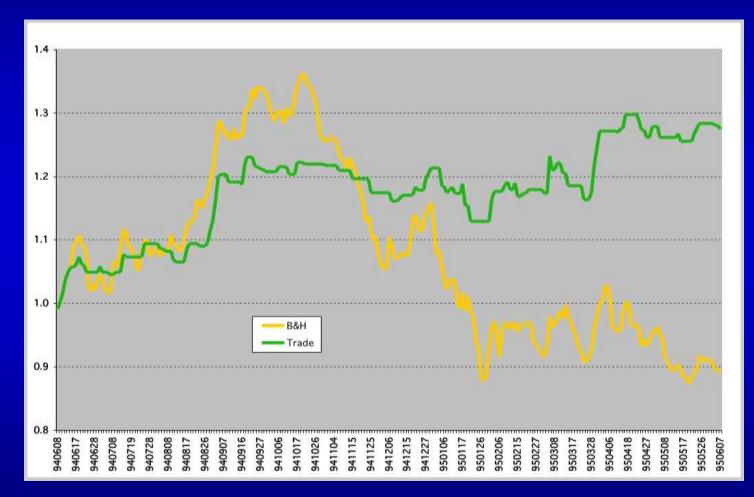


- Automation 90% accurate
 -> 10-fold speed-up
- Id'd "stock outs" and "competitor creep"
- Key technological breakthrough: iterative identification

"Planogram": symbolic summary of product configuration. Used for planning and analysis



Case 9) WestWind Foundation: Hedge Fund Strategy to Time Market



Resampling Technology used to determine whether gains were "real":

READ file "fund_1yr" date position return MULTIPLY position return trade SUM trade original PRINT original

REPEAT 1000 SHUFFLE position pos MULTIPLY pos return trade SUM trade total SCORE total Z END HISTOGRAM Z

COUNT Z > original better DIVIDE better 1000 prop_bet PRINT prop_bet

