### VISUAL ANALYTICS

**APPLICATION AREAS** 

**LECTURE 8** 

Petra Isenberg



### TOC

- Business Intelligence
- Legal Matters
- Work at Aviz

## ANALYSIS IN THE BUSINESS SECTOR

#### **BUSINESS INTELLIGENCE**

"An interactive process for exploring and analyzing structured and domain-specific information to discern trends or patterns, thereby deriving insights and drawing conclusions.

The business intelligence process includes communicating findings and effecting change."

Gartner

→ there are many more definitions and they don't all say the same thing

### BUSINESS INTELLIGENCE

- Typical Goals:
  - increase profitability
  - decrease costs
  - improve customer relationship management
  - decrease risks (e.g. credit risk analysis)
- Main goal: aid in making decisions

## ANALYSIS IN ENTERPRISES IN GENERAL

- business analysts
- data analysts
- data scientists
- typical user population for analysis & visualization tools

Sean Kandel, Andreas Paepcke, Joseph M. Hellerstein, Jeffrey Heer: Enterprise Data Analysis and Visualization: An Interview Study. IEEE Trans. Vis. Comput. Graph. 18(12): 2917-2926 (2012

### **A STUDY**

researchers conducted semi-structured interviews in enterprises:

- 35 analysts (26 male)
- 25 organizations: healthcare, retail, finance, social networking, insurance, ...

Sean Kandel, Andreas Paepcke, Joseph M. Hellerstein, Jeffrey Heer: Enterprise Data Analysis and Visualization: An Interview Study. IEEE Trans. Vis. Comput. Graph. 18(12): 2917-2926 (2012

#### QUESTIONS

- What tasks do analysts perform?
- What kinds of data sources and formats do they work with?
- What tools do they regularly use and how do they use them?
- · How do analysts vary in terms of programming proficiency?
- How do analysts vary in terms of statistical proficiency?
- What are the "results" of analysis?
- What happens to these results "downstream"?
- What are recurring bottlenecks and pain points?
- How important is scalability?
- How important is sociability?
- What is the relationship between analysts and other business units?
- Where are analysts situated within their corporate hierarchy?

### THE ANALYST

#### Three archetypes found

#### 1) hackers

- most comfortable manipulating data
- used 3+ programming languages (R, Python, SQL, ...)
- complex workflows
- work quite isolated (don't need a lot of help)
- used visualizations: Tableau, Excel, PPT, D3, ...

### THE ANALYST

## Three archetypes found 2) scripters

- most analysis done in R/Matlab
- not versed in custom operations (parsing, scraping)
- generally worked on data from data warehouse (with help from IT staff)
- applied models to data
- did exploratory analysis with visualization

### THE ANALYST

#### Three archetypes found

- 3) application user
  - performed most operation in spreadsheet or analysis app (SAS/JMP, SPSS, ...)
  - needed help preparing data
  - typically worked on smaller datasets

### ANALYSTS

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### IT TEAM

- Crucial role in process
  - getting, maintaining, accessing data
  - operationalize workflows
  - provide documentation

### OTHER FINDINGS

- data:
  - stored in variety of repositories and formats
- consumers of analysis:
  - many different departments, also other analysts
  - static reports shared most commonly, sometimes dynamic dashboards

### OTHER FINDINGS

- collaboration
  - work on analysis task mostly done individually
  - resources shared, however: data, scripts, results, documentation

- discovery
  - where is my data?
  - what does my data mean? (unclear field names, missing units, timezones, ...)

- wrangling
  - processing semi-structured data
  - data integration from multiple sources
  - advanced aggregation and filtering

I spend more than half of my time integrating, cleansing and transforming data without doing any actual analysis. Most of the time I'm lucky if I get to do any analysis. Most of the time once you transform the data you just do an average... the insights can be scarily obvious. It's fun when you get to do something somewhat analytical.

- profiling
  - assessing & improving data quality
     (missing data, wrong formats, ...)
  - checking assumptions
     (data distribution, semantics of data, ...)

- modeling
  - finding the right features to analyze
  - scale of data
  - visualization of statistical models missing

- reporting
  - how to document assumptions
  - flexibility in reports missing

- workflows are non-linear
  - moving data between tools necessary (creates formatting issues)
  - creating repeatable, reliable, and scalable workflows

### BUSINESS ANALYSIS TOOLS

a research view

### TOOLS PREDECESSORS

- Management Information Systems (MIS)
- Management Decision Systems (MDS)
- Decision Support Tools (DSS)
- Executive Information Systems (EIS)
- Analysis Information Systems (AIS)

• ...

### TOOLS

Warning: I haven't tried any of them, so can't make recommendations

- Professional
  - SAP Business Intelligence
  - IBM Cognos

**—** ...

- Open Source
  - SpagoBl
  - Pentaho

**—** ...

- Research
  - see next slides

### DOTLINK360

Basole, R.C.; Clear, T.; Mengdie Hu; Mehrotra, H.; Stasko, J., "Understanding Interfirm Relationships in Business Ecosystems with Interactive Visualization," *Visualization and Computer Graphics, IEEE Transactions on*, vol.19, no.12, pp.2526,2535, Dec. 2013 doi: 10.1109/TVCG.2013.209

2013

#### dotlink360

Understanding Interfirm Relationships in Business Ecosystems with Interactive Visualization



Rahul C. Basole | Trustin Clear | Mengdie Hu | Harshit Mehrotra | John Stasko contact: basole@gatech.edu



# BUSINESS ECOSYSTEM INTELLIGENCE

- market analysts
  - understand competitive trends/strategies/threats/opportunities
- executives
  - identify strategic collaborations & customers, find areas for innovation
- venture capitalists
  - identify investment opportunities, see how they fit in the business landscape

### DOTLINK360



### DESIGN PROCESS

- field study of analysts
- derive set of design requirements
- develop initial version
- user testing
- redesign

### DATA SOURCES

- Thomson Reuters SDC Platinum database
  - commercial database
  - ca. 700,000 global alliances, agreements, joint-ventures, since 1985
- Capital IQ Compustat
  - contains e.g. quarterly financial and accounting data for active and inactive publicly listed companies

### DATA TRANSFORMATION

- turned data into a network
  - nodes = companies
  - edges = agreements
- nodes have multiple attributes
- edges have multiple types
- +temporal data
- → time-varying multivariate network

### DESIGN REQUIREMENTS

- Field study
  - online survey + interviews
  - 24 senior industry individuals (market analysts, executives, venture capitalists)
  - each >10 years of experience

### DESIGN REQUIREMENTS

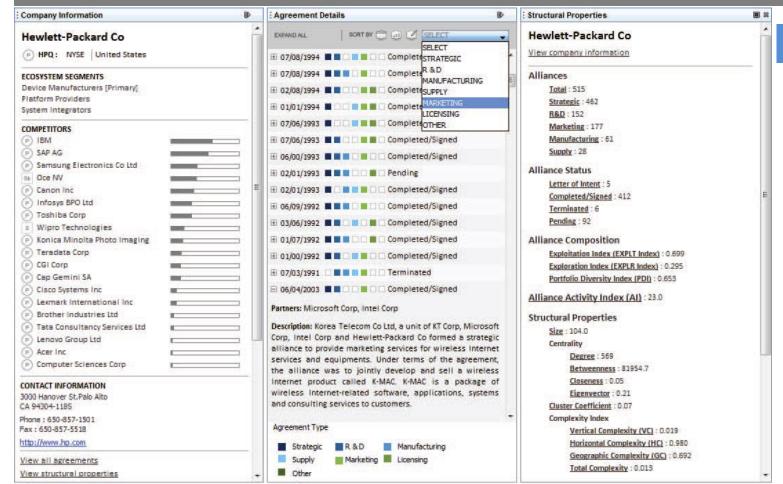
top-down & bottom-up examination of an ecosystem are critical

"it is helpful to have a birds-eye view of the ecosystem, but at the same time [the system] should be able to allow users to drill down into individual companies and segments."

### DESIGN REQUIREMENTS

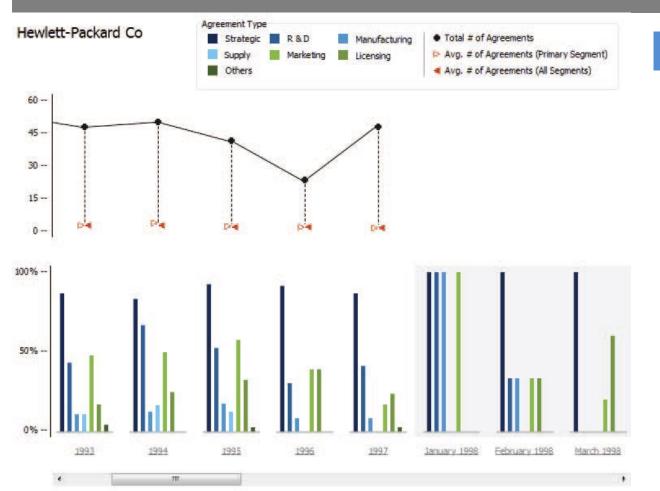
- understanding interfirm connectivity, composition, and temporality is vital
- comparative perspectives drive insight
- first: communicate agreement summaries (structural information) → then: offer details
- provide a familiar metaphor while supporting direct and prompt interaction, not complex queries and commands
- add common network-related analysis tasks (see the InfoVis lecture on graphs and networks)

### MULTIPLE VIEWS



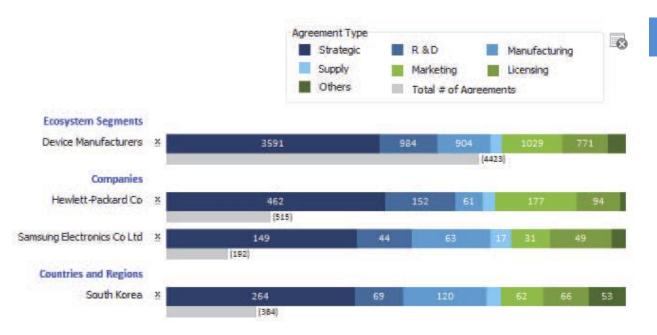
details

### **MULTIPLE VIEWS**



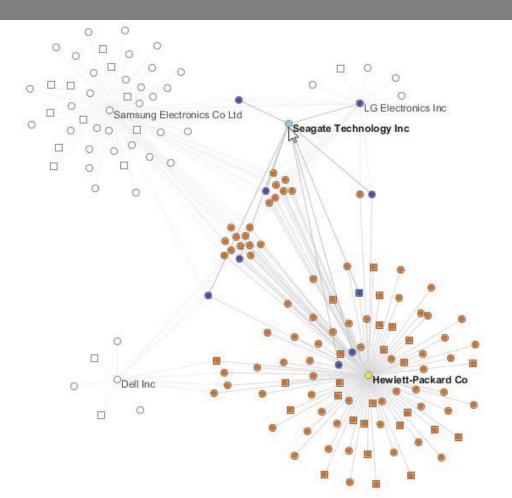
timeline

### MULTIPLE VIEWS



portfolio overview

## MULTIPLE VIEWS



comparison

... and many more

Remco Chang, Mohammad Ghoniem, Robert Kosara, William Ribarsky, Jing Yang, Evan Suma, Caroline Ziemkiewicz, Daniel Kern, Agus Sudjianto

IEEE Visual Analytics Science and Technology (VAST) 2007.

- financial institutions have obligation to discover suspicions financial transactions
  - can be fined if not found or shut down
- many transactions are purely digital & banks often are only the middle "man"
- large banks need to monitor hundreds of thousands of transactions per day

## WIREVIS

Remco Chang, Mohammad Ghoniem, Robert Kosara, William Ribarsky, Jing Yang, Evan Suma, Caroline Ziemkiewicz, Daniel Kern, Agus Sudjianto

IEEE Visual Analytics Science and Technology (VAST) 2007.

## WireVis

Visualization of Categorical, Time-Varying Data From Financial Transactions

**UNC Charlotte** 

Remco Chang Mohammad Ghoniem Robert Kosara William Ribarsky Jing Yang Evan Suma Caroline Ziemkiewicz

Bank of America

Daniel Kern Agus Sudjianto

### WIREVIS

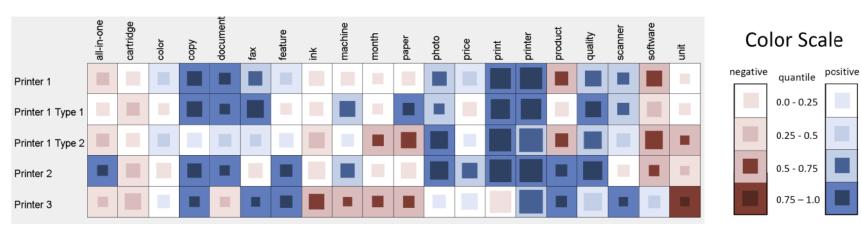
### goals

- provide overview of transactions over any period of time → apply to large data
- allow identification of patterns over time and keywords
- replace blind queries with in-place analysis
- provide search-by-example technique

### WIREVIS

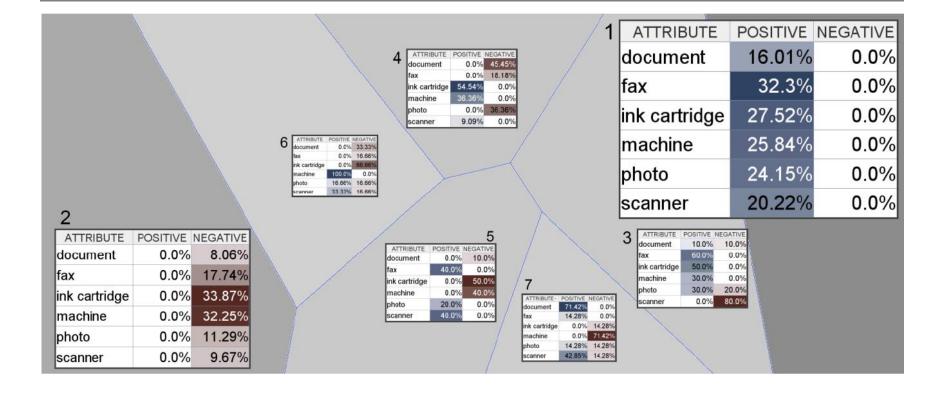
- wire transaction data
  - semi-structured data record
  - fixed data (sender, receiver, etc.)
  - optional free text
  - temporal, categorical, quantitative
- current (2007) procedure
  - filter based on risk assessment (e.g. money exceed threshold, sender/receiver is a highrisk country or organization)
  - look at spreadsheets

# VISUAL ANALYSIS OF CUSTOMER FEEDBACK



Oelke, D.; Ming Hao; Rohrdantz, C.; Keim, D.A.; Dayal, U.; Haug, L.; Janetzko, H., "Visual opinion analysis of customer feedback data," *Visual Analytics Science and Technology, 2009. VAST 2009. IEEE Symposium on*, vol., no., pp.187,194, 12-13 Oct. 2009 doi: 10.1109/VAST.2009.5333919

## VISUAL ANALYSIS OF CUSTOMER FEEDBACK



## LEGAL MATTERS

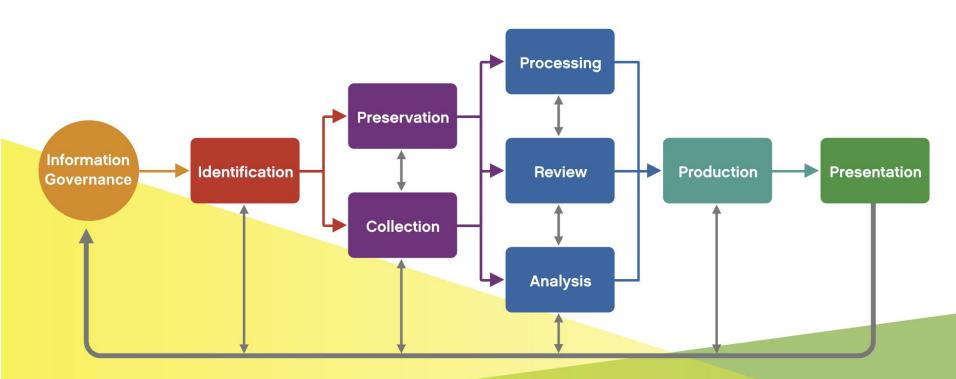
or

working with large corpora of electronic texts

## E-DISCOVERY (MOSTLY USA)

- legal electronic document discovery
  - for use in law suits
  - regulatory information requests
  - investigations, audits, freedom of information act, ...
- documents typically given to court & opponent
- if complaint received, corporations have to produce all related information

#### **Electronic Discovery Reference Model**



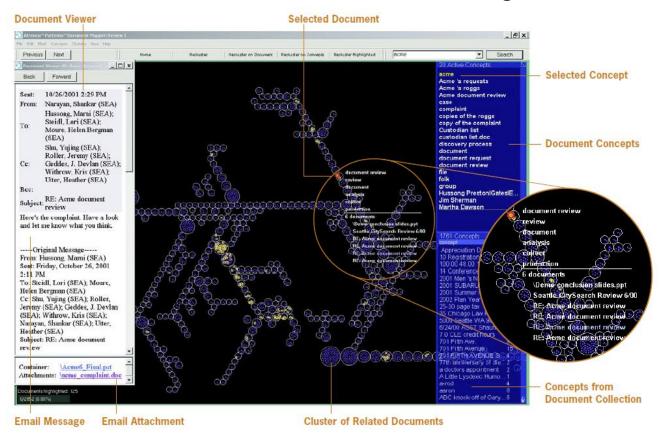
**VOLUME** 

RELEVANCE



## **EXAMPLE TOOLS**

#### Commercial: Attenex Patterns (now FTI Ringtail)



## SUCCESS STORIES OF VA TOOL

- Lovells (6<sup>th</sup> largest law firm worldwide) was investigating 35Gb of email data
  - traditional method: estimated cost: 1 year, 4-5 million \$US
  - with VA software: three months, cost 1 million \$US
- JMBM law form represented spinal surgeon & inventor in intellectual property suit
  - 20 claims, 50 million pages of documents, contracts, patents, etc.
  - with VA software (used on 7 computers), 44 million pages sifted through in 4 months
  - important doc found, surgeon won, received \$570 million in compensation

Figure 1. Magic Quadrant for E-Discovery Software



Source: Gartner (June 2014)

## RESEARCH

- working on this level in research is difficult
  - man power for providing support
  - "research aspect" needs most attention
  - adoption of prototypes unlikely when stakes are high (\$\$\$)
- thus, next: smaller research projects in regards to document analysis

## PARALLEL TAG CLOUDS

C. Collins, F. B. Viégas, and M. Wattenberg, "Parallel Tag Clouds to Explore and Analyze Facted Text Corpora," in *Proc. of the IEEE Symp. on Visual Analytics Science and Technology (VAST)*, 2009.

Parallel Tag Clouds to Explore and Analyze Faceted Text Corpora

Christopher Collins Fernanda B. Viégas Martin Wattenberg

### FACET ATLAS

Nan Cao; Jimeng Sun; Yu-Ru Lin; Gotz, D.; Shixia Liu; Huamin Qu, "FacetAtlas: Multifaceted Visualization for Rich Text Corpora," *Visualization and Computer Graphics, IEEE Transactions on*, vol.16, no.6, pp.1172,1181, Nov.-Dec. 2010 doi: 10.1109/TVCG.2010.154

FacetAtlas: Multifaceted Visualization for Rich Text Corpora

InfoVis 2010

NanCao, Jimeng Sun, Yu-Ru Lin, David Gotz, Shixia Liu, Huamin Qu

## CAMBIERA

## Collaborative Brushing and Linking for Co-located Visual Analytics of Document Collections

Petra Isenberg & Danyel Fisher Microsoft Research