
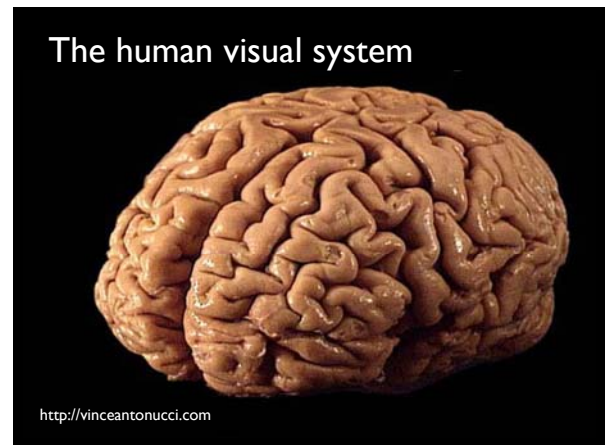
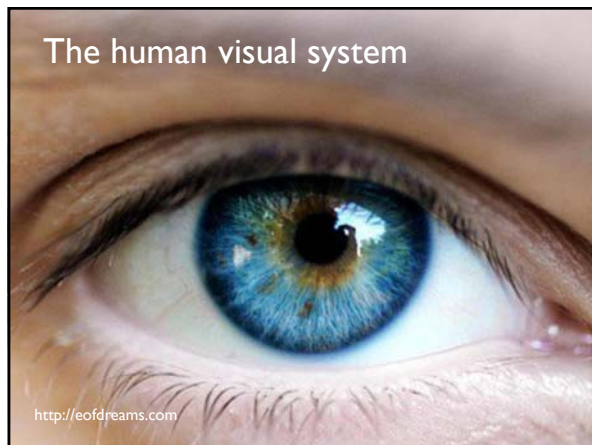
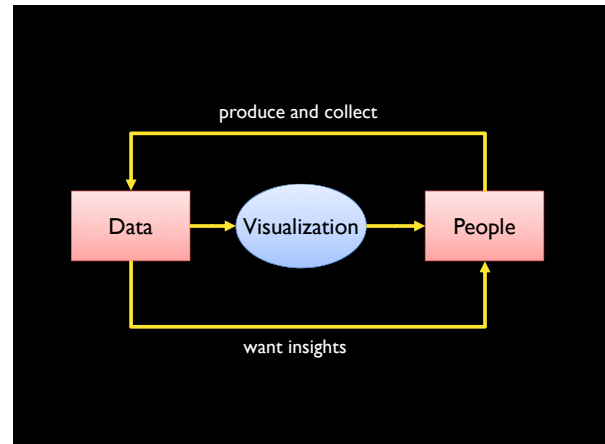


**Visualization for Transparency**  
 Jack van Wijk  
 Responsible Data Science Seminar  
 Amsterdam  
 December 14<sup>th</sup>, 2016



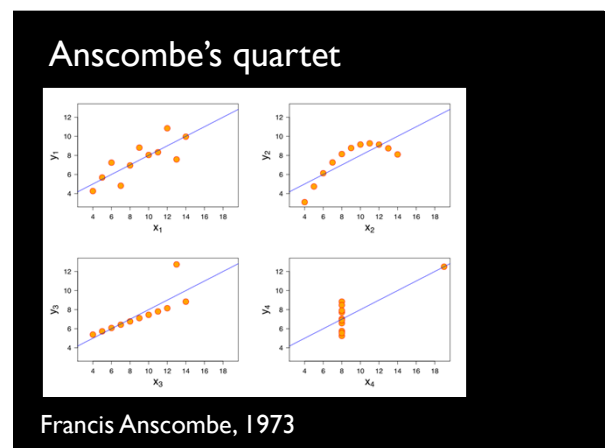
**TU/e** Technische Universiteit Eindhoven  
 University of Technology  
 Where innovation starts

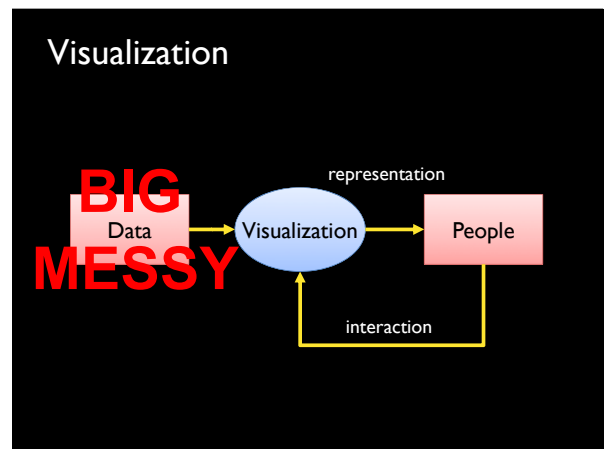
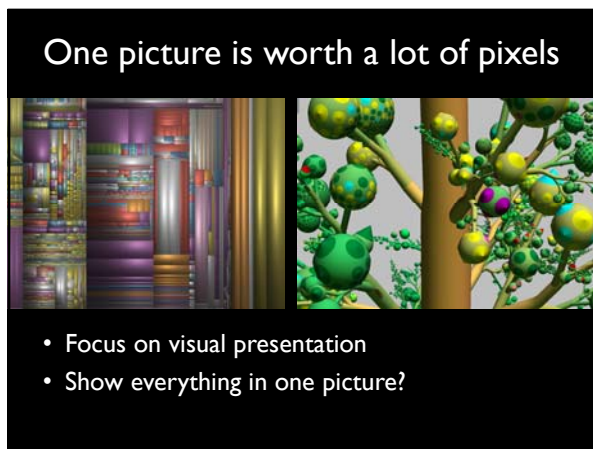
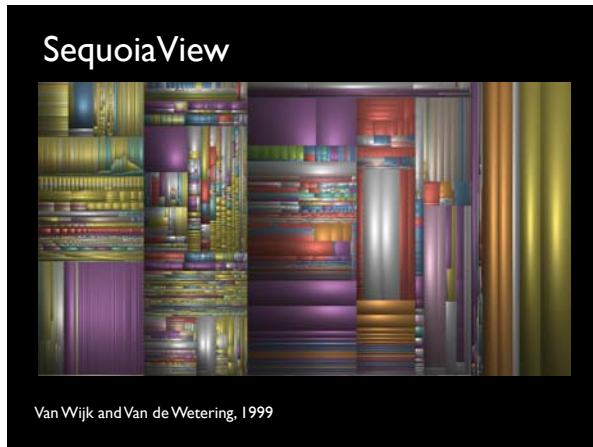
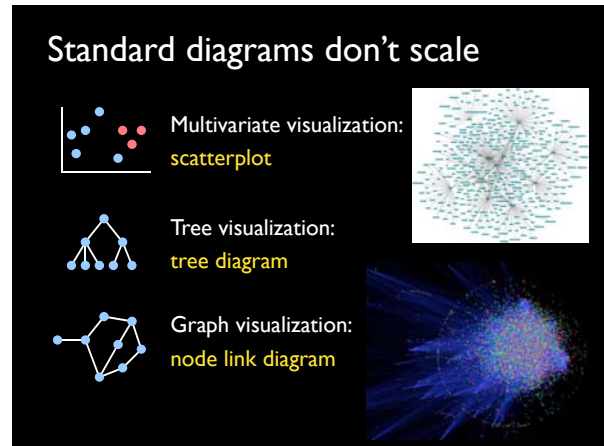
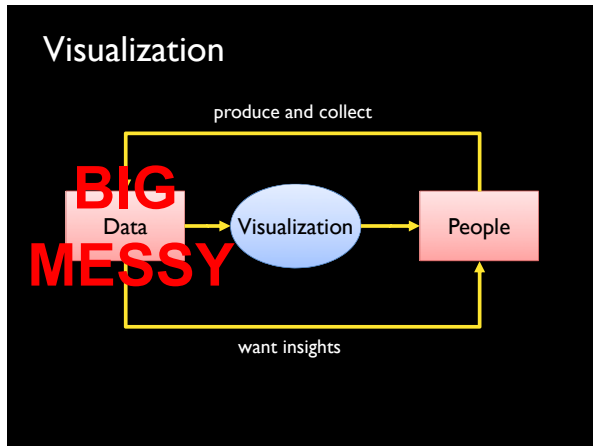


**Anscombe's quartet**

I		II		III		IV	
$x_1$	$y_1$	$x_2$	$y_2$	$x_3$	$y_3$	$x_4$	$y_4$
10.0	8.04	10.0	9.14	10.0	7.46	8.0	6.58
8.0	6.95	8.0	8.14	8.0	6.77	8.0	5.76
13.0	7.58	13.0	8.74	13.0	12.74	8.0	7.71
9.0	8.81	9.0	8.77	9.0	7.11	8.0	8.84
11.0	8.33	11.0	9.26	11.0	7.81	8.0	8.47
14.0	9.96	14.0	8.10	14.0	8.84	8.0	7.04
6.0	7.24	6.0	6.13	6.0	6.08	8.0	5.25
4.0	4.26	4.0	3.10	4.0	5.39	19.0	12.50
12.0	10.84	12.0	9.13	12.0	8.15	8.0	5.56
7.0	4.82	7.0	7.26	7.0	6.42	8.0	7.91
5.0	5.68	5.0	4.74	5.0	5.73	8.0	6.89

Four data-sets: same average  $x$  and  $y$ , same variance, same correlation, ...





### Information Visualization

- The use of computer-supported, **interactive**, visual representations of abstract data to amplify cognition (Card et al., 1999)

```

    graph LR
      Data[Data] --> Visualization((Visualization))
      Visualization --> User[User]
      User --> Visualization
  
```

### InfoGraphic ≠ InfoVis

The Guardian, 2010

### Information Visualization

- The use of computer-supported, interactive, visual representations of abstract data to amplify cognition (Card et al., 1999)
- Visualization: exploration process, not a picture.

```

    graph LR
      Data[Data] --> Visualization((Visualization))
      Visualization --> User[User]
      User --> Visualization
  
```

### Visualization

**BIG MESSY**

```

    graph LR
      Data[Data] -- representation --> Visualization((Visualization))
      Visualization --> People[People]
      People -- interaction --> Visualization
  
```

### Statistics, machine learning, data mining, ...

**BIG MESSY**

```

    graph LR
      Data[Data] --> Model((Model))
      Model --> People[People]
  
```

### Visual Analytics

```

    graph LR
      Data[Data] --> Model((Model))
      Model --> People[People]
      People -- steer model --> Model
      Data --> Visualization((Visualization))
      Visualization --> People
      People -- control --> Visualization
  
```

Model: statistics, machine learning, data mining, ...

## Visual Analytics

### Computers

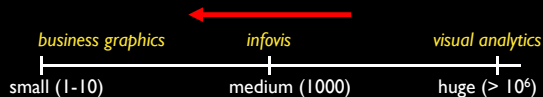
- fast
- precise
- computation
- search and store data
- graphics

### Humans

- Flexible, inventive
- Solve problems
- Handle new situations
- Handle incomplete and/or inconsistent information
- Have domain knowledge and experience
- Can **see** things that are hard to compute

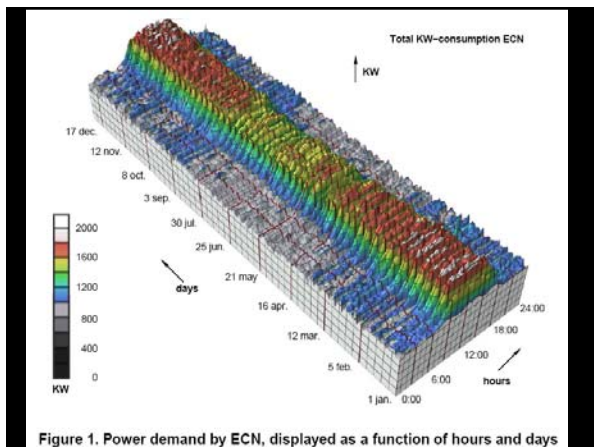
Visual Analytics: Synergy between computers and humans

## Data size

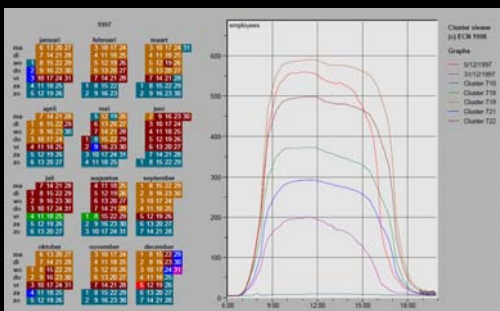


Try to move to the left:

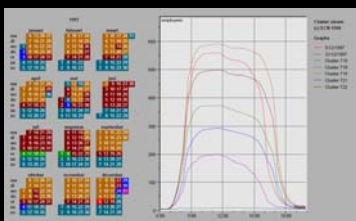
- Filter, aggregate, statistics, machine learning, ... without losing essential information



## Visual Analytics Time Series Data



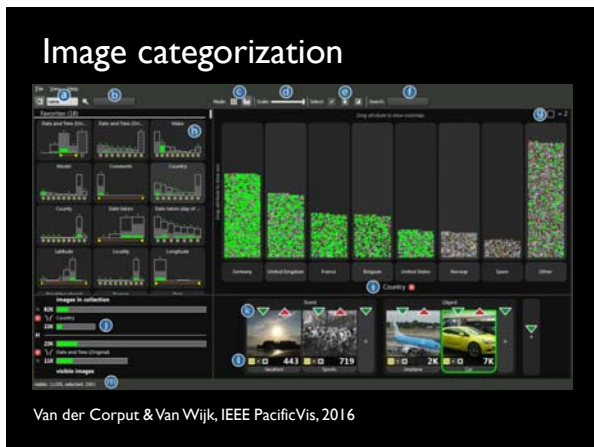
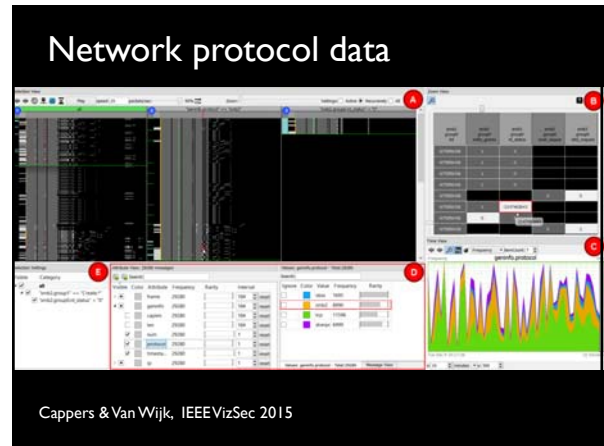
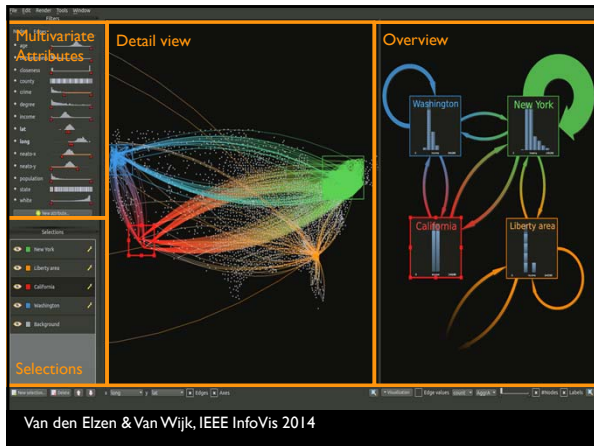
## The pattern



- combination of simple presentations,
- use interaction and statistics

## Challenges

- Handling big, messy data
- Dealing with models
- Examples from Eindhoven
- Lessons learned, challenges ahead



Dealing with big, messy data

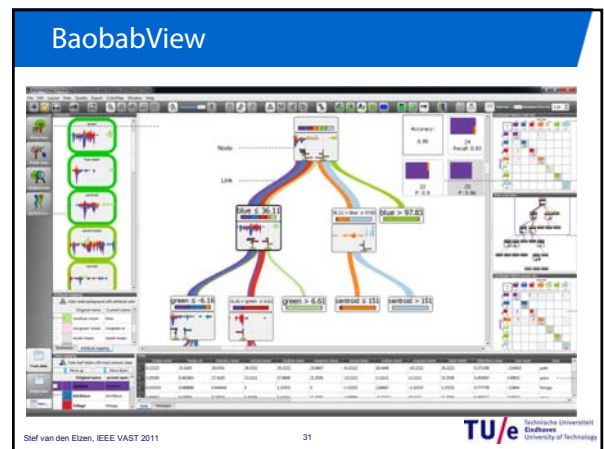
Custom cases can be solved, using a combination of simple presentations, interaction and statistics

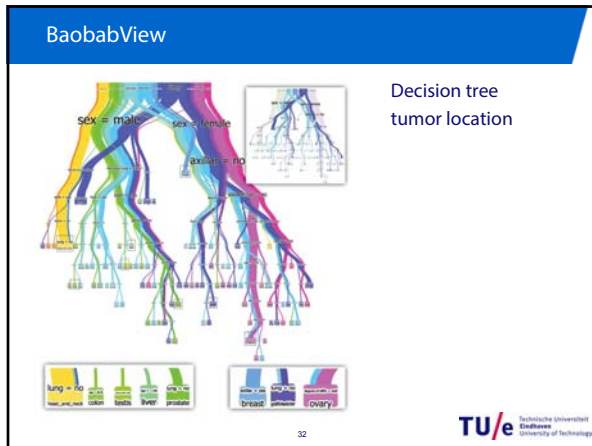
Generic challenges:

- How to scale up?
- How to generalize?

Dealing with models

- How to define ML/DM models?
  - What model, what parameter settings?
  - Which features of the data?
- How to understand the result?
- *How to make the black box transparent?*





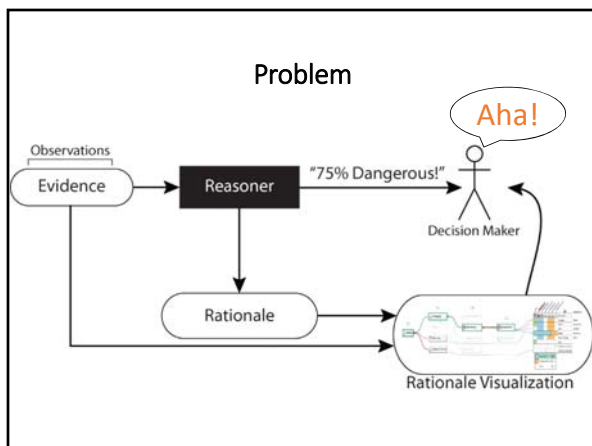
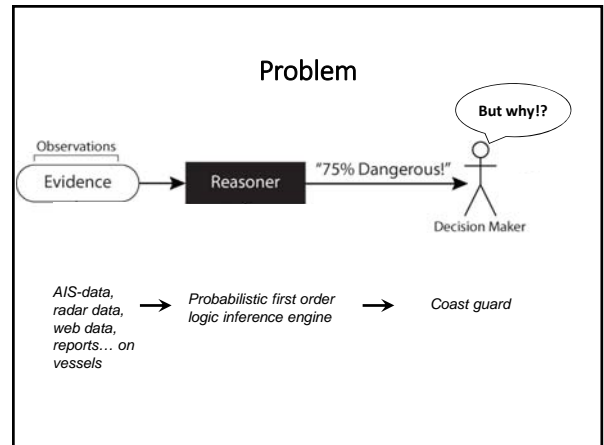
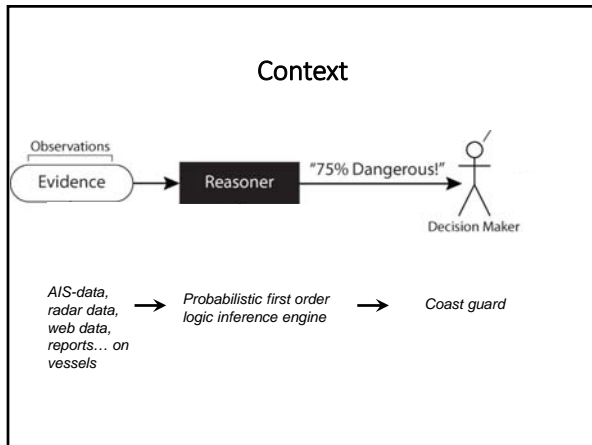
**EuroVis 2015**  
The EG / VGTC  
Conference on  
Visualization

25 - 29  
May  
Cagliari / Sardinia / Italy

## Rationale Visualization for Safety and Security

Roeland Scheepens  
Steffen Michels  
Huub van de Wetering  
Jarke J. van Wijk

TU/e Technische Universiteit  
Eindhoven  
University of Technology



### Example

The reasoner may be confused because it may be a smuggler, is not a pilot or a fug, and has collided in the past.

The reasoner may be a smuggler because it is trying to hide its identity.

The reasoner is trying to find its identity by updating the clarity of its search path.

CONMIT / A public-private research community

THALES Embedded Systems Innovation

BY TNO TU/e Technische Universiteit Eindhoven University of Technology

## How to understand a model?

Custom cases can be solved, using a combination of simple presentations, interaction and statistics

Challenges:

- How to scale?
- How to generalize?
- Can we understand neural networks...?

## Conclusions

Data and models can be made transparent, using a combination of simple presentations, interaction and statistics

*and with smart students and a lot of effort!*

Challenges:

- How to scale?
- How to generalize?
- *How to develop custom solutions efficiently?*

Thank you!