

Visualizing Data: Perception, Interaction, Computation

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LMU Munich

Diplom in Computer Science, 2001-2007



BMW Group Research & Technology

PhD in Computer Science, 2007-2010



Univ. of British Columbia, Canada

PostDoc, 2010-2012



Univ. of Vienna, Austria

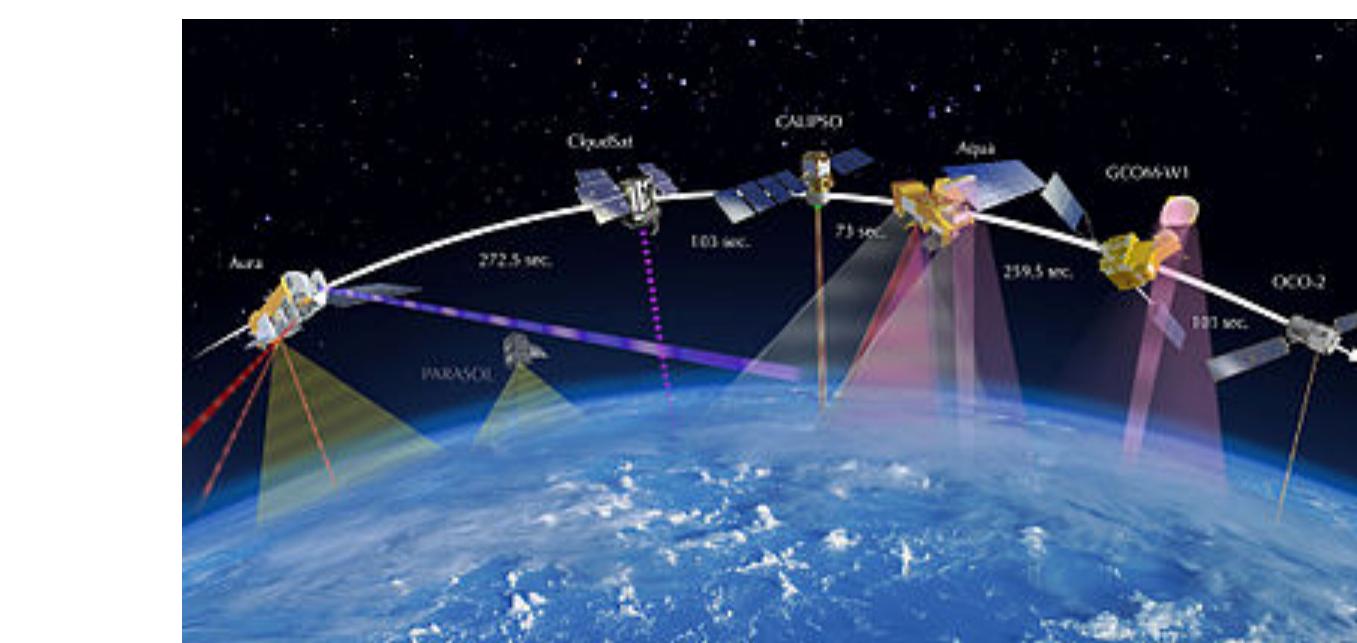
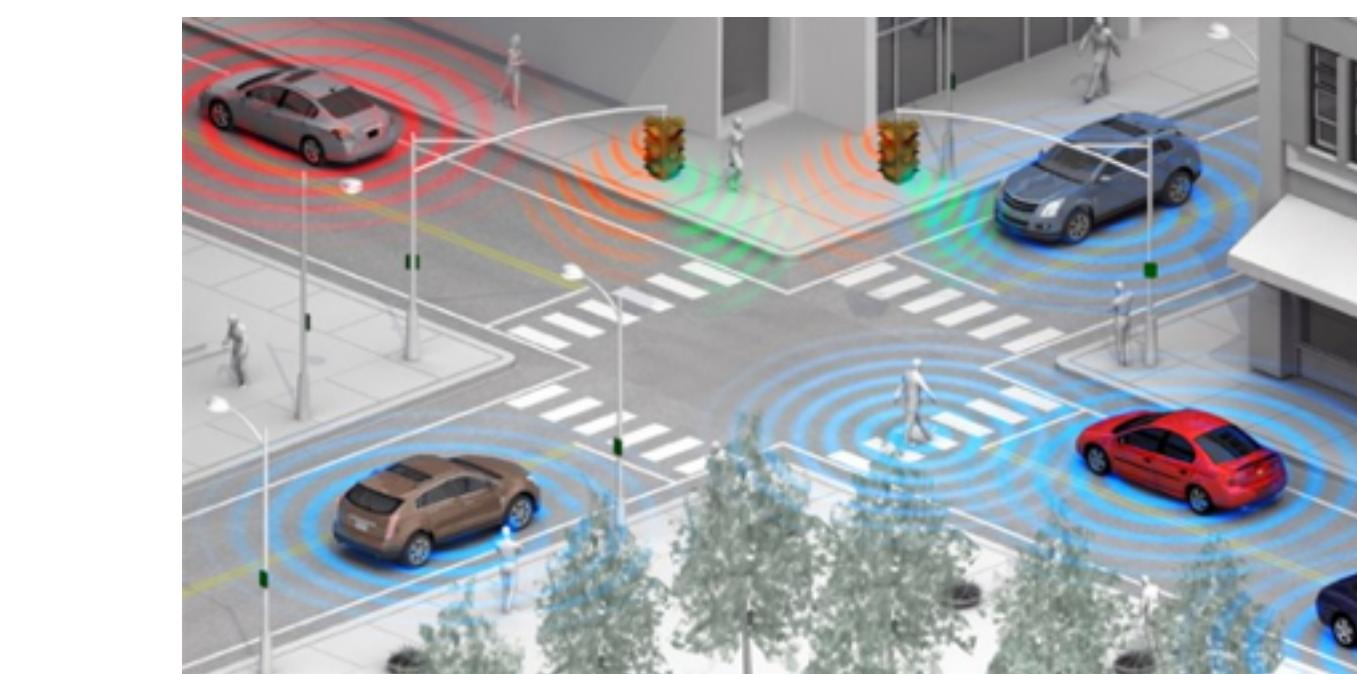
Assistant Professor, 2013-now

Human-Computer
Interaction (HCI)

Information
Visualization

Data
Science
(algorithmic
analysis)

Data



https://en.wikipedia.org/wiki/Earth_observation_satellite

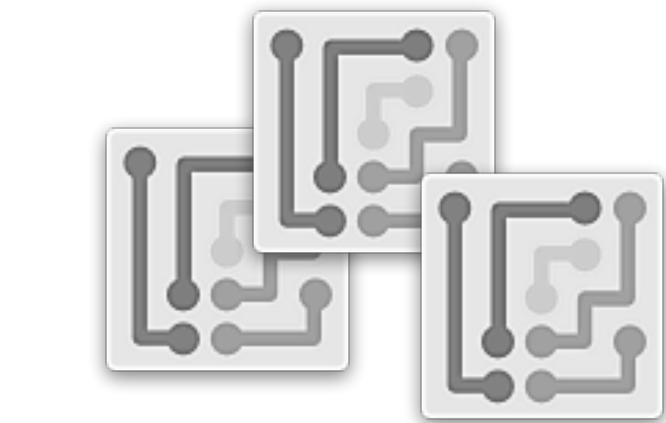
<https://www.extremetech.com/extreme/226739-how-smart-cities-will-work>

Big Data/ Data Science

Data

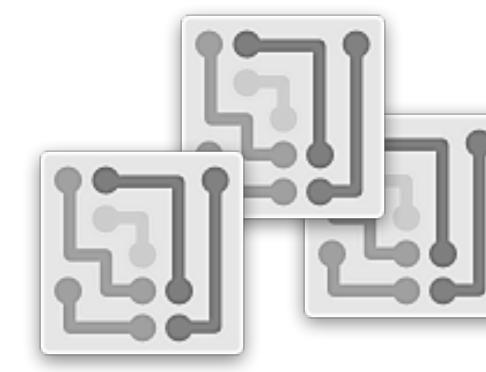


Algorithms



Visualization

Scientists
Engineers
Journalists
Policy Makers



Understand patterns

Gain insights

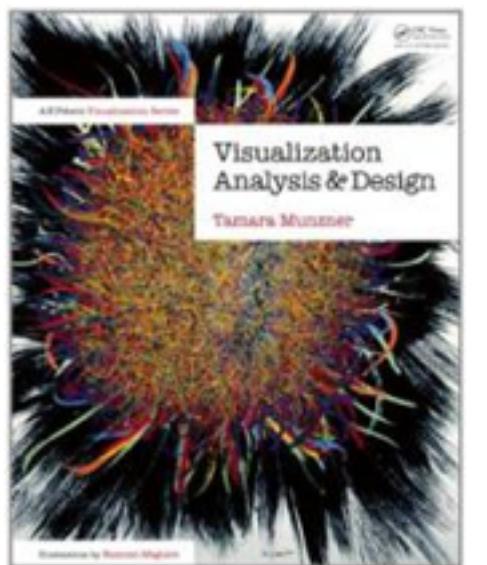
Make decisions

Communicate

Definition

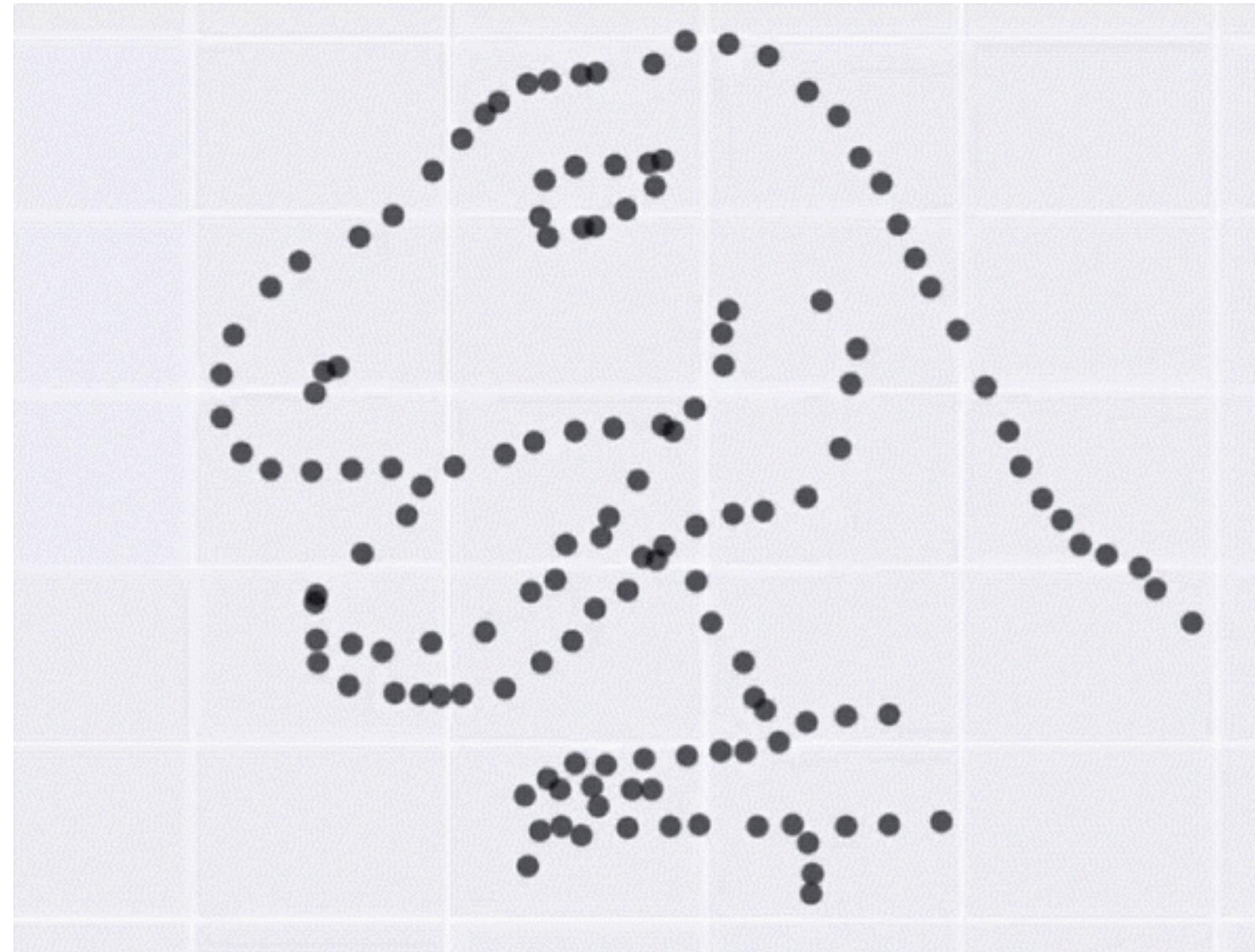
Computer-based visualization systems provide visual representations of datasets intended to help people carry out some task more effectively.

Munzner, 2014



Definition

Computer-based visualization systems provide **visual representations** of datasets intended to help people carry out some task more effectively.



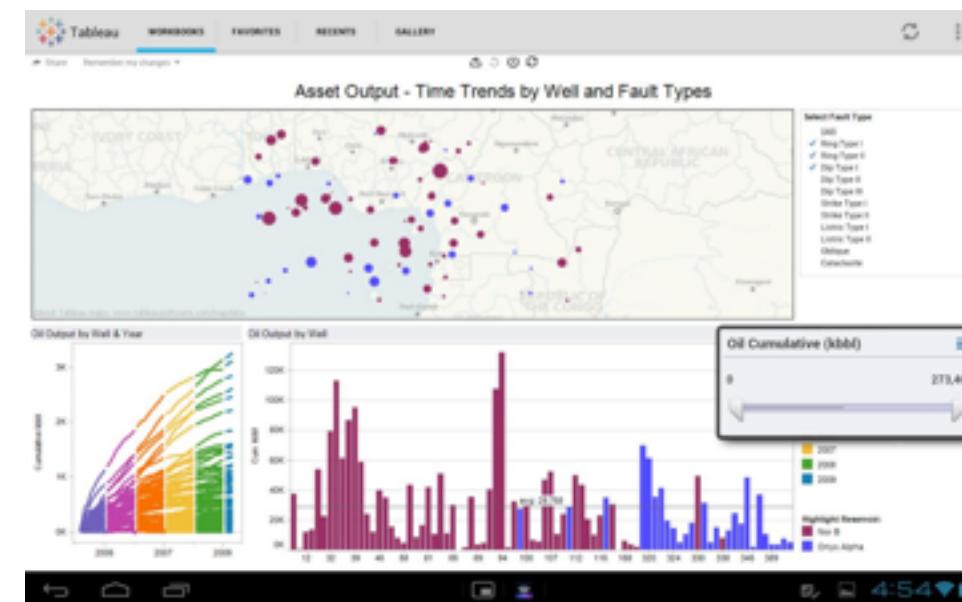
Same stats

x mean	54
y mean	48
x SD	17
y SD	27
x/y correlation	-0.1

Matejka, Fitzmaurice.
Same Stats, Different Graphs: Generating
Datasets with Varied
ACM CHI, 2017.

Definition

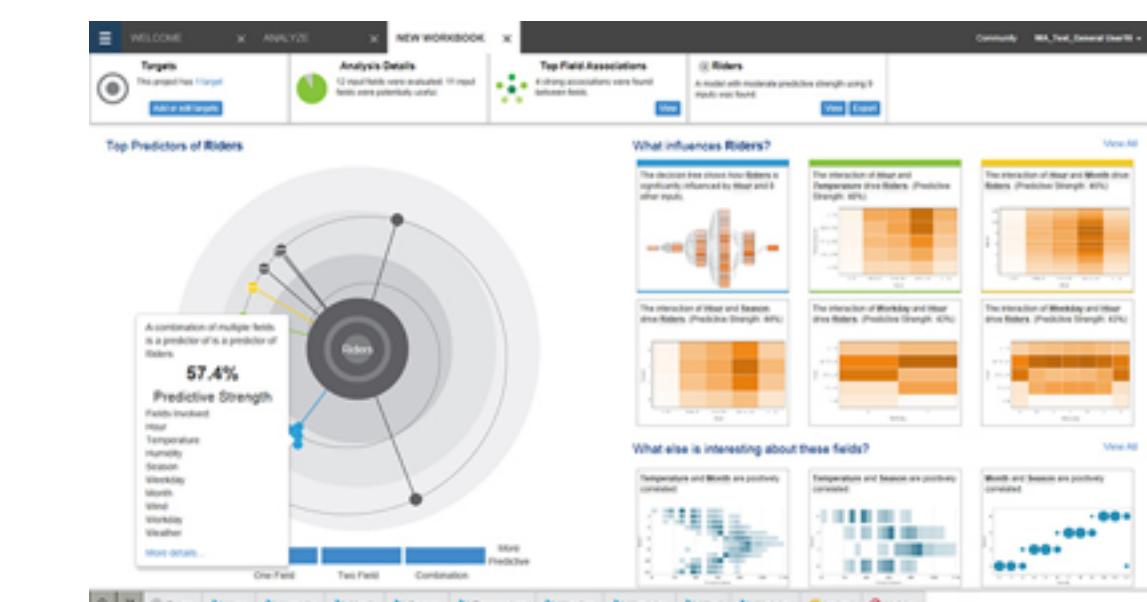
Computer-based visualization systems provide visual representations of datasets intended to help people carry out some task more effectively.



Tableau



SAS



IBM Watson Analytics



Improvise

Definition

Computer-based visualization systems provide visual representations of datasets intended to help **people carry out some task** more effectively.



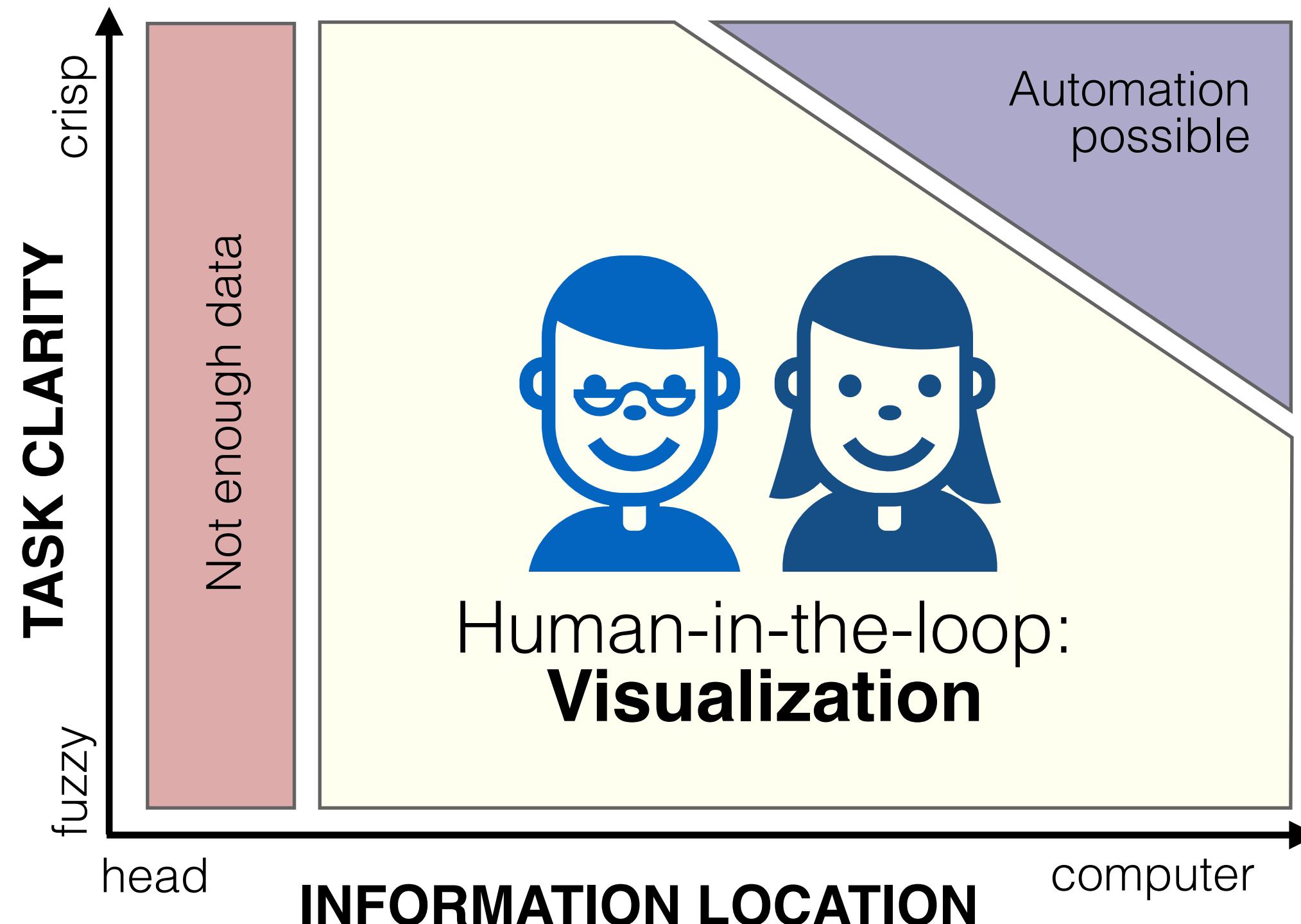
... not necessary if task can be fully automated

Definition

Computer-based visualization systems provide visual representations of datasets intended to help **people carry out some task** more effectively.

**Wait, why humans in the loop?
We have great algorithms!**

Why human-in-the-loop?



Examples for “automation possible”:

- handwritten text recognition
- autonomous driving
- ...

Examples for ill-defined problems:

- exploratory analysis of scientific problems
- (collaborative) decision-making problems
- model building & validation
- ...

Michael Sedlmair, Miriah Meyer, Tamara Munzner
Design Study Methodology: Reflections from the Trenches and the Stacks (IEEE InfoVis 2012).

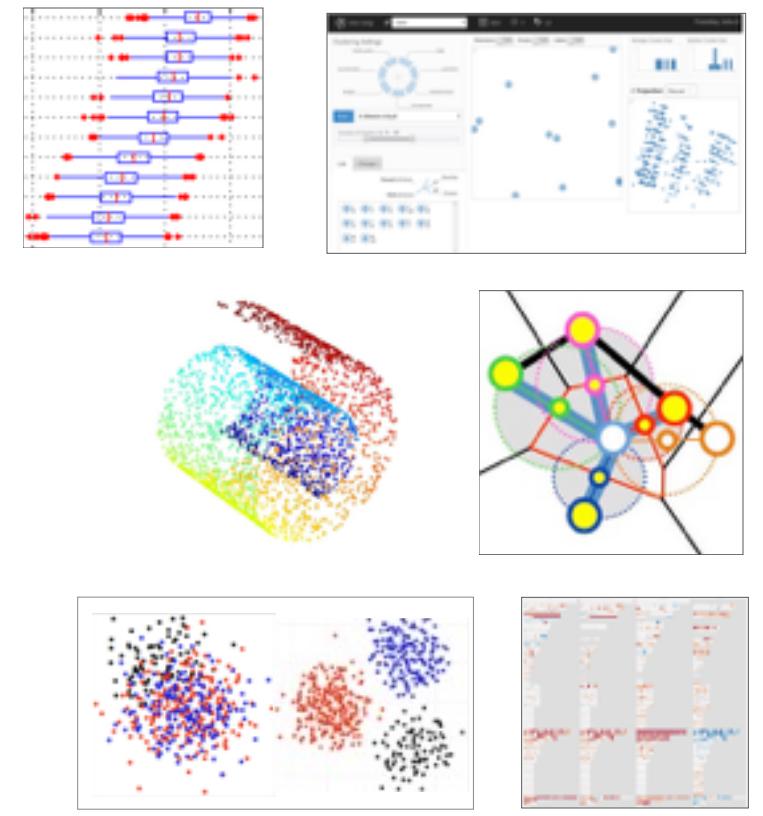
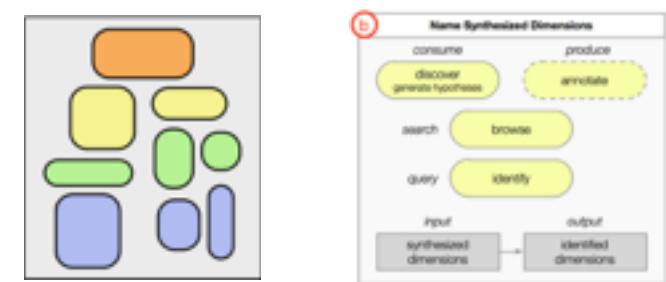
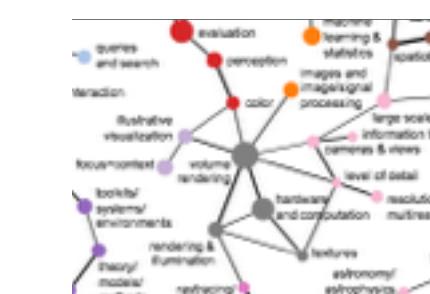
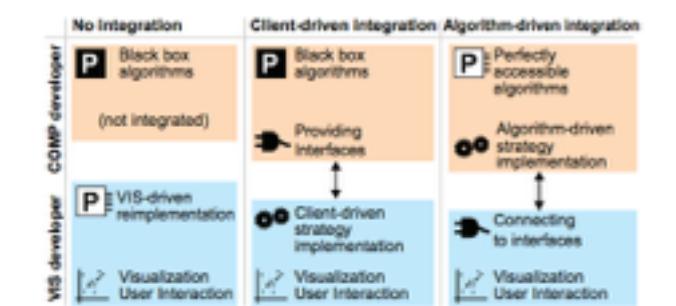
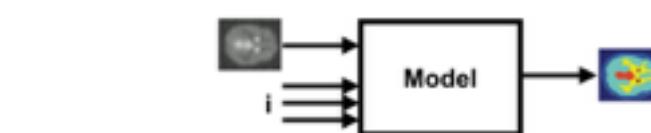
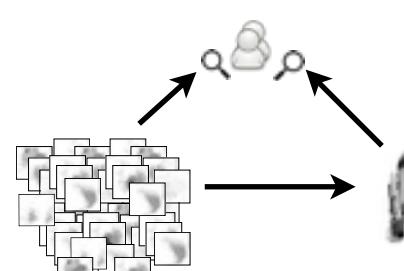
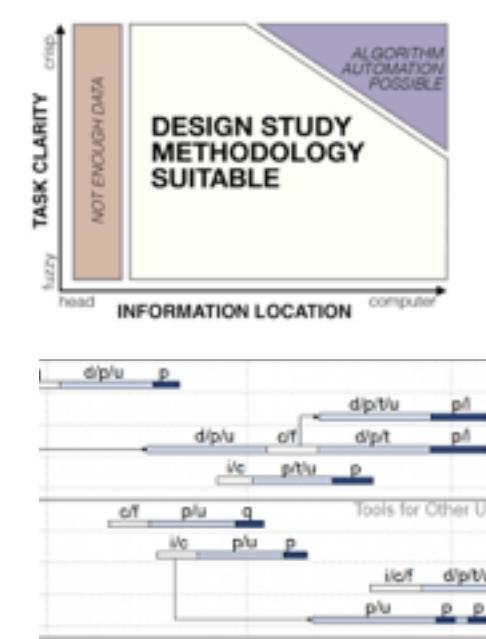


problem-driven

MY WORK

technique-driven

methods & theory



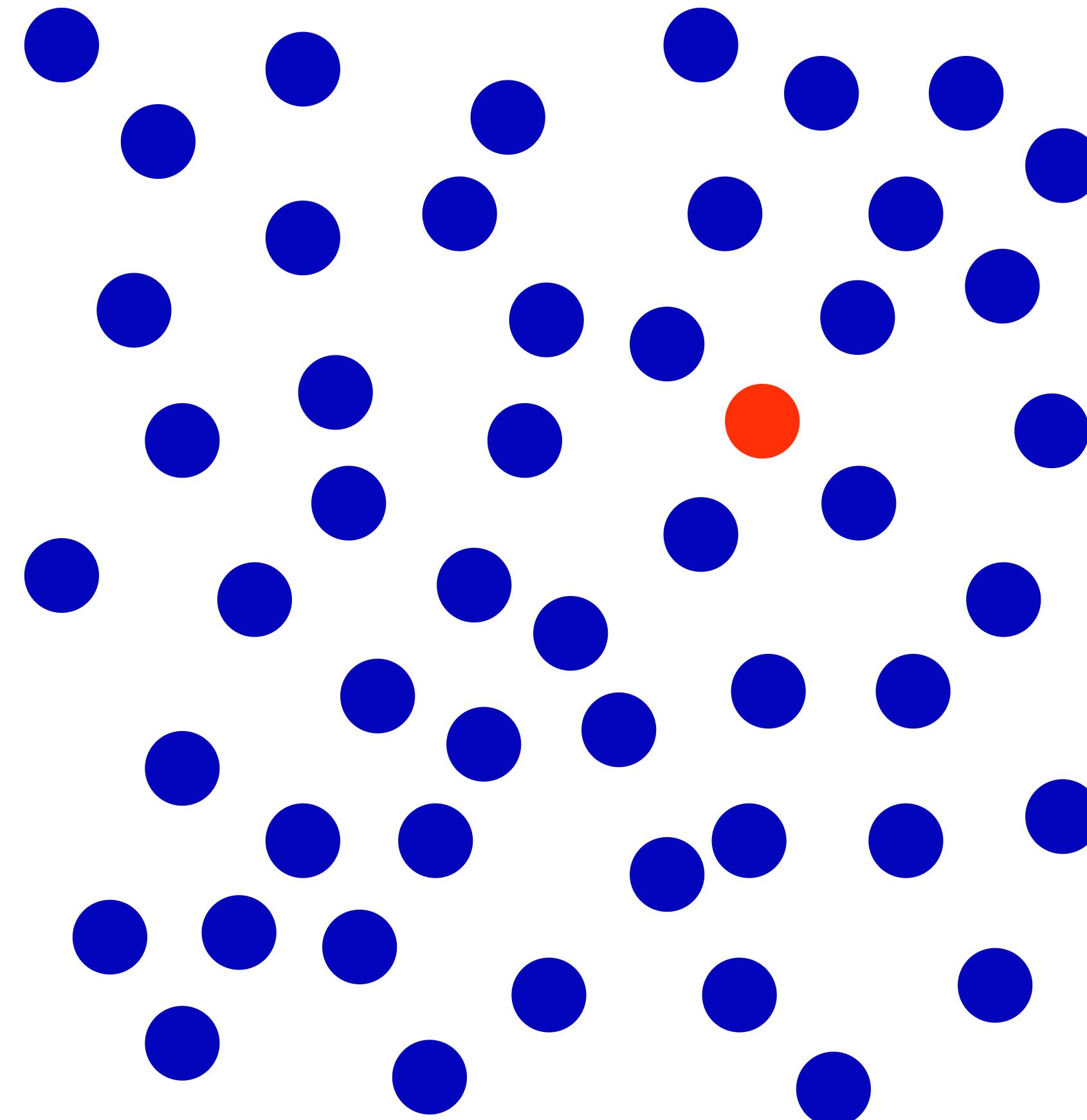
Outline: How to visualize data?

- Leverage perception
- Make it interactive
- Combine computation and visualization

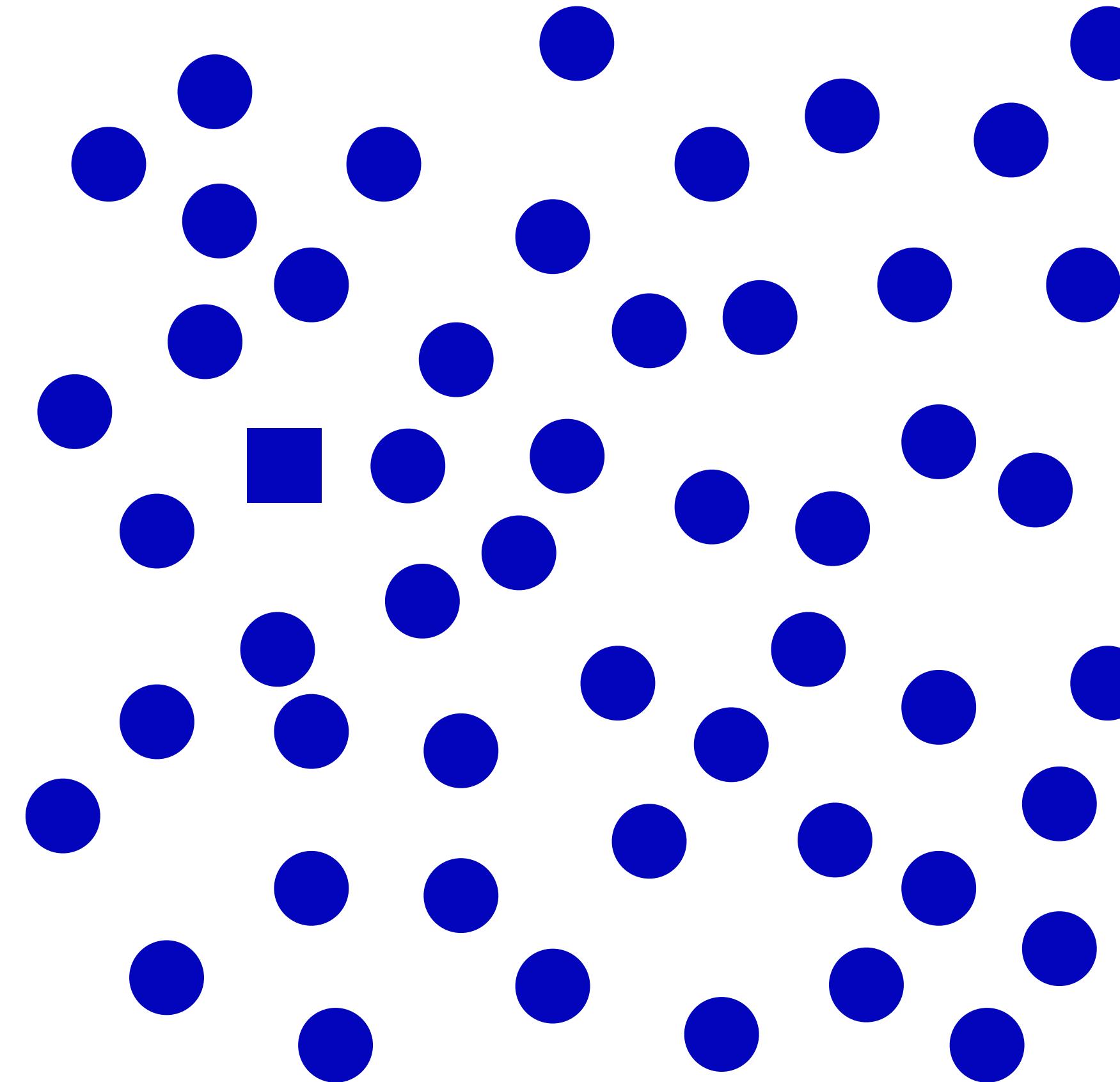
Leverage perception

...

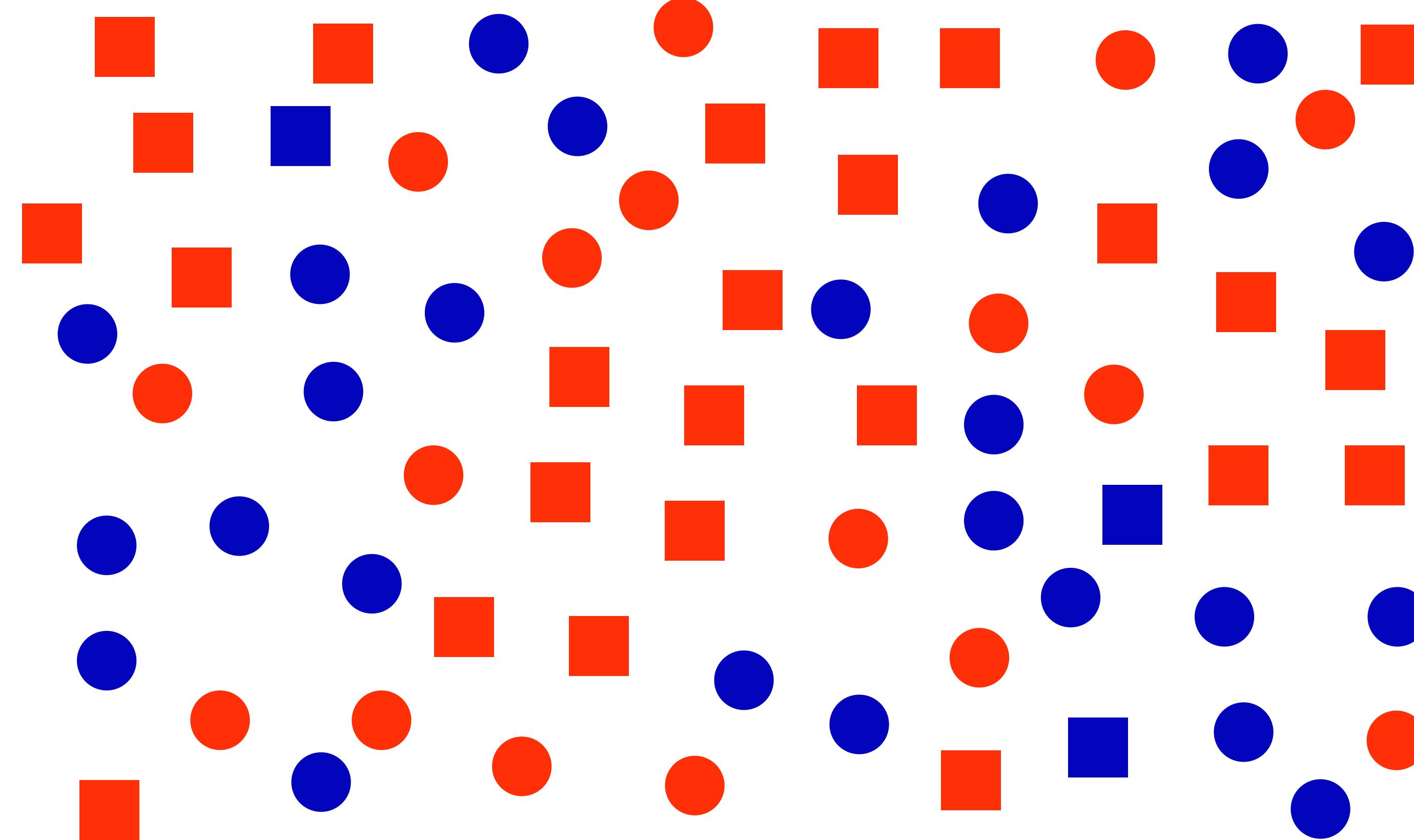
Human perception: Popout



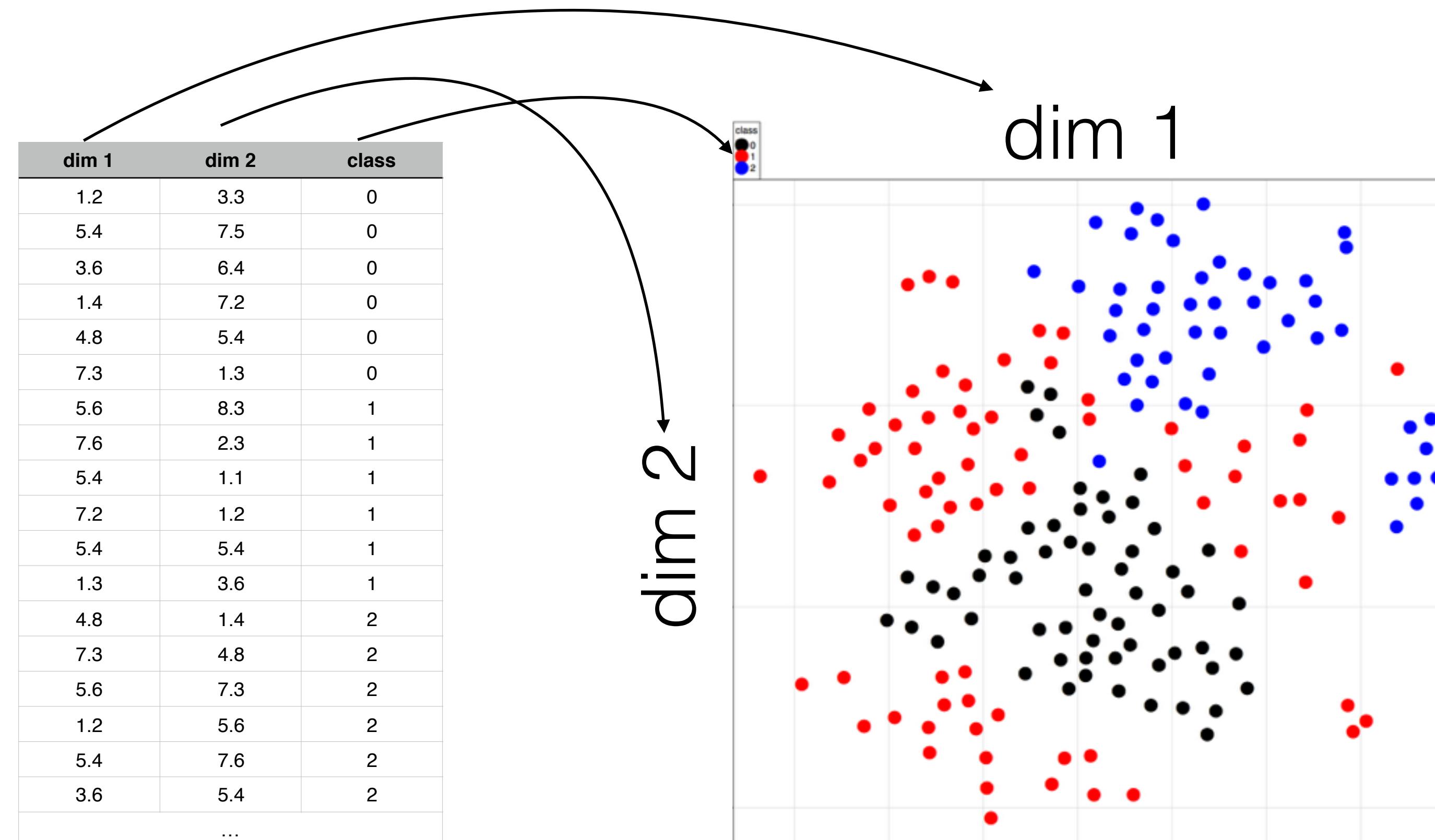
Human perception: Popout



Popout: Combining channels?



2D Scatterplot

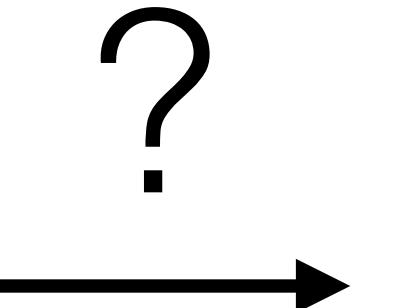


data

visualization

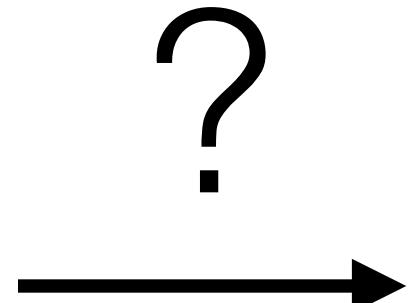
3D Scatterplot?

dim 1	dim 2	dim 3	class
1.2	3.3	5.4	0
5.4	7.5	1.3	0
3.6	6.4	4.8	0
1.4	7.2	7.3	0
4.8	5.4	5.6	0
7.3	1.3	1.2	0
5.6	8.3	5.4	1
7.6	2.3	3.6	1
5.4	1.1	3.3	1
7.2	1.2	7.5	1
5.4	5.4	6.4	1
1.3	3.6	7.2	1
4.8	1.4	5.4	2
7.3	4.8	1.3	2
5.6	7.3	8.3	2
1.2	5.6	3.3	2
5.4	7.6	7.5	2
3.6	5.4	1.2	2
...			

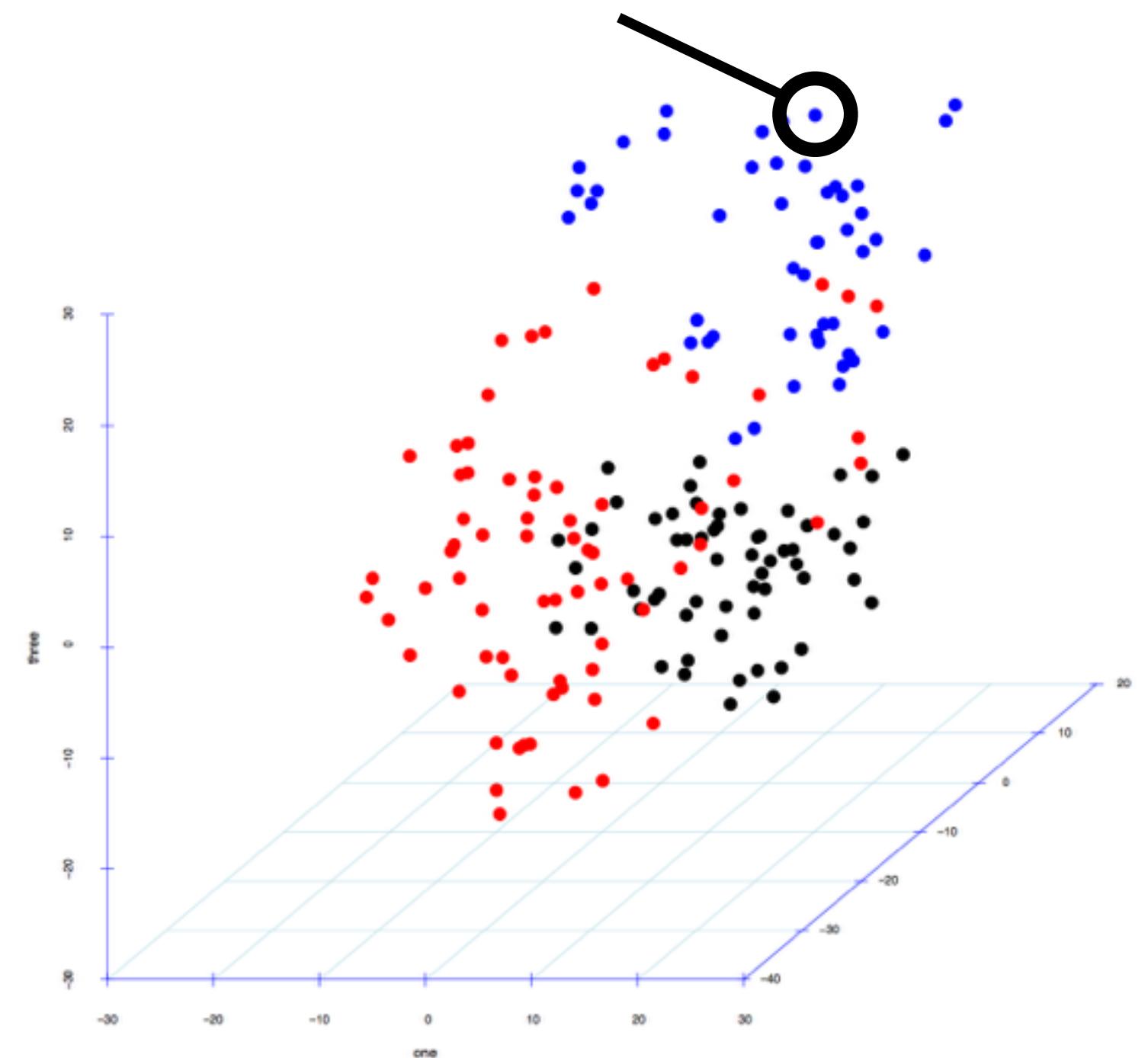


3D Scatterplot?

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7.6	2.3	3.6	1
5.4	1.1	3.3	1
7.2	1.2	7.5	1
5.4	5.4	6.4	1
1.3	3.6	7.2	1
4.8	1.4	5.4	2
7.3	4.8	1.3	2
5.6	7.3	8.3	2
1.2	5.6	3.3	2
5.4	7.6	7.5	2
3.6	5.4	1.2	2
...			



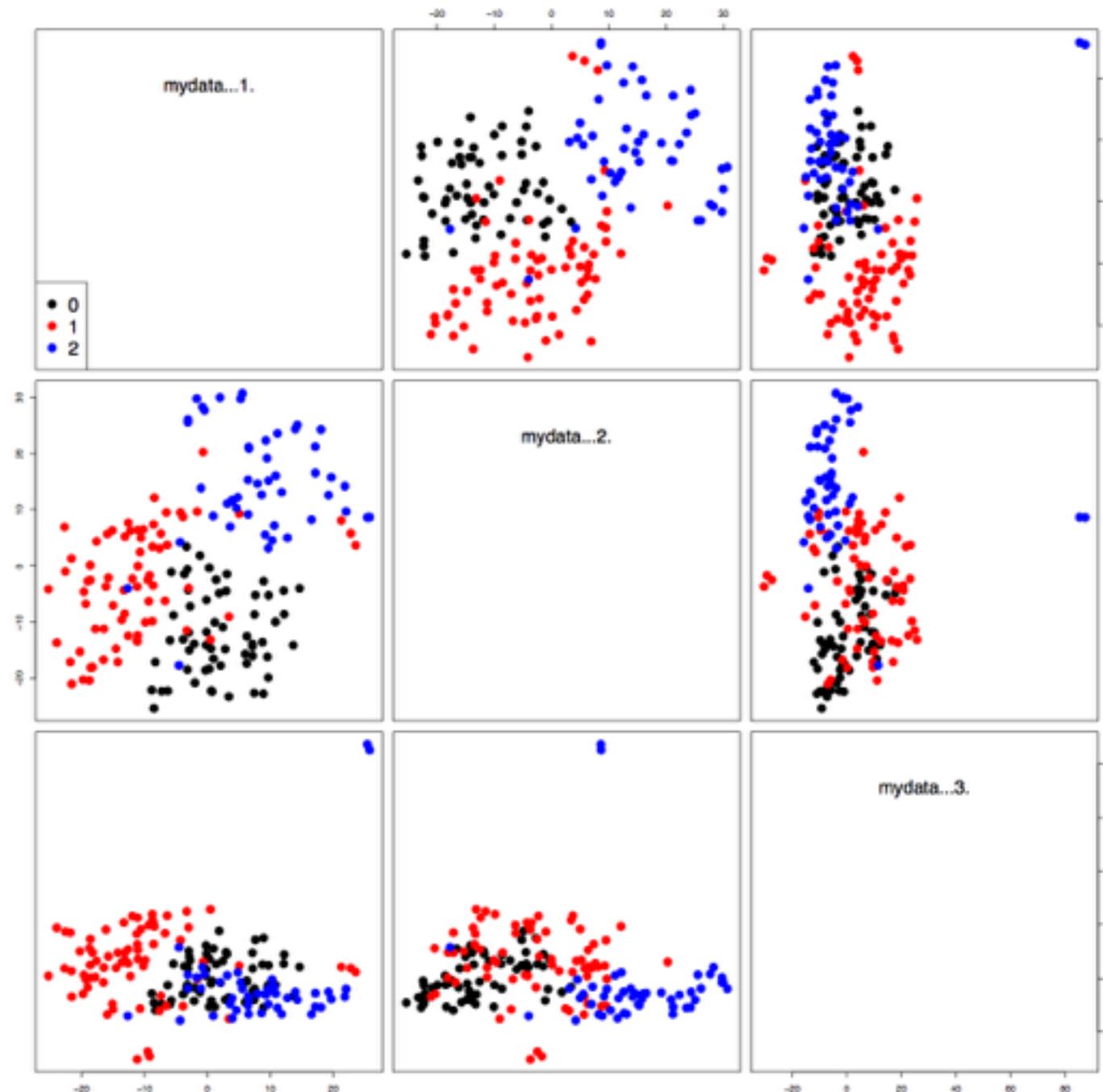
what's the value of this point?



No 3D for abstract data!

Scatterplot Matrix

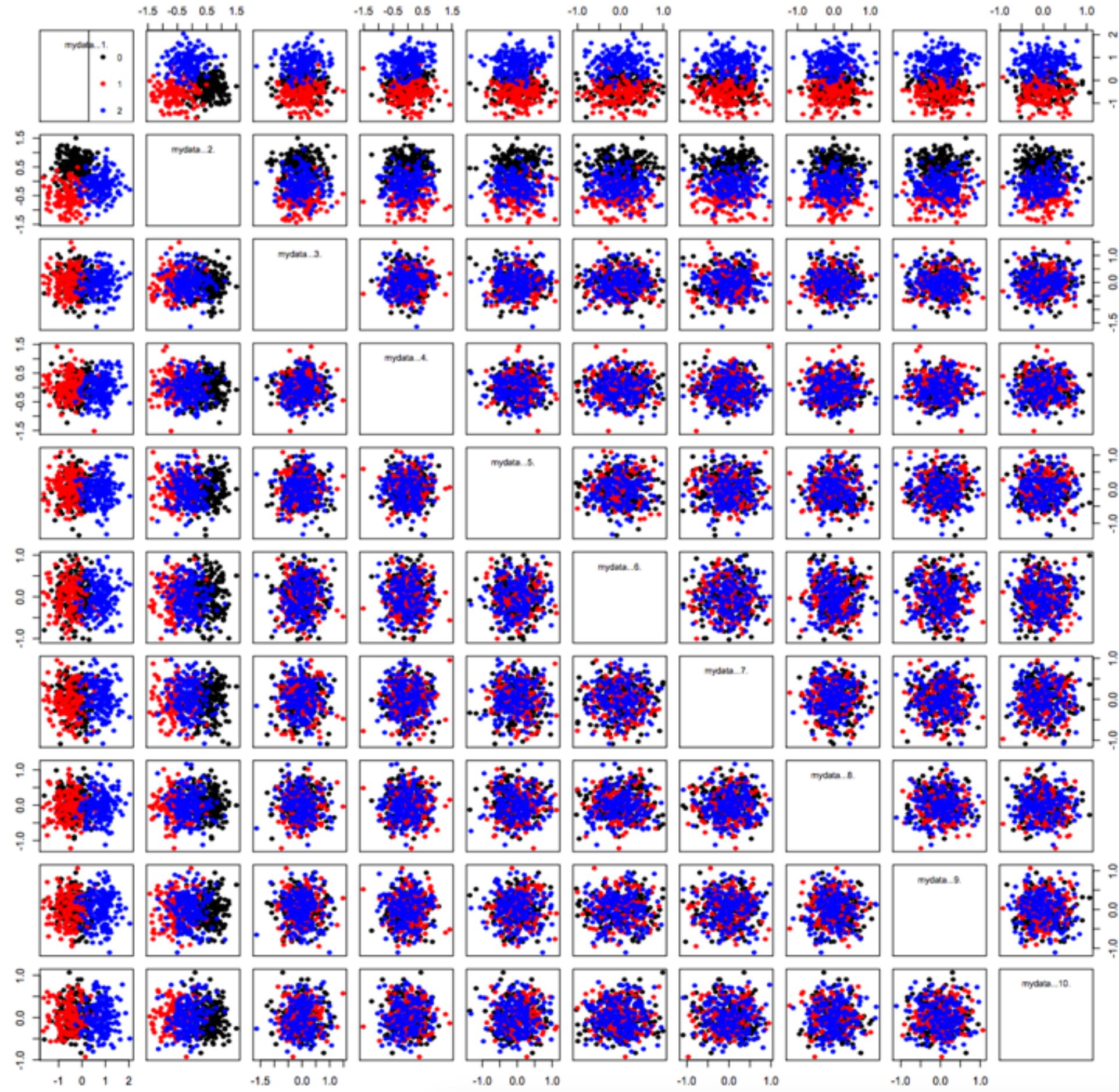
dim 1	dim 2	dim 3	class
1.2	3.3	5.4	0
5.4	7.5	1.3	0
3.6	6.4	4.8	0
1.4	7.2	7.3	0
4.8	5.4	5.6	0
7.3	1.3	1.2	0
5.6	8.3	5.4	1
7.6	2.3	3.6	1
5.4	1.1	3.3	1
7.2	1.2	7.5	1
5.4	5.4	6.4	1
1.3	3.6	7.2	1
4.8	1.4	5.4	2
7.3	4.8	1.3	2
5.6	7.3	8.3	2
1.2	5.6	3.3	2
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3.6	5.4	1.2	2
...			



Sedlmair, Munzner, Tory: Empirical Guidance on Scatterplot and Dimension Reduction Technique Choices (InfoVis 2013).



Scatterplot Matrix with more dimensions?

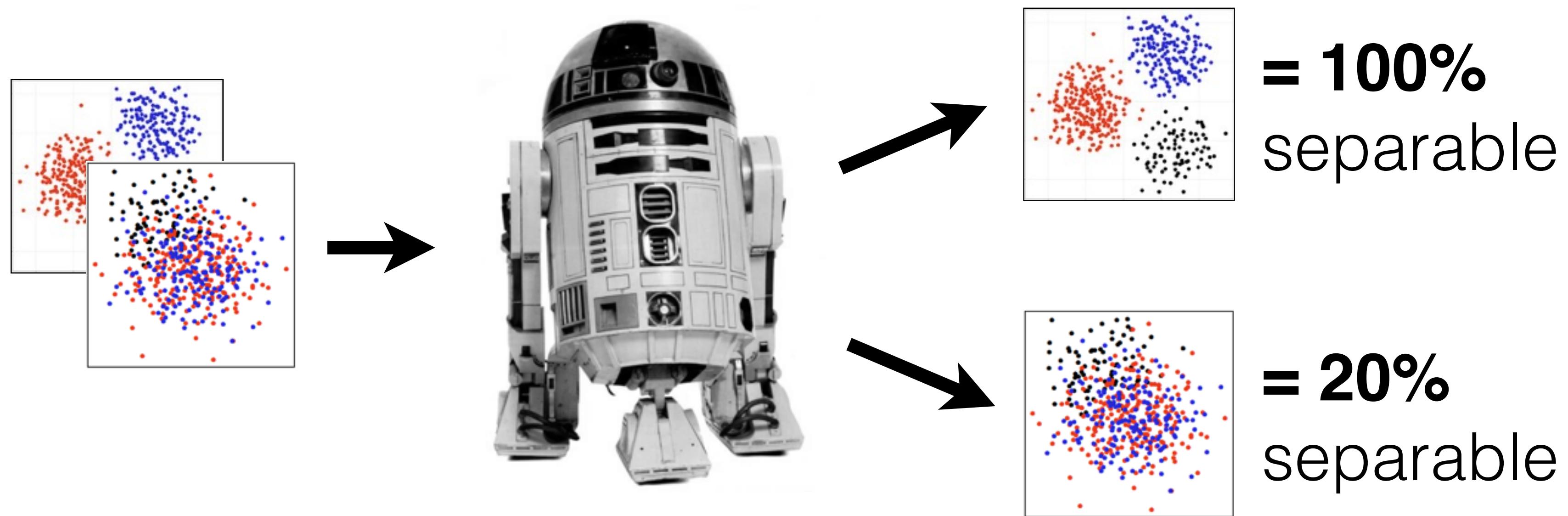


10D → 45 Views

200D → ~20k Views

Modeling human perception

e.g. class separation measures

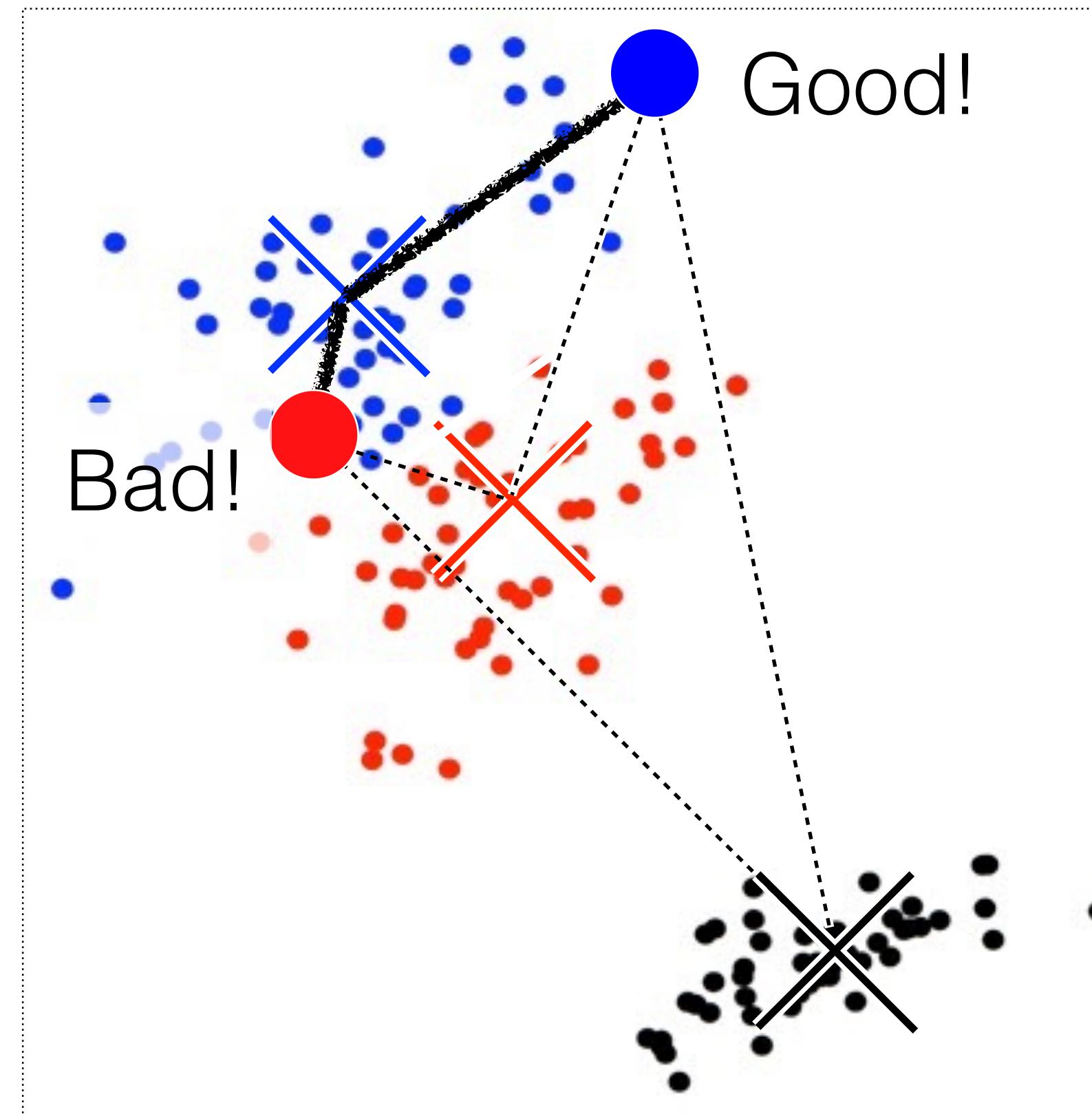


Aupetit, **Sedlmair**: SepMe: 2002 New Visual Separation Measures (PacificVis 2016).

Sedlmair, Aupetit: Data-driven Evaluation of Visual Quality Measures (EuroVis 2015).

Sedlmair, Tatu, Munzner, Tory: A Taxonomy of Visual Cluster Separation Factors (EuroVis 2012).

Example: Centroid measure



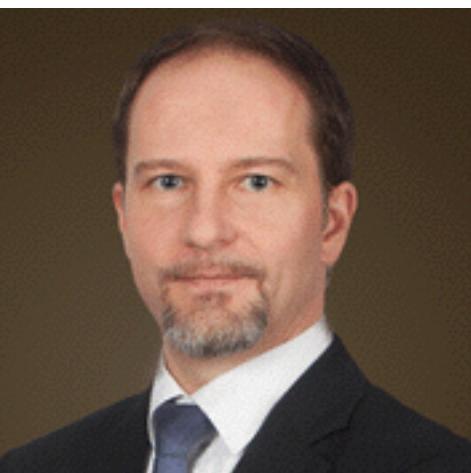
Centroid measure
= **93%** separable

Sips et al.: Selecting good views of high-dimensional data using class consistency (EuroVis 2009).

How good is it?

- Evaluation of 35 measures
- How to evaluate?
 - HCI-style user studies?
 - Using a Computer Vision Framework!

Sedlmair, Aupetit: Data-driven
Evaluation of Visual Quality
Measures (EuroVis 2015).



A “Computer Vision” framework



Freelimages.com

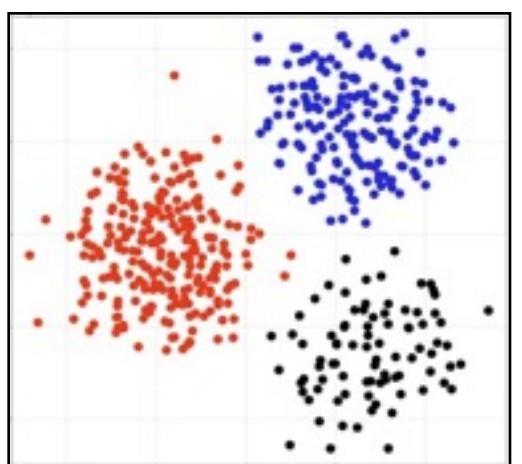
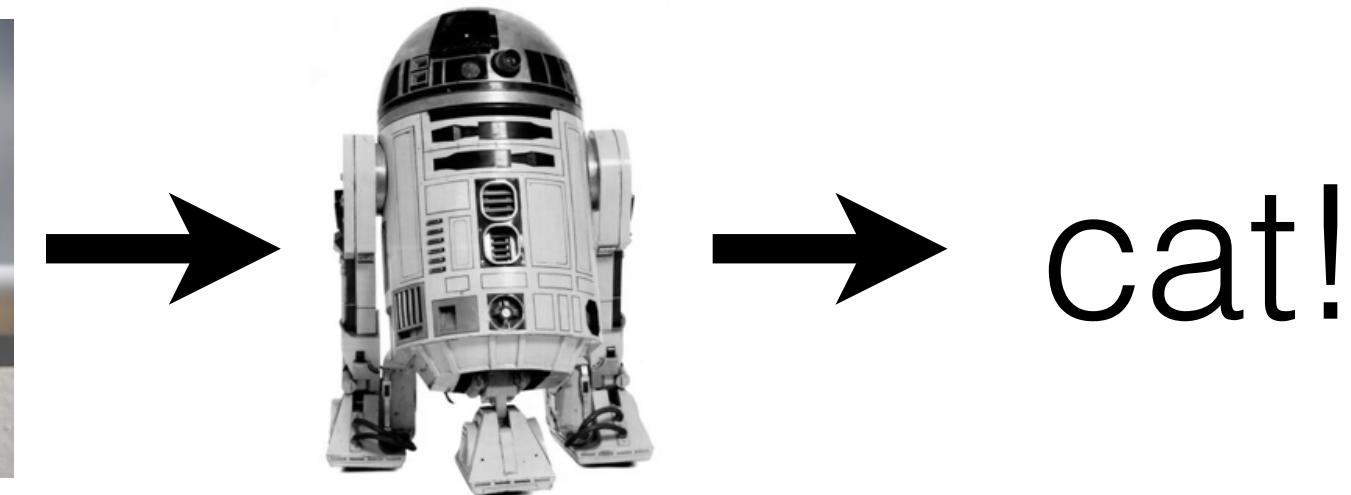
or



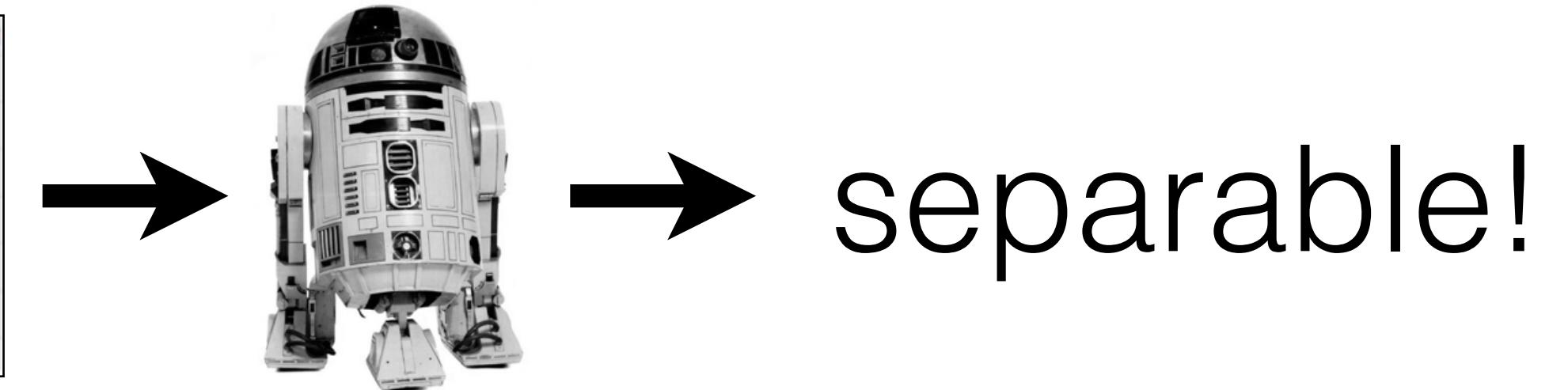
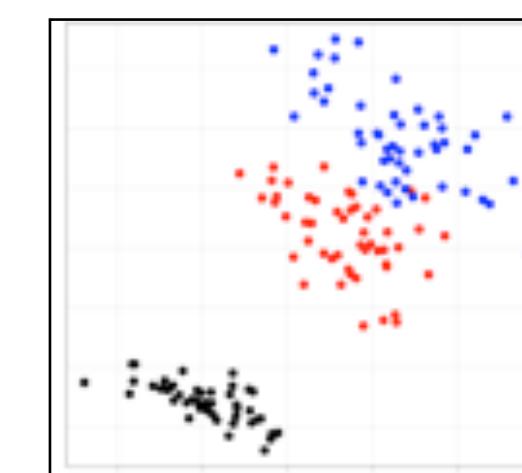
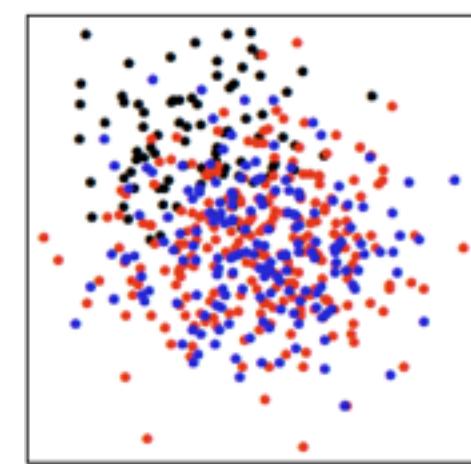
Freelimages.com



Freelimages.com

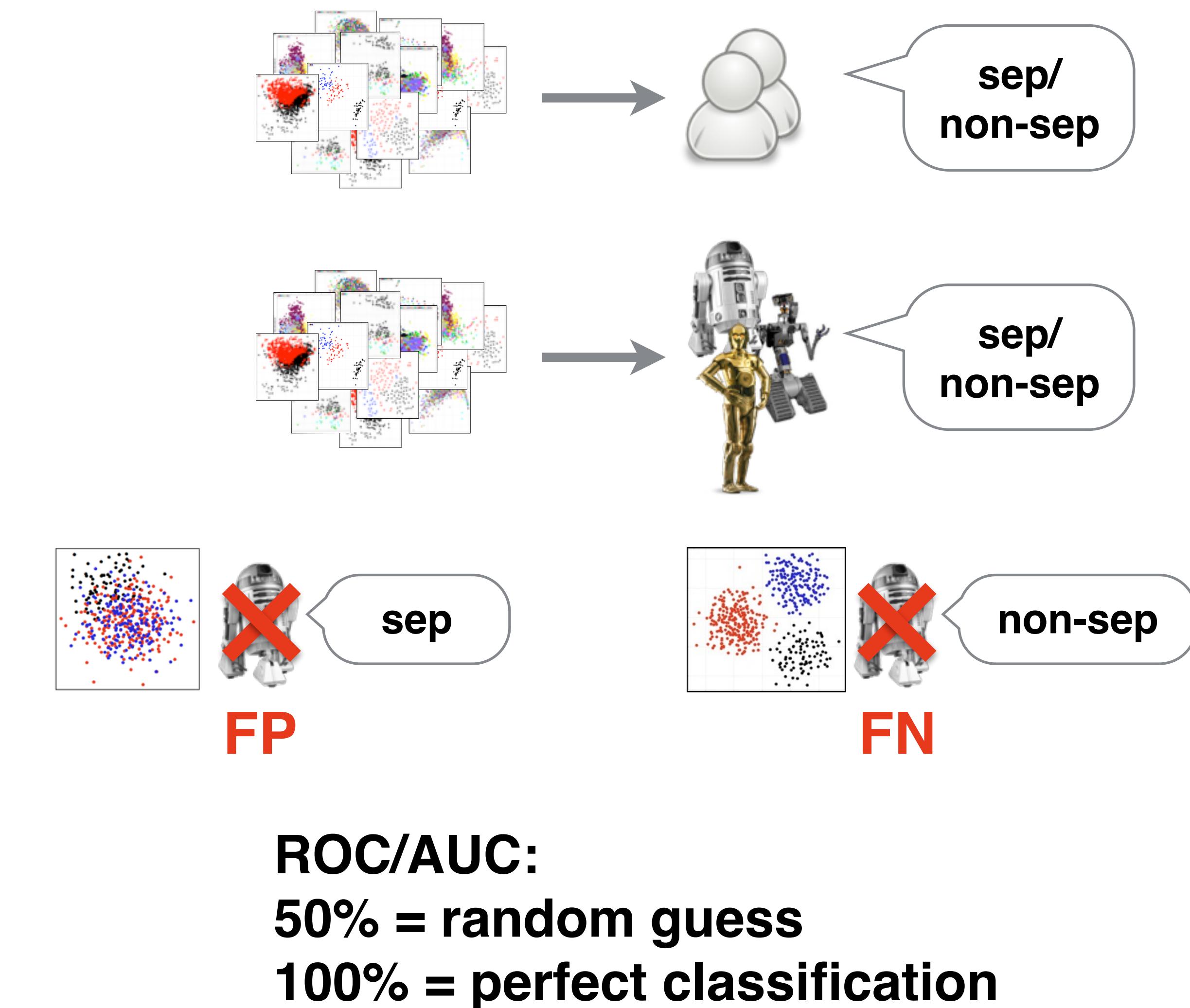


or

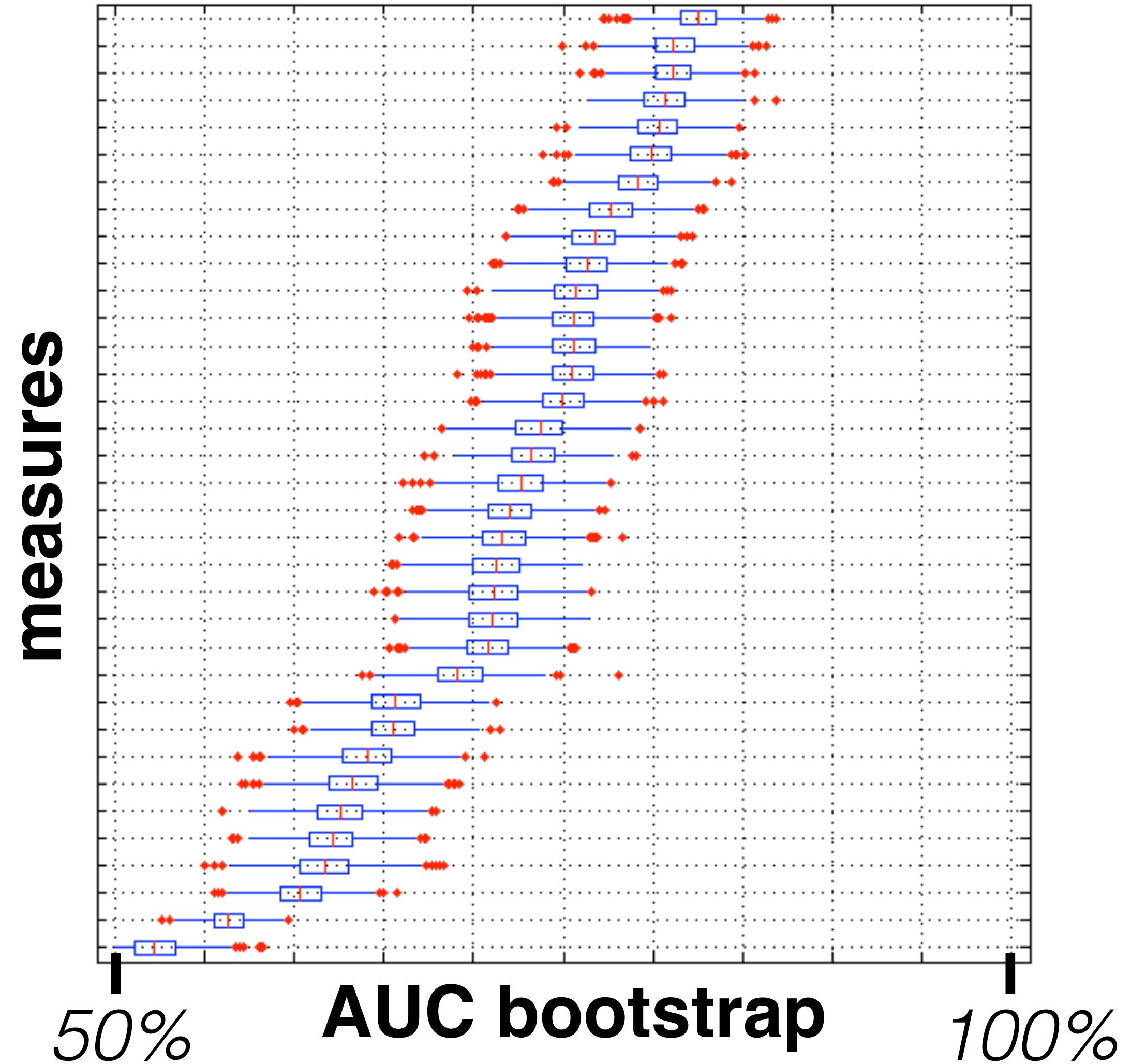


Evaluate with standard Machine Learning pipeline

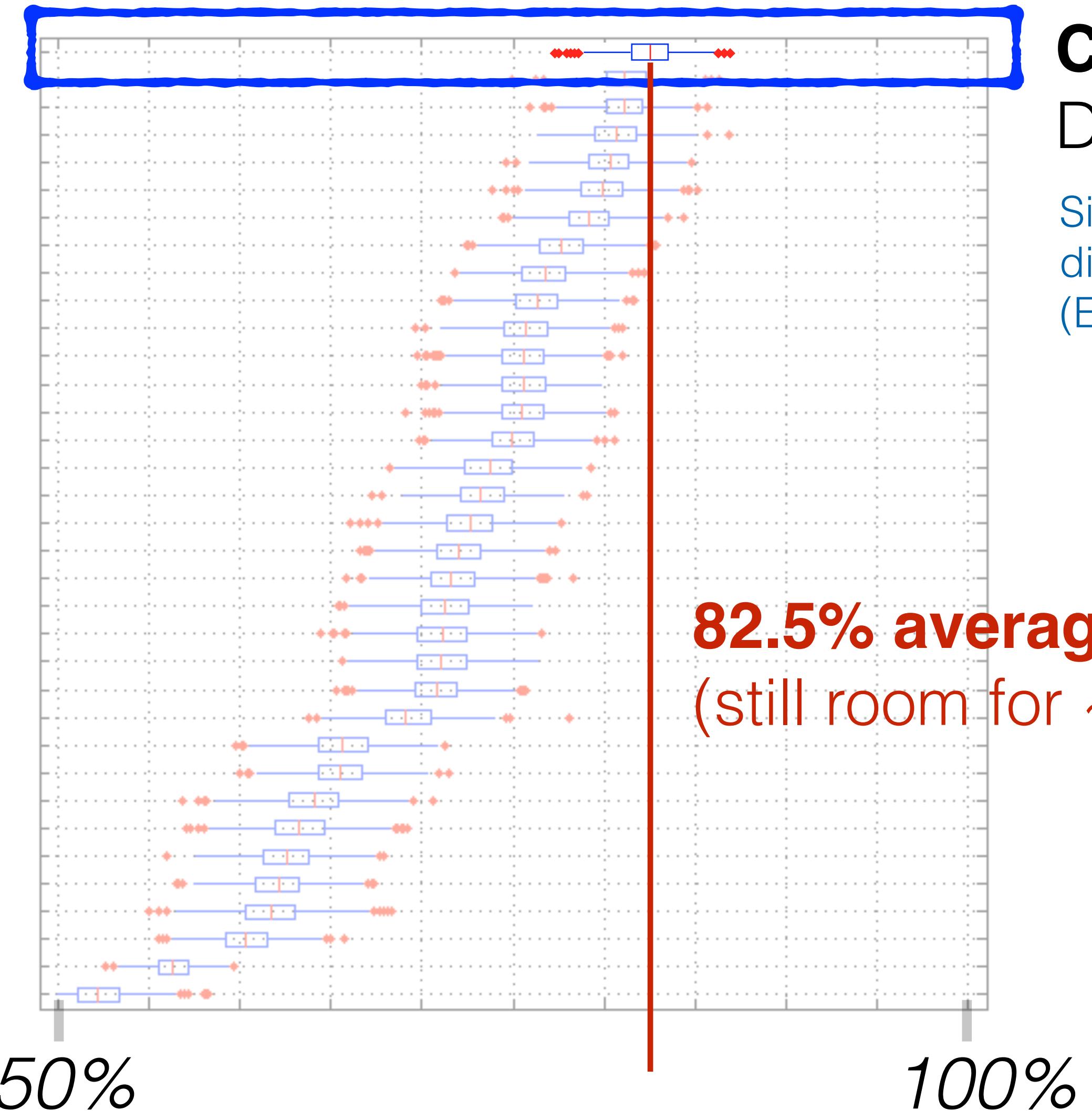
1. gather large collection of “ground truth” data
2. predict with different measures
3. evaluate quality: FP/FN
4. ROC/AUC
5. bootstrapping to generalize



Results

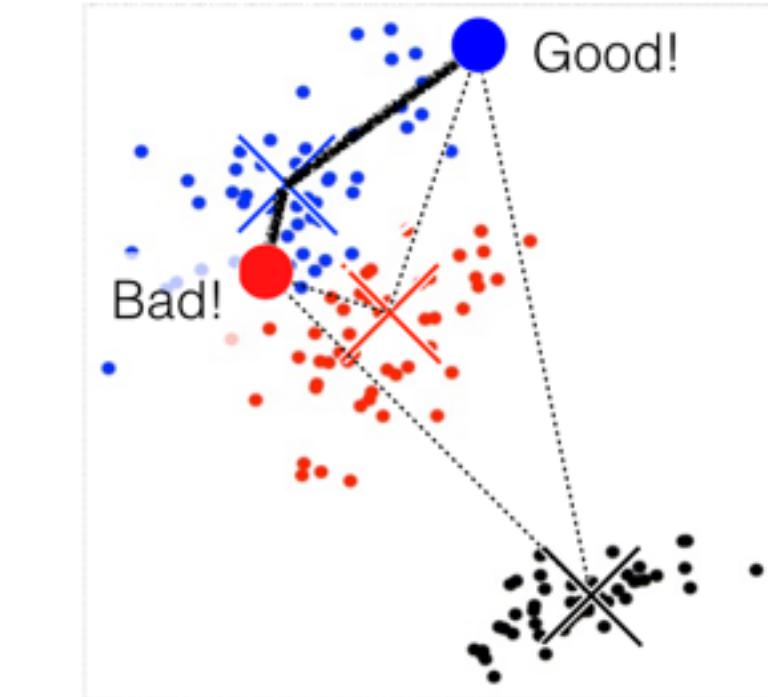


The Winner



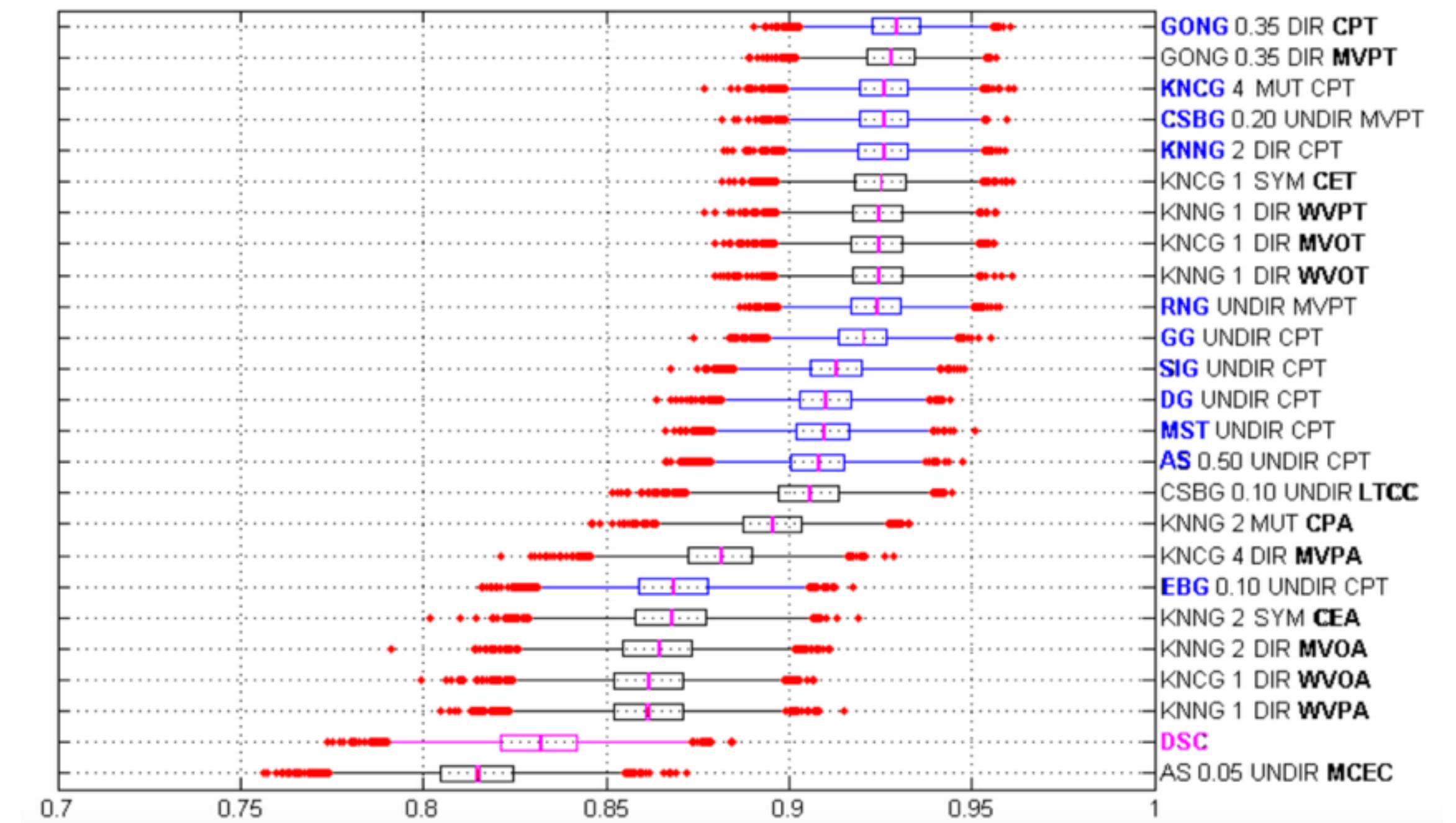
Centroid Measure
DSC - Distance Consistency

Sips et al.: Selecting good views of high-dimensional data using class consistency
(EuroVis 2009).

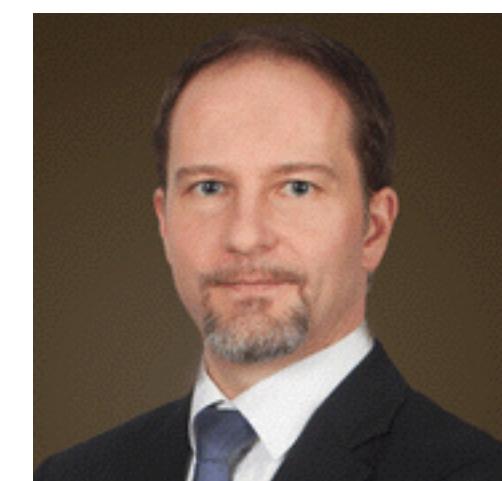


Follow-up work: PacificVis 2016

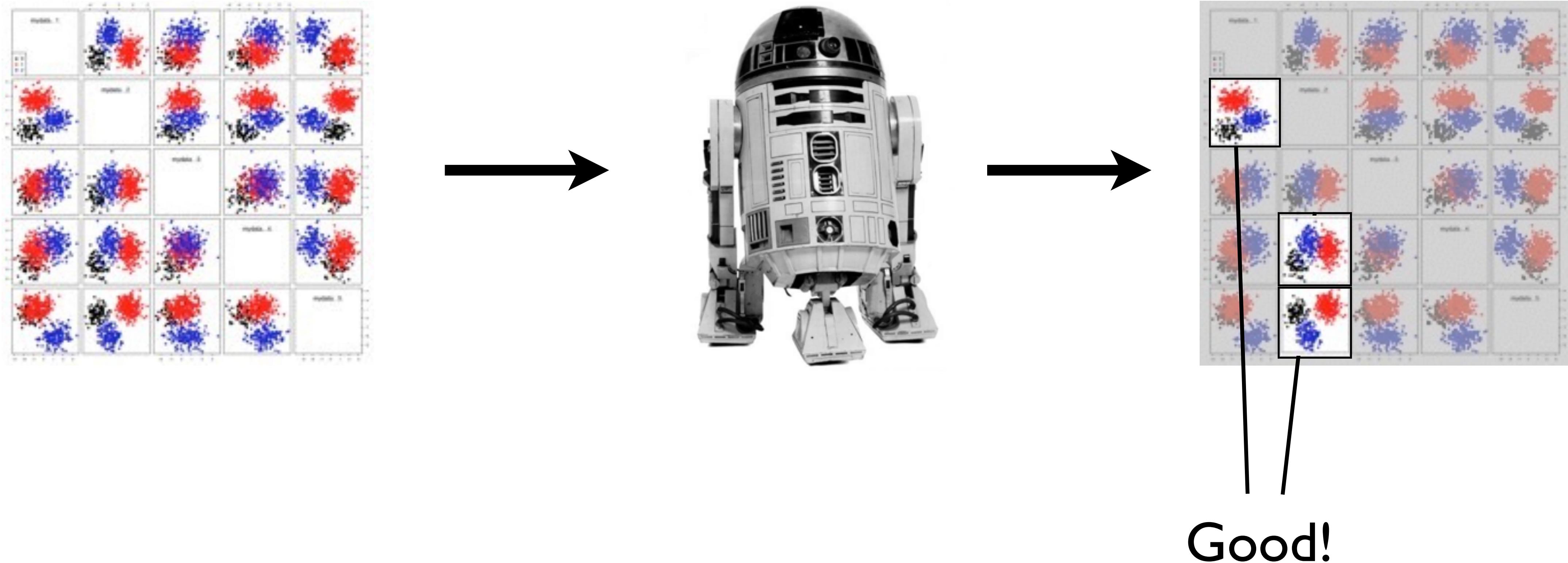
- 2002 New Visual Separation Measures
- 58% of our 2002 new measures outperform best state-of-the-art DSC
- Our best measure
 - GONG 0.35 DIR CPT
 - Prediction accuracy 92.9% (11.7% better than DSC)



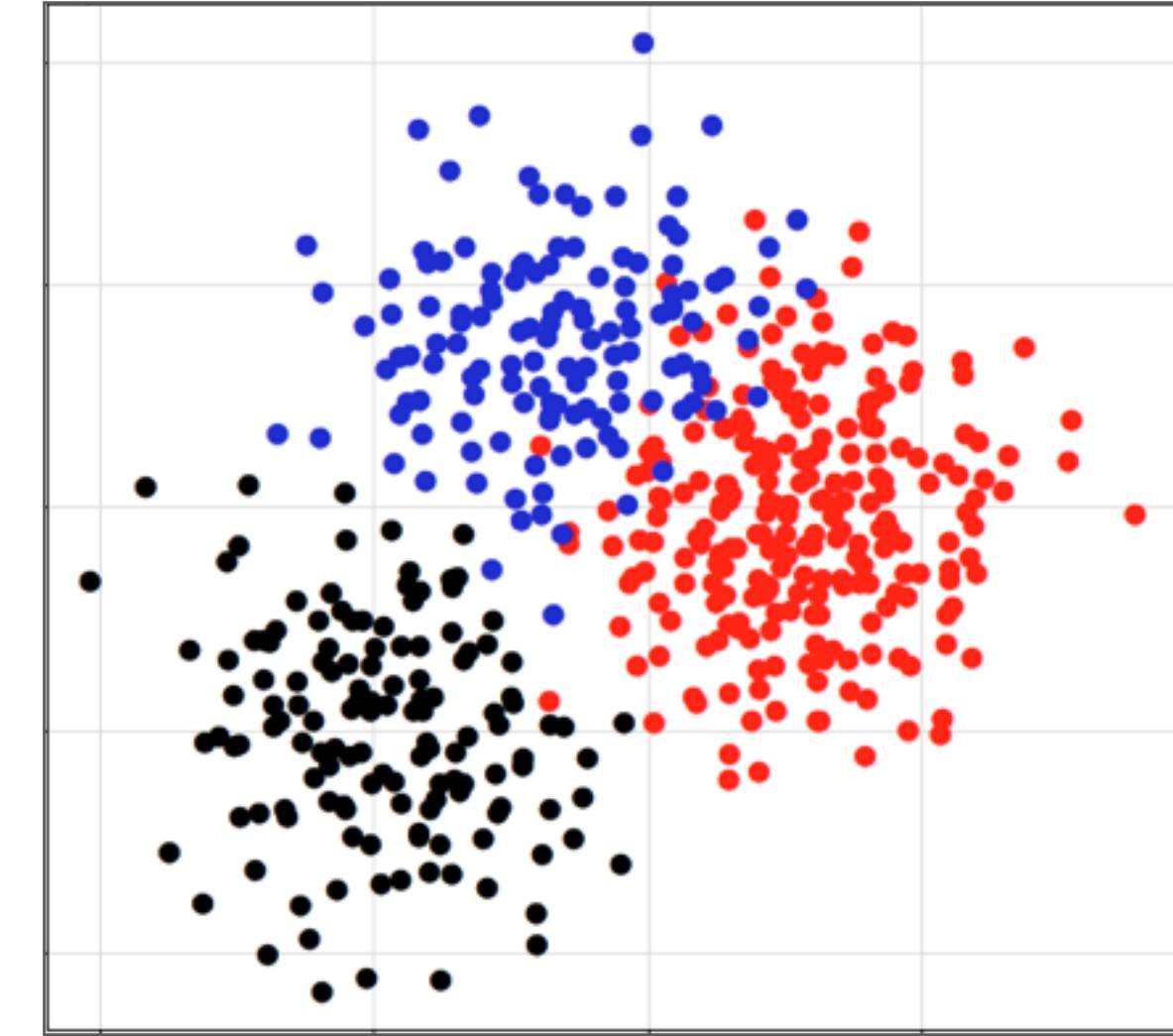
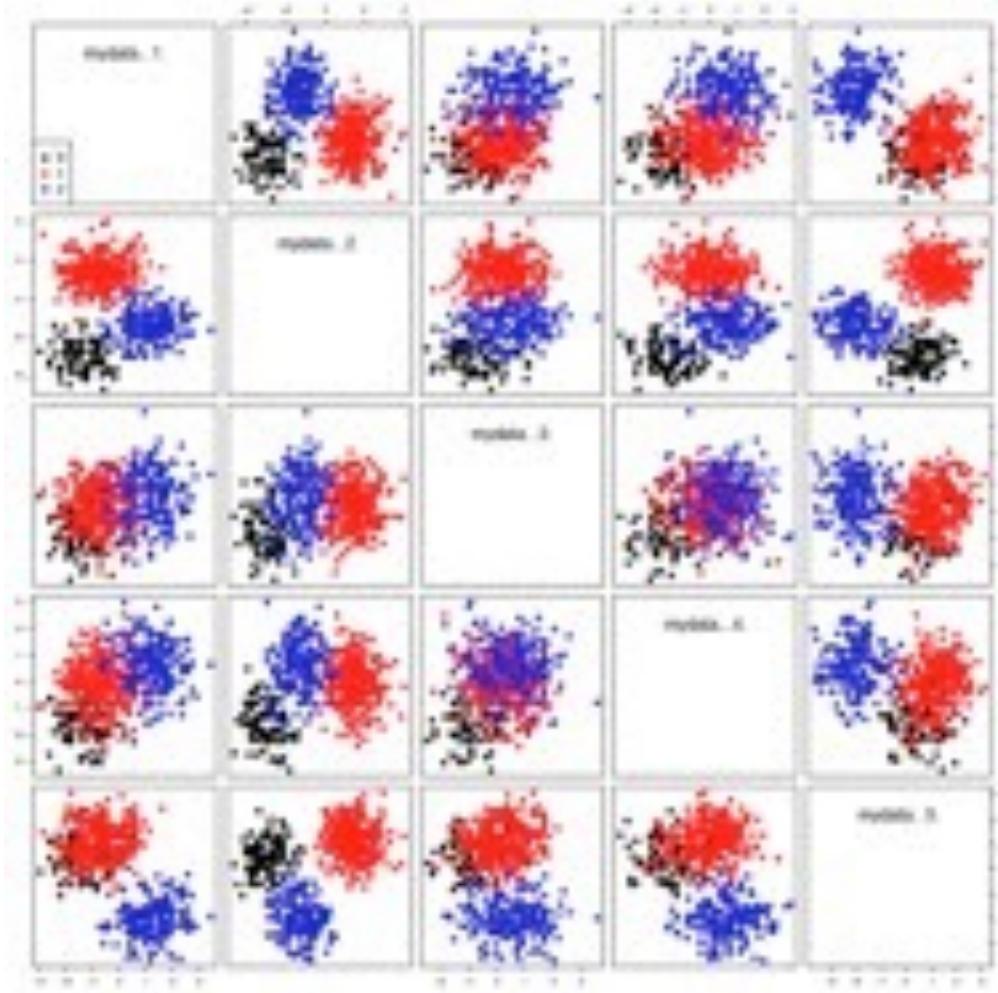
Aupetit, **Sedlmair**: SepMe: 2002 New Visual Separation Measures (PacificVis 2016).



Measures help humans finding the good views



Perception-driven Dimensionality Reduction



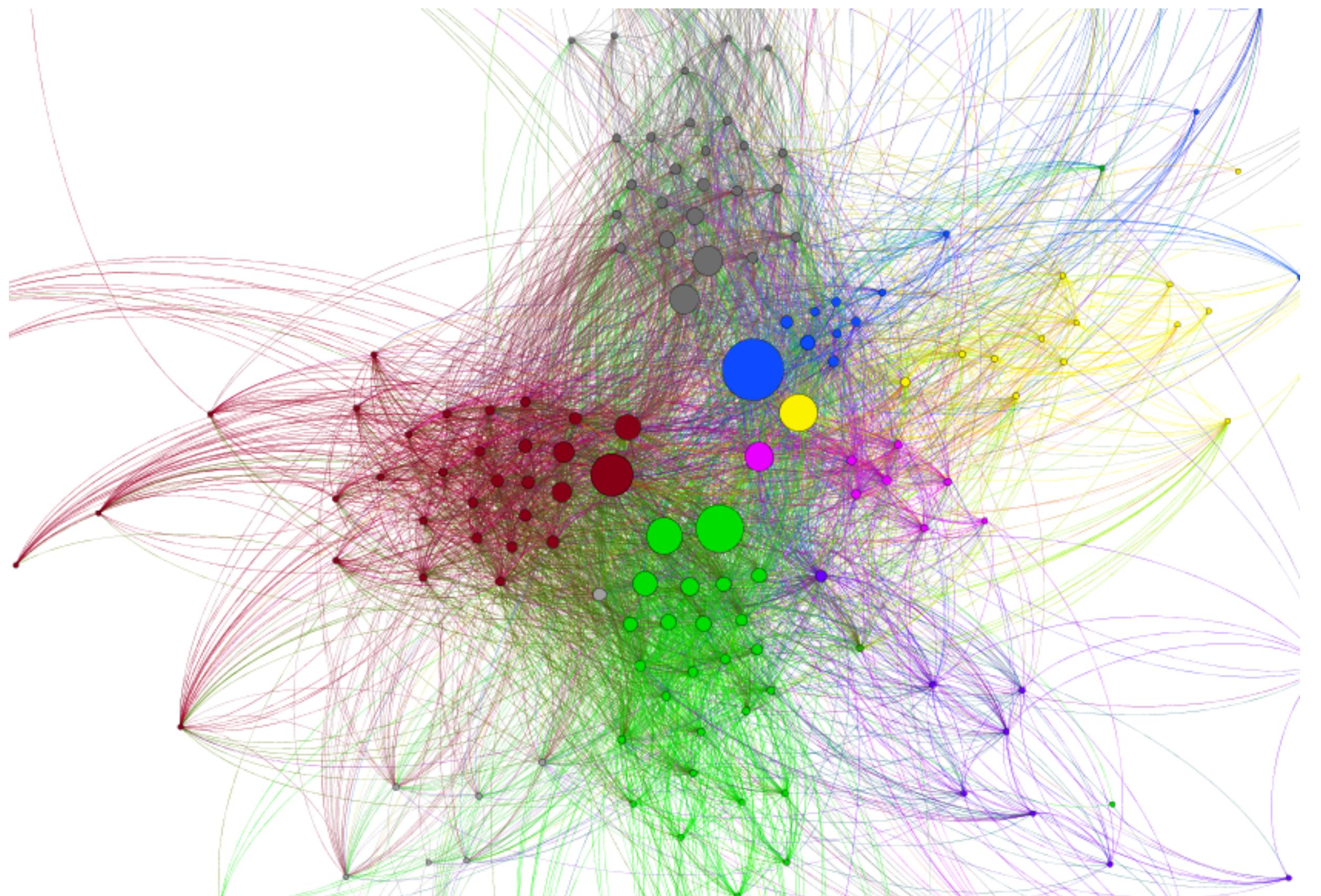
- Find the linear projection that perceptually separates classes as well as possible
- Better than LDA!

Yunhai Wang, Kang Feng, Xiaowei Chu, Jian Zhang, Chi-Wing Fu, **Sedlmair**, Xiaohui Yu, Baoquan Chen.
A Perception-Driven Approach to Supervised Dimensionality Reduction for Visualization. (TVCG 2017)



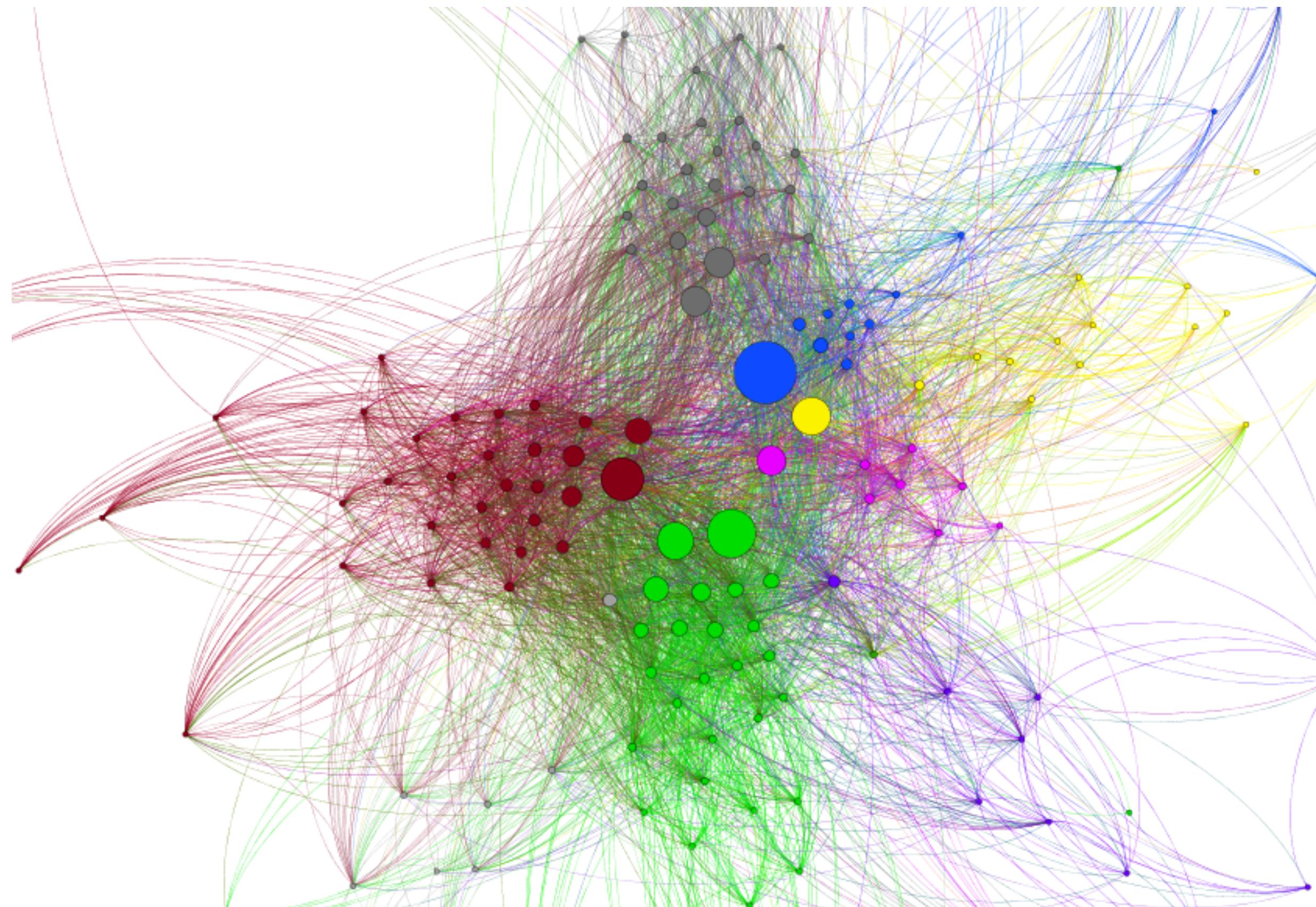
Make it interactive ...

Showing all the data?



<https://twitter.com/axelmaireder/media>

Showing all the data?



<https://twitter.com/axelmaireder/media>



Hairball

Make it interactive

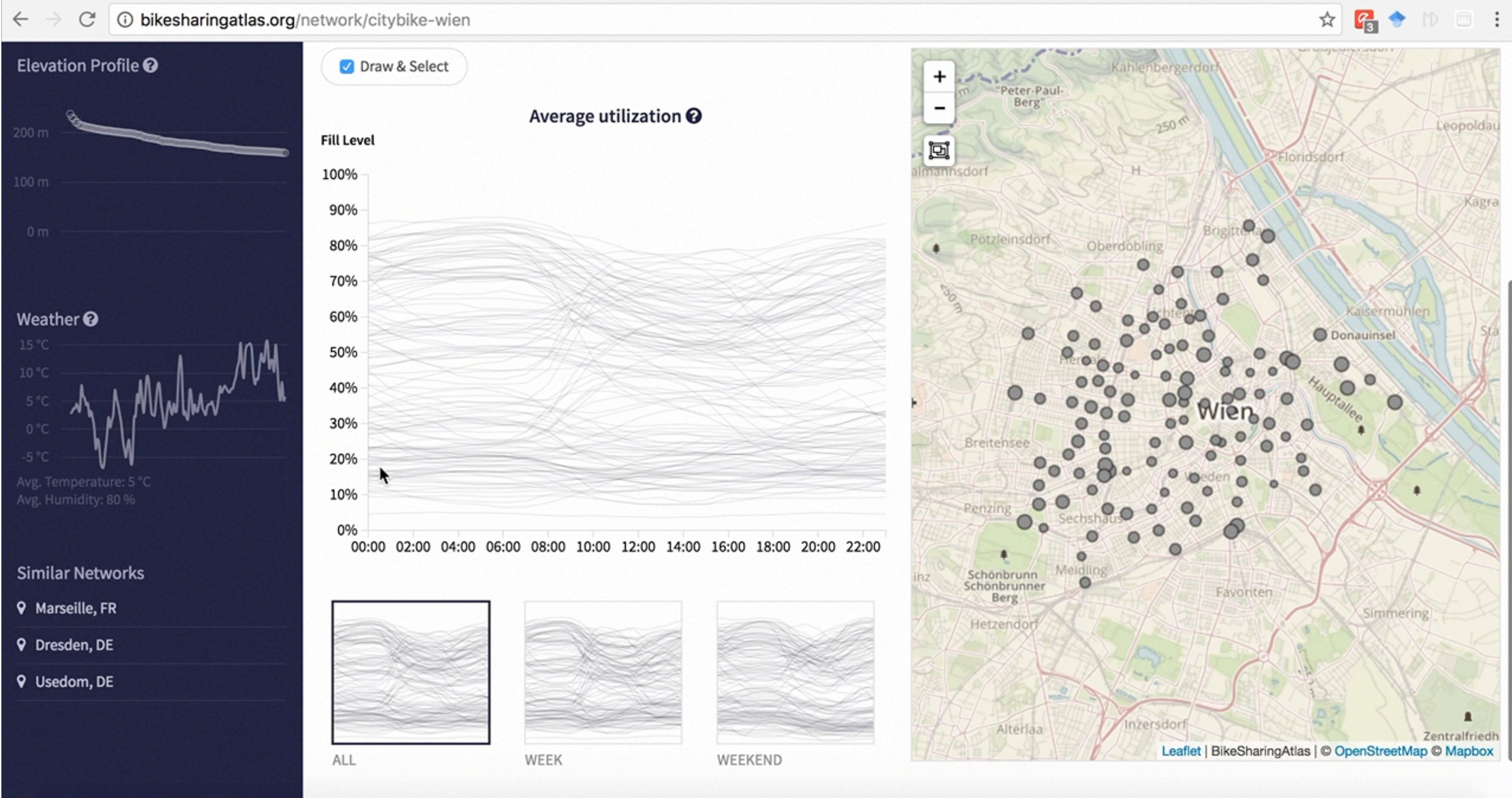


example:
bikesharingatlas.org

- 468 networks
- 21.500 stations
- Fill levels every 15 min, for 1.5 years
(i.e. >1 billion fill level values)

Oppermann, Möller, **Sedlmair**.
To appear in International Journal of Transportation.





bikesharingatlas.org



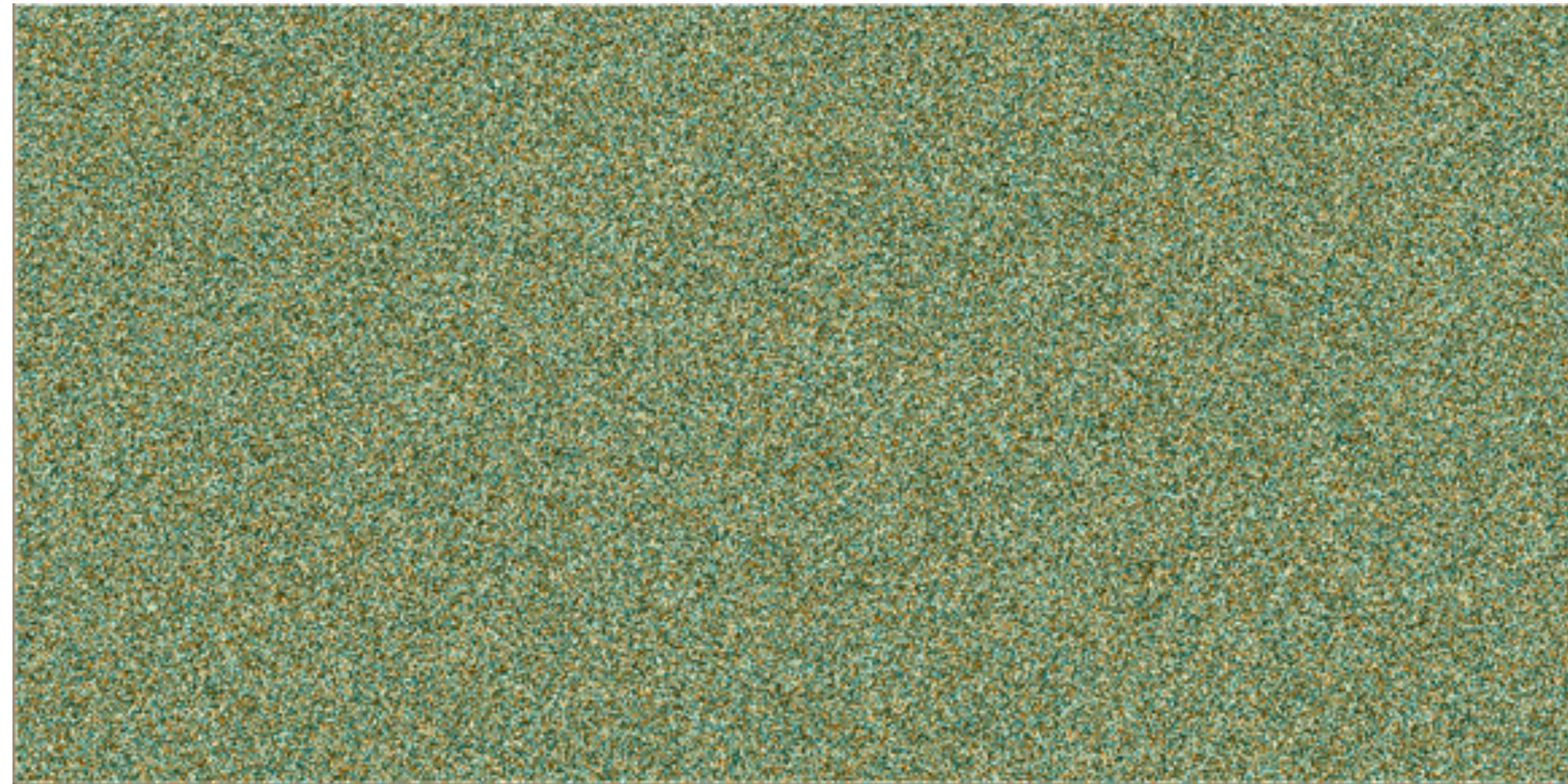
- Multiple views - linking & brushing
- Novel patterns and insights
- Different user groups
(network operators, city planners, politicians, journalists, general public, etc.)

Oppermann, Möller, **Sedlmair**.

*To appear in International
Journal of Transportation.*

Combine
Computation & Visualization

Even too much data for interaction



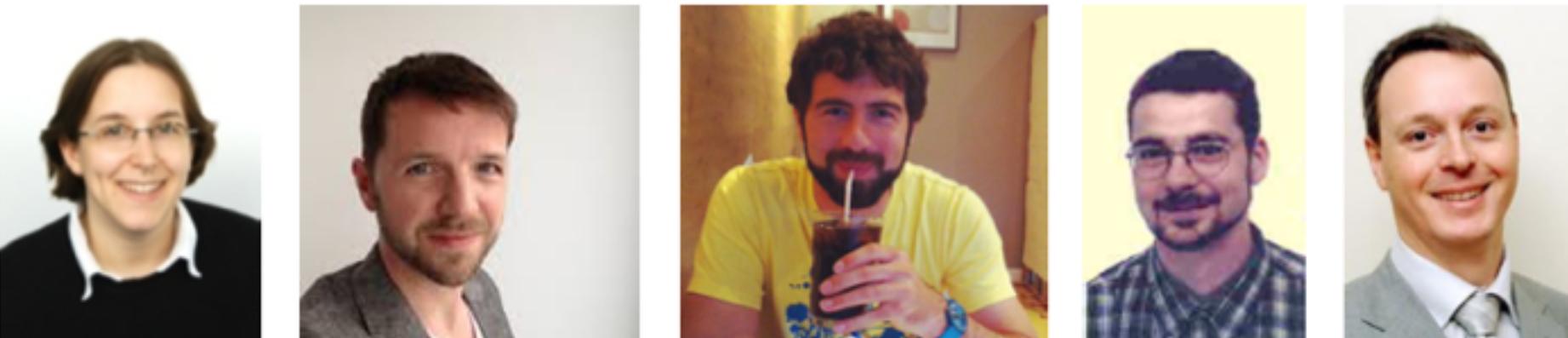
<https://www.geek.com/geek-cetera/pi-visualized-to-4-million-decimal-places-in-one-image-1486849/>

Combine visual and computational

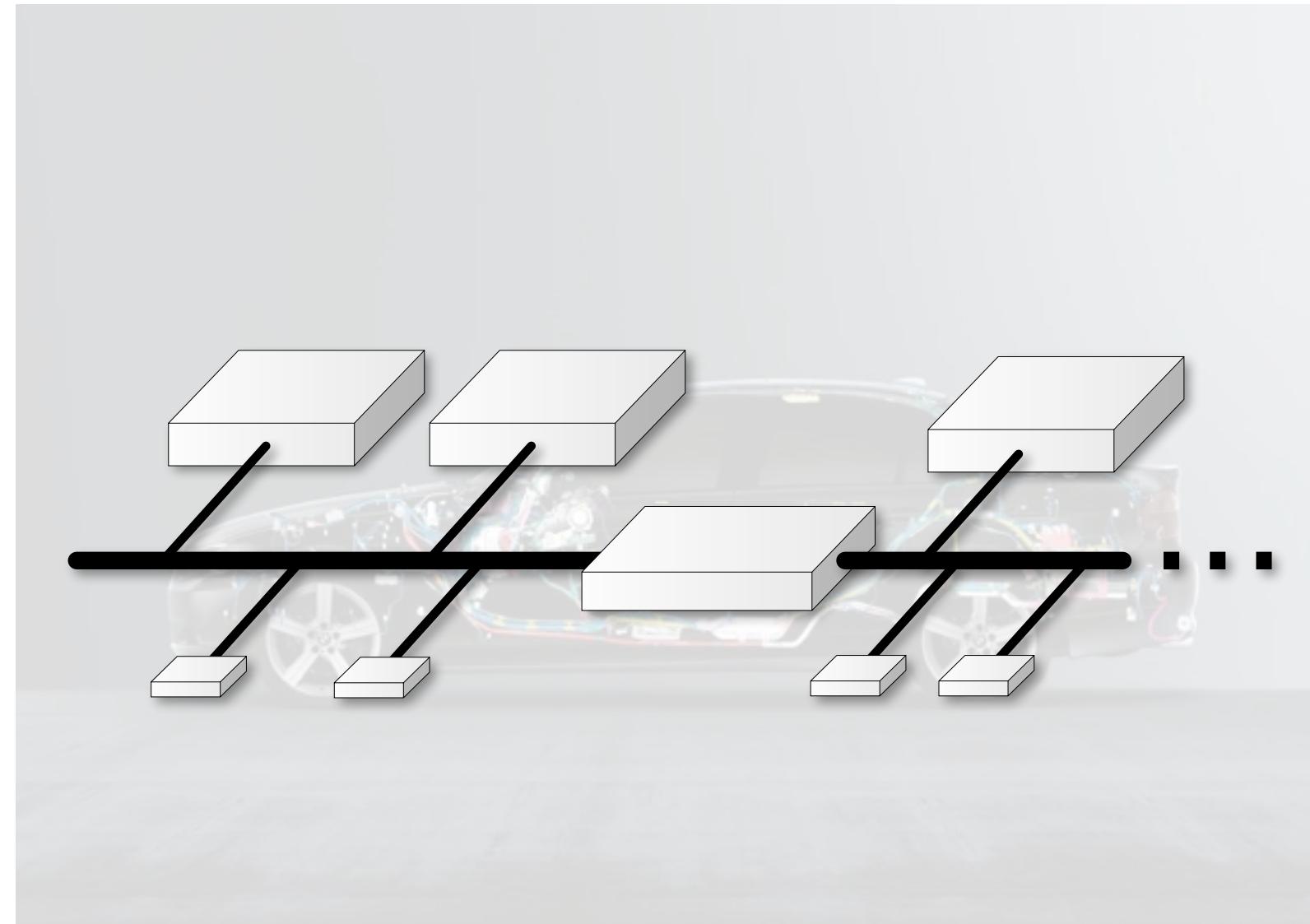


example: Cardiogram @BMW

Sedlmair, Isenberg, Baur, Mauerer,
Pigorsch, Butz: Cardiogram: Visual Analytics
for Automotive Engineers (CHI 2011).



Combine visual and computational

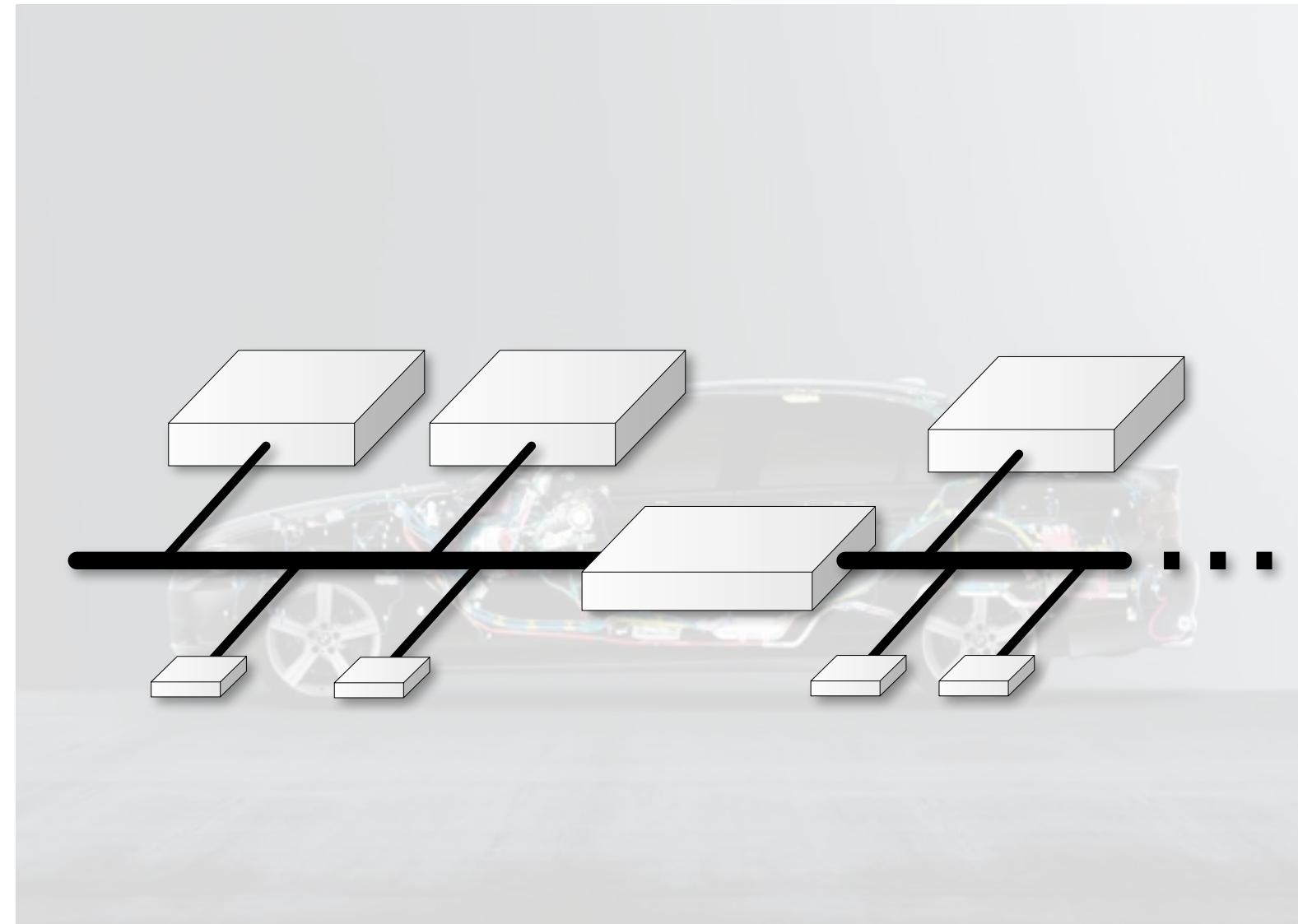


example: Cardiogram @BMW

- 70 interconnected ECUs

Sedlmair, Isenberg, Baur, Mauerer,
Pigorsch, Butz: Cardiogram: Visual Analytics
for Automotive Engineers (CHI 2011).

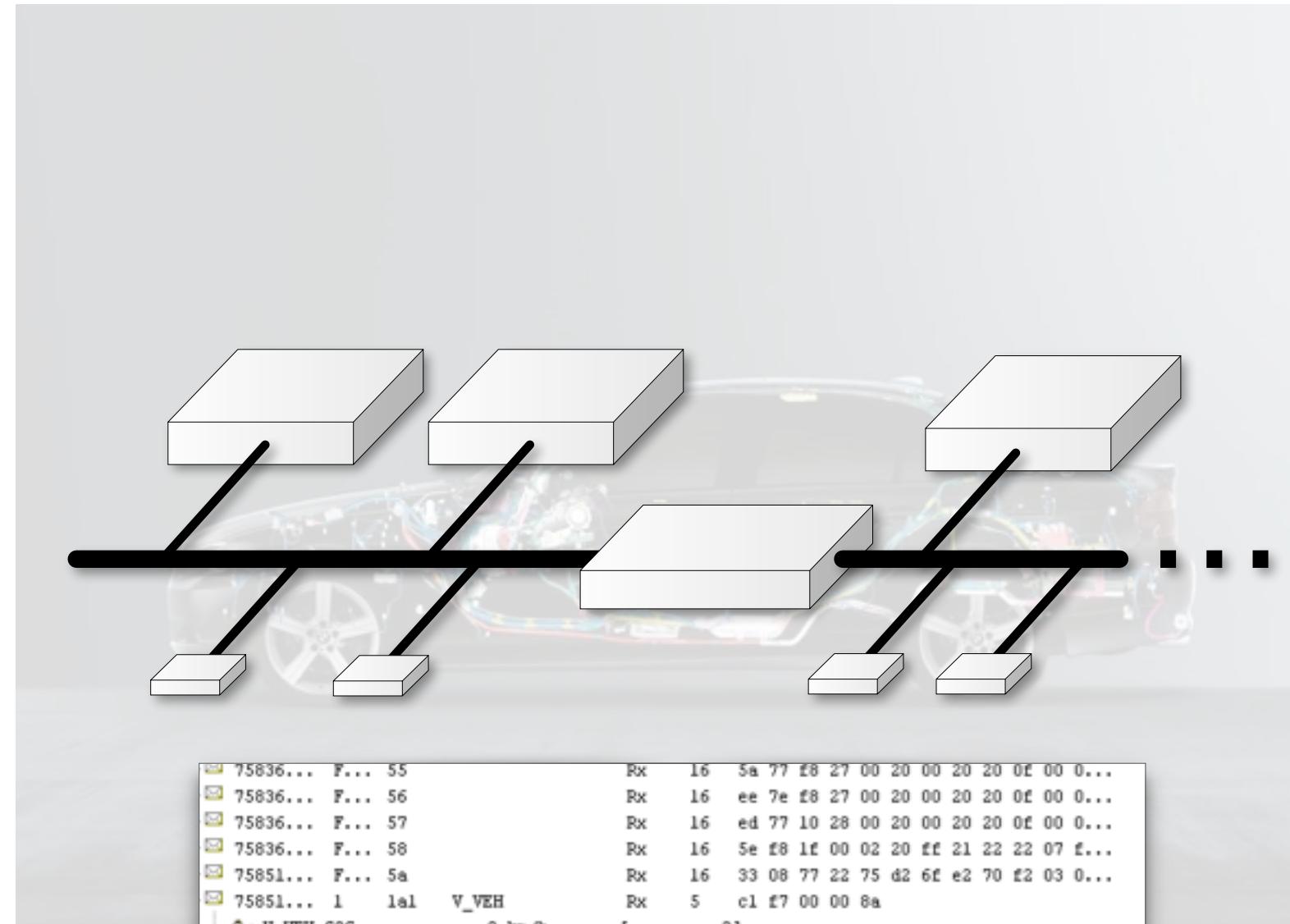
Combine visual and computational



Sedlmair, Isenberg, Baur, Mauerer,
Pigorsch, Butz: Cardiogram: Visual Analytics
for Automotive Engineers (CHI 2011).

- example: Cardiogram @BMW
- 70 interconnected ECUs
 - **task:** finding errors
 - **process:** analyse test drives
 - **data:** recorded traces
(15k messages/sec)

Combine visual and computational



```
75836... F... 55 Rx 16 5a 77 f8 27 00 20 00 20 20 0f 00 0...
75836... F... 56 Rx 16 ee 7e f8 27 00 20 00 20 20 0f 00 0...
75836... F... 57 Rx 16 ed 77 10 28 00 20 00 20 20 0f 00 0...
75836... F... 58 Rx 16 5e f8 1f 00 02 20 ff 21 22 22 07 f...
75851... F... 5a Rx 16 33 08 77 22 75 d2 6f e2 70 f2 03 0...
75851... 1 la1 V_VEH Rx 5 c1 f7 00 00 8a
~ V_VEH_COG 0 km/h [ 0]
~ ST_ECU_V_VEH Signal ungültig [ f]
~ QU_V_VEH_COG Signalwert ist gültig, Zustand/S [ a]
~ DVCO_VEH Fahrzeug steht [ 0]
~ CRC_V_VEH 193 [ c1]
~ ALIV_V_VEH 7 [ 7]
75851... F... 5c Rx 8 00 00 00 00 ff ff 00 10
75851... 1 1c4 Rx 6 00 00 00 00 ff ff
75851... F... 12t Rx 72 00 00 00 00 00 ff 59 87 21 4c 0...
75851... 1 1c5 Rx 6 02 00 04 00 ff ff
75836... F... 1 Rx 16 10 7d 18 28 00 20 00 20 0f 00 0...
75836... F... 2 Rx 16 ca 76 f9 27 00 20 00 20 0f 00 0...
75836... F... 3 Rx 16 6a 7d fa 27 00 20 00 20 0f 00 0...
75836... F... 4 Rx 16 a3 76 0e 28 00 20 00 20 0f 00 0...
75836... F... 5 Rx 16 2c f7 1f 00 02 20 ff 21 22 22 07 f...
75851... F... 7 Rx 16 6d f9 76 12 75 d2 6f f2 70 f2 01 1...
75851... F... 12 Rx 0
75851... 2 301 AVL_STEA_DV Rx 7 51 15 f8 7f ff 7f 11
75851... 3 301 AVL_STEA_DV Rx 7 51 15 f8 7f ff 7f 11
75851... 4 137 Rx 2 fd 00
75851... 3 d9 ANG_ACPD Rx 8 9b 99 00 c0 00 e0 7f f0
75848... 1 299 Rx 5 9f ff ff ff ff
75851... F... 21 Rx 0
75851... 4 a5 TORQ_CRSH_1 Rx 8 45 f5 48 f7 7f 00 00 fc
75851... 3 a5 TORQ_CRSH_1 Rx 8 fe f9 48 f7 7f 00 00 fc
75836... F... 23 Rx 16 7c 10 05 05 ea f3 53 20 74 10 20 f...
75851... F... 26 Rx 0
```

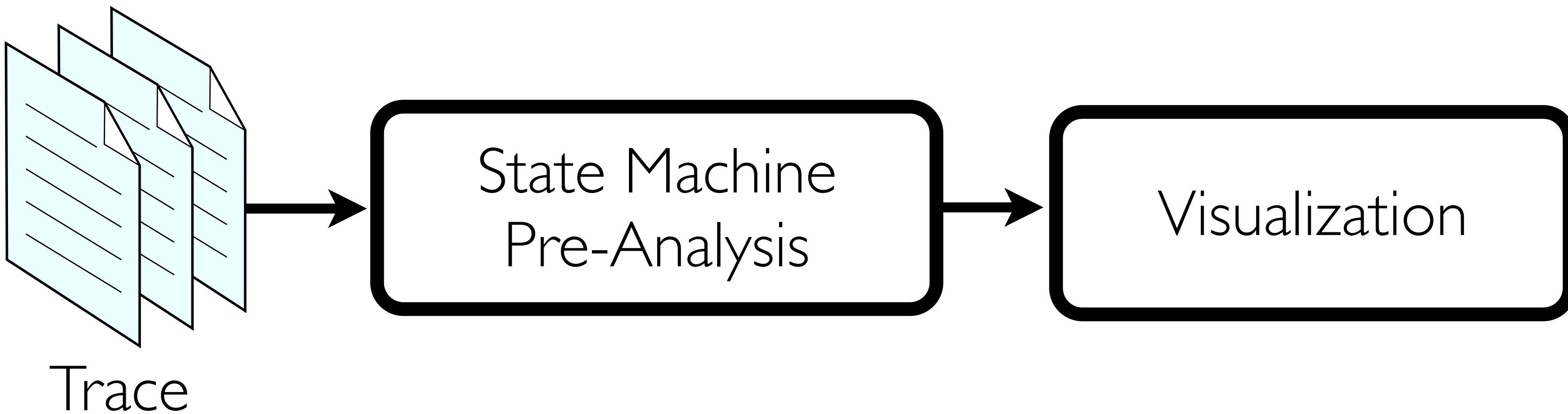
example: Cardiogram @BMW

- 70 interconnected ECUs
- **task:** finding errors
- **process:** analyse test drives
- **data:** recorded traces
(15k messages/sec)
- **current practices:** mainly textual lists

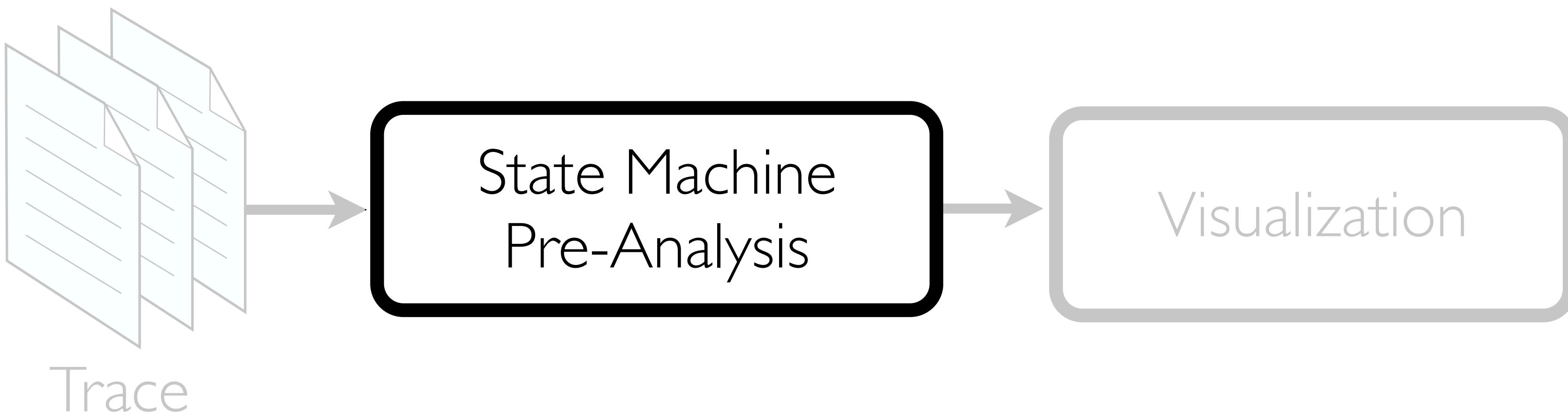
Main challenges

- handling masses of test traces (large and many)
- understanding correlation between trace and car behavior
- ... (*and others*)

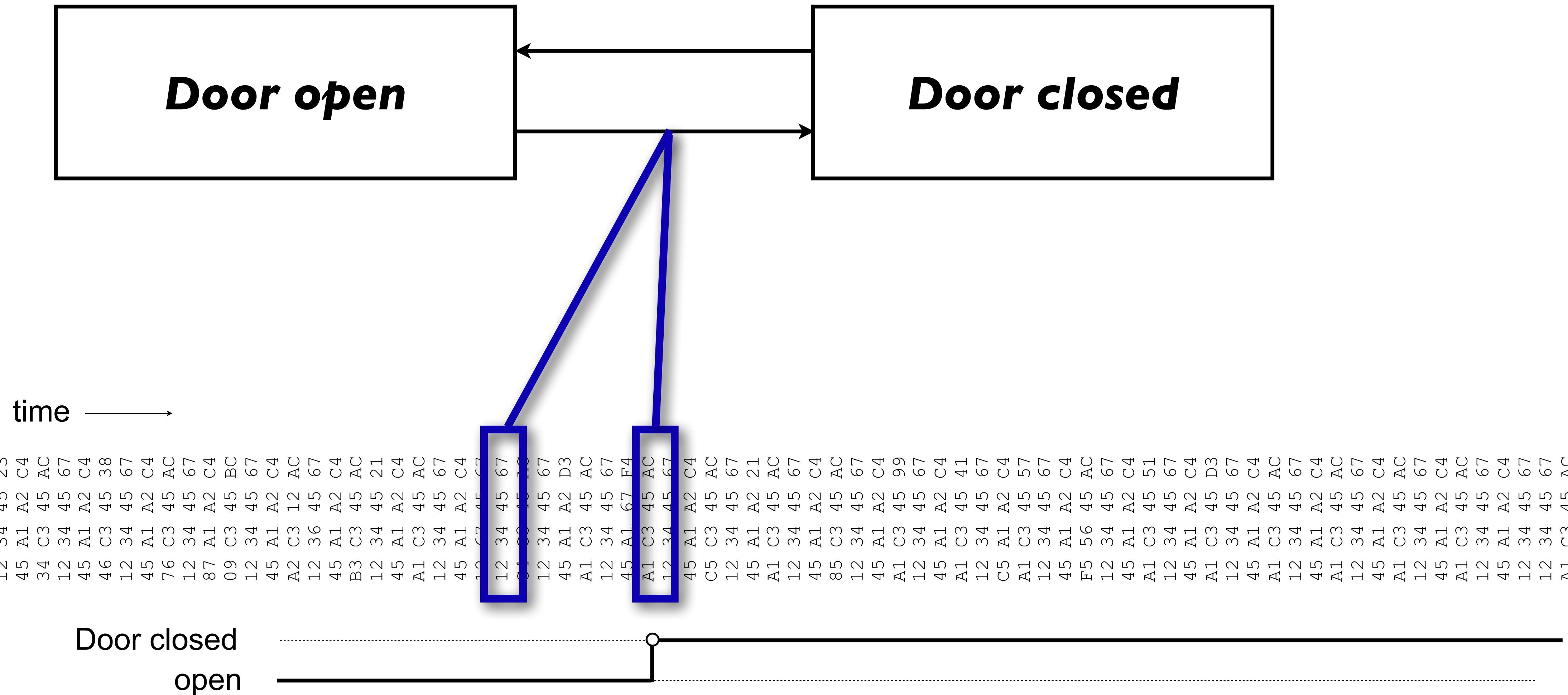
Our solution: Cardiogram



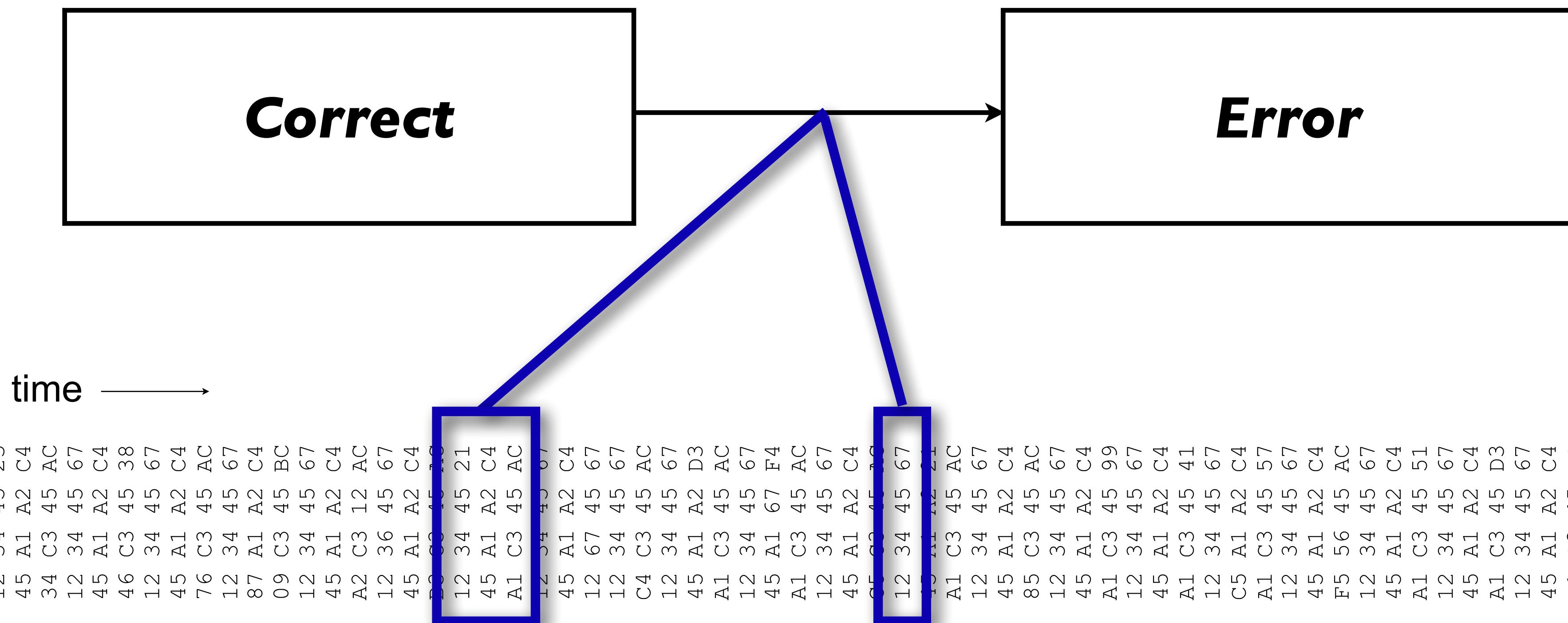
State machine pre-analysis



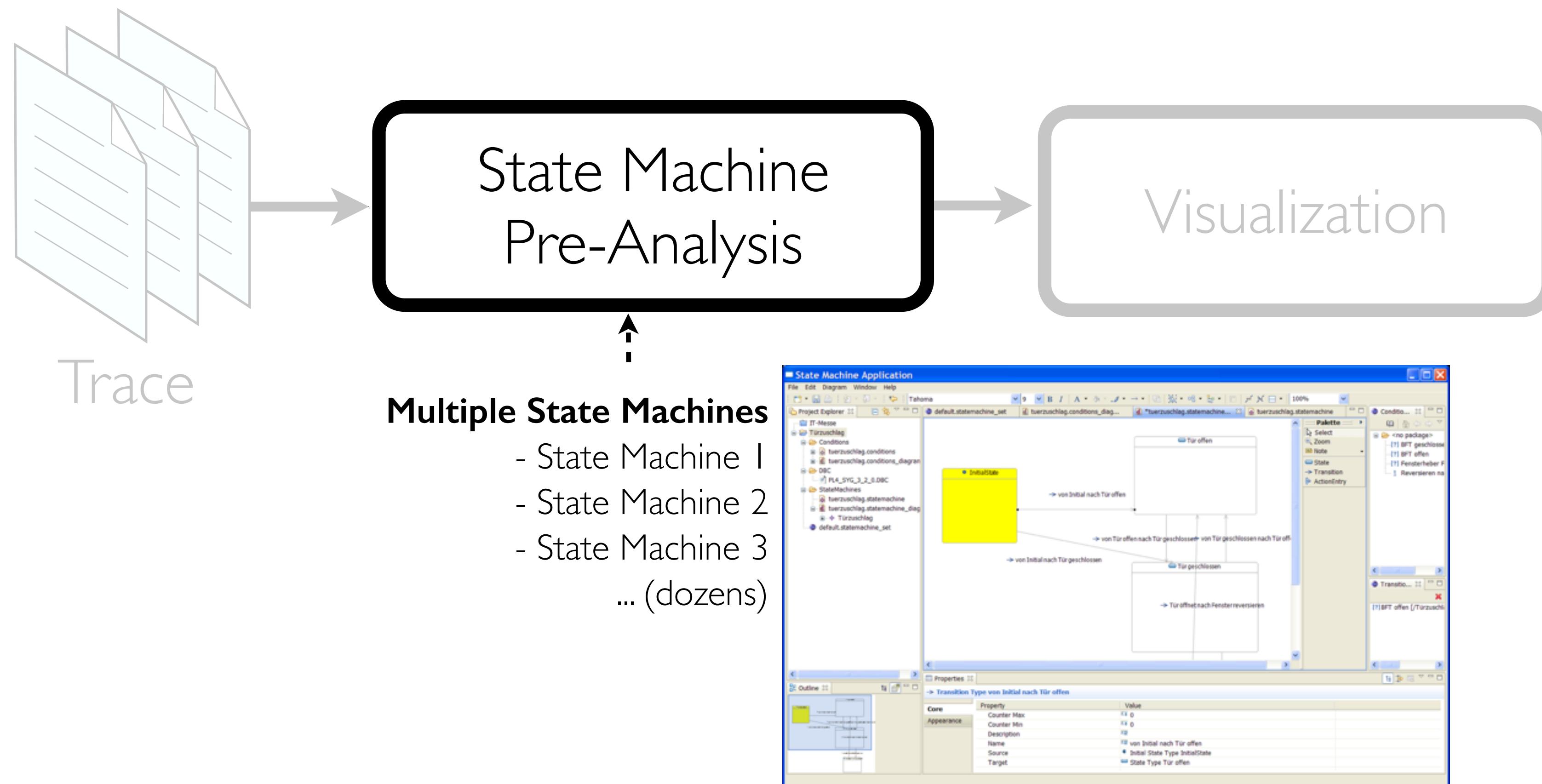
State machines: Behavior



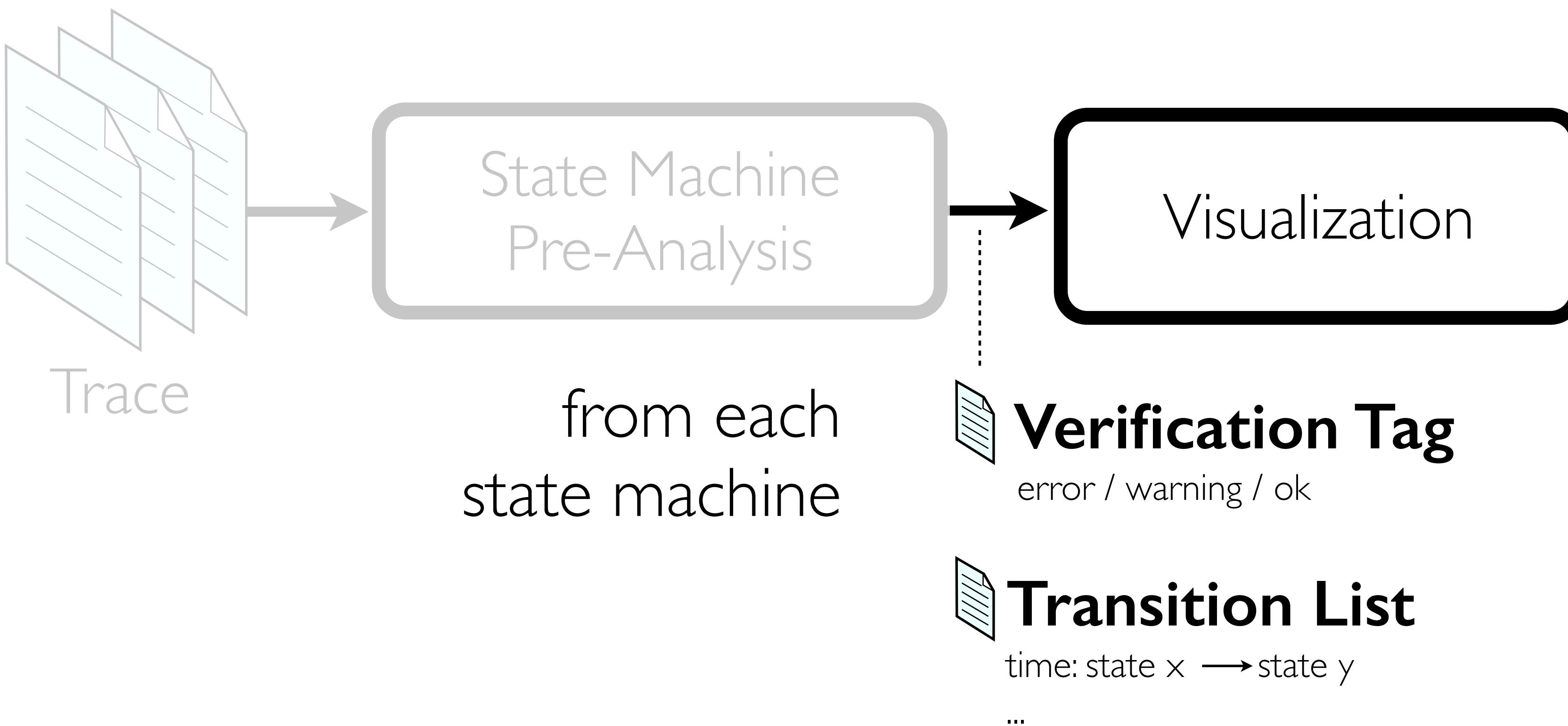
State machines: Error detection

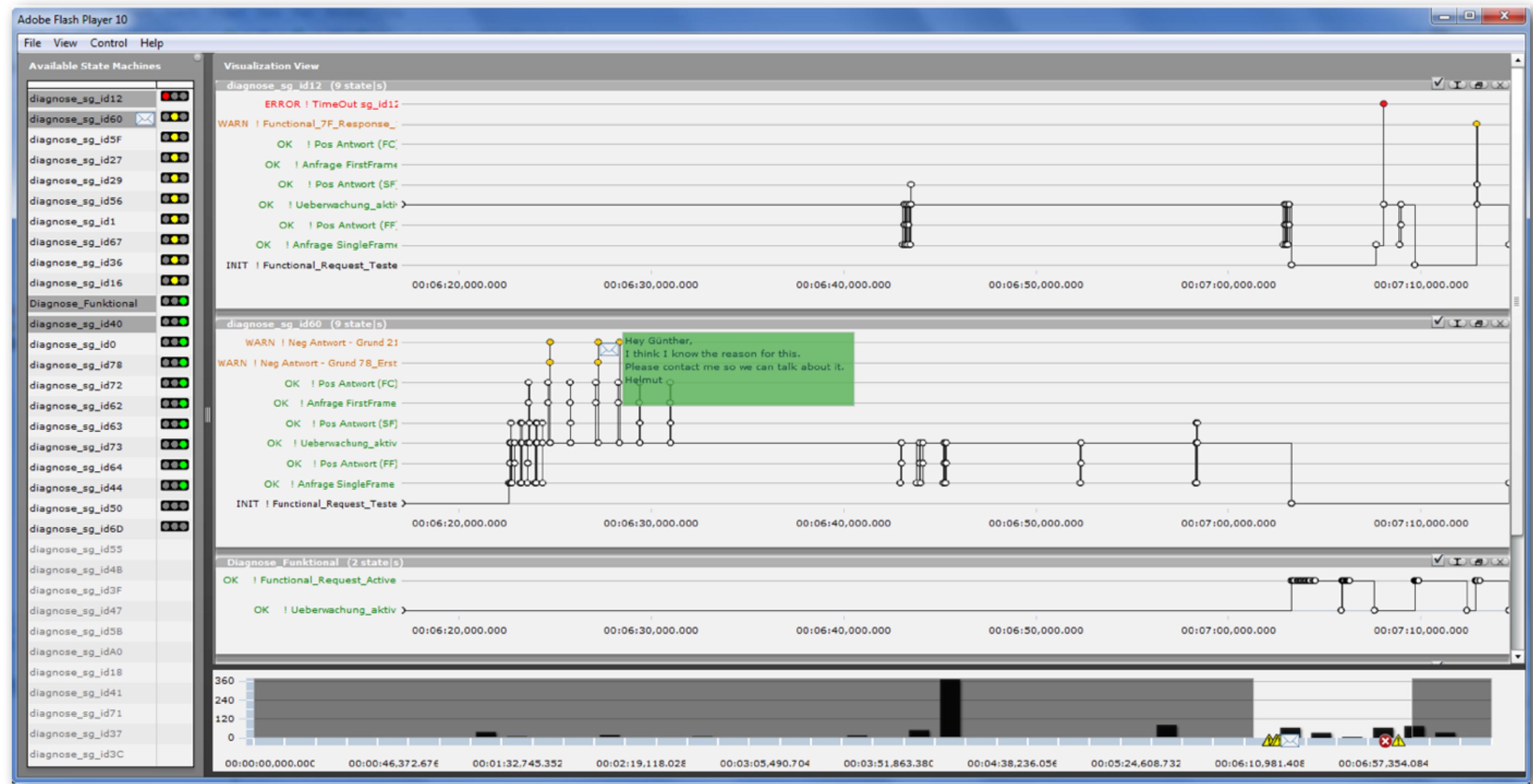


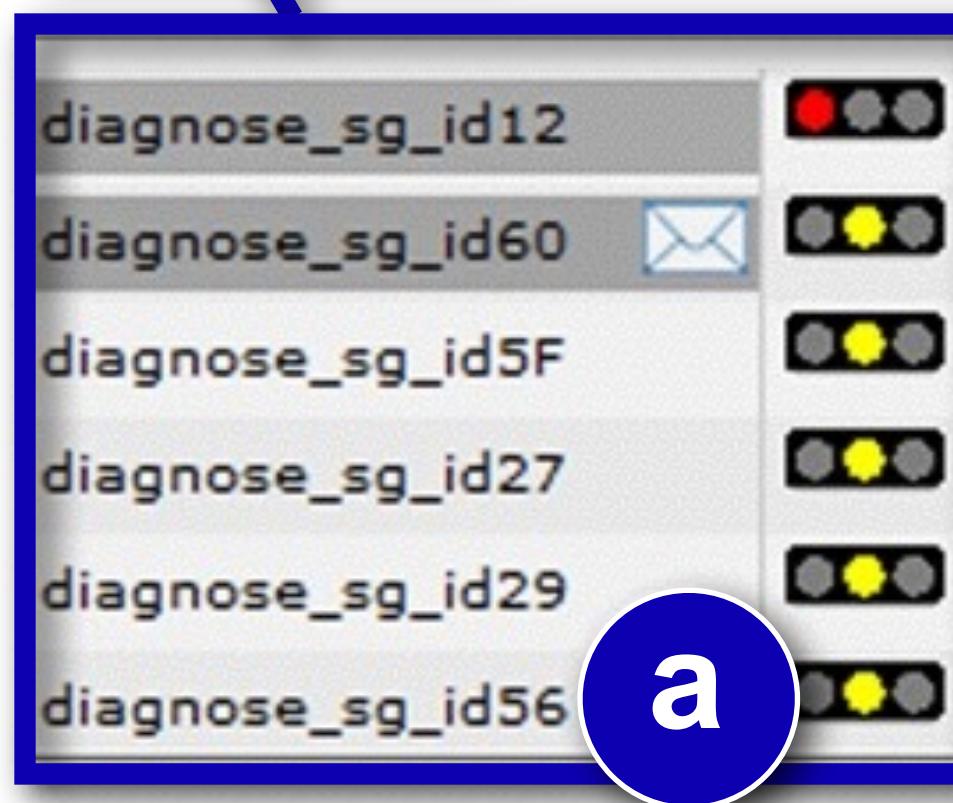
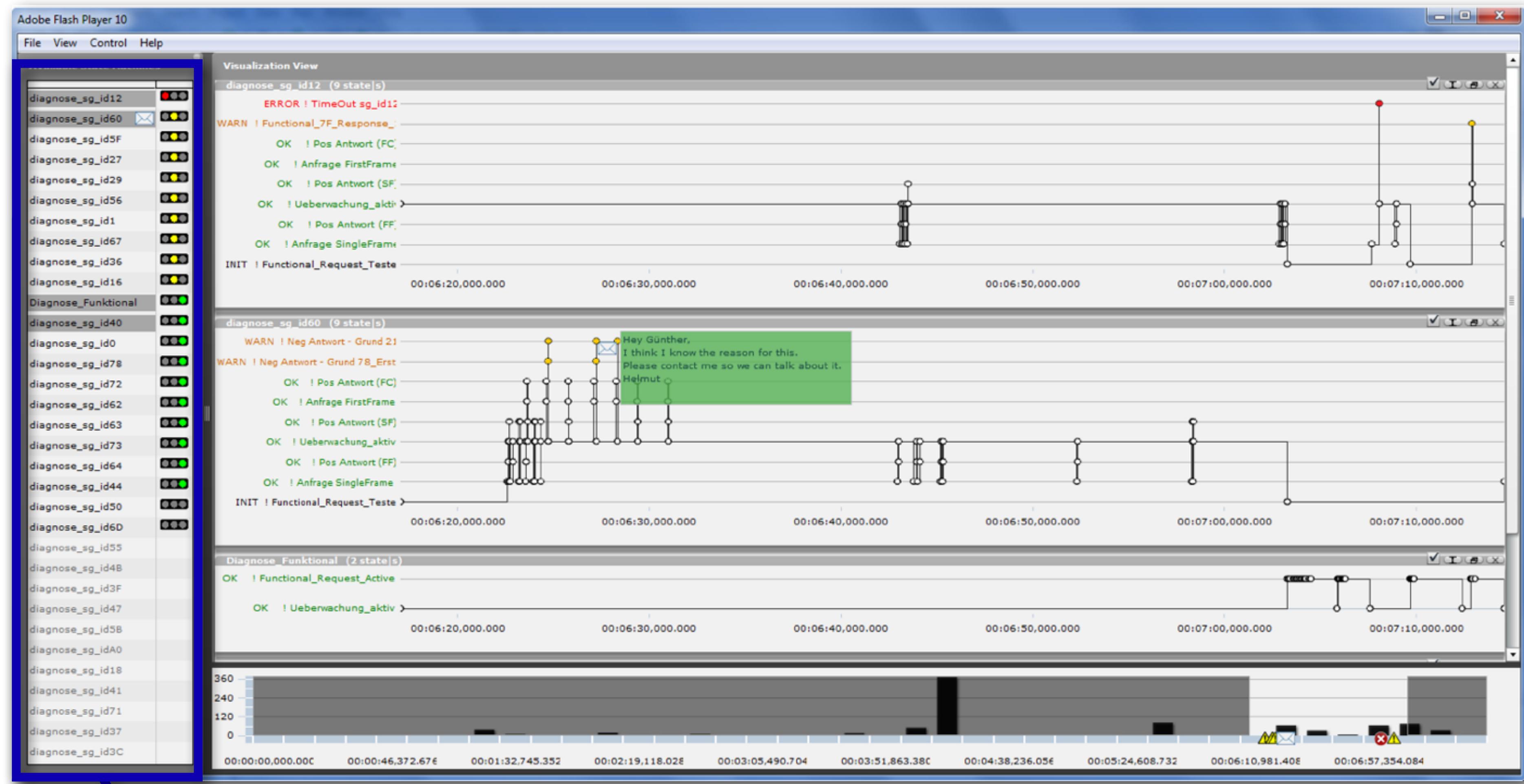
Create/select state machines

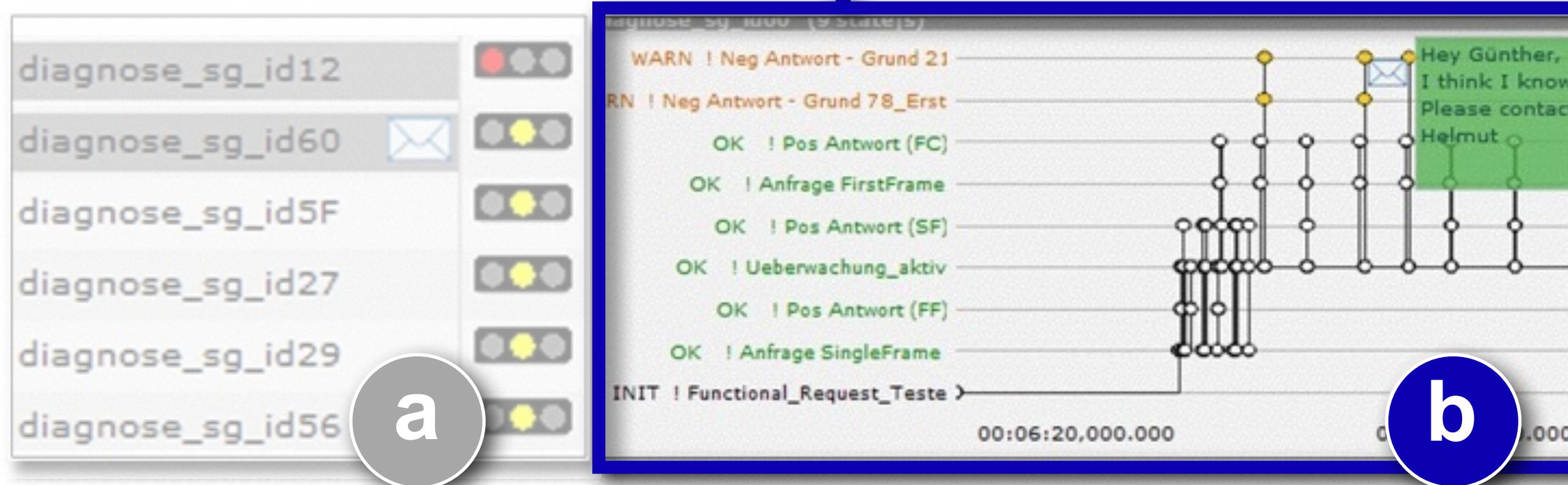
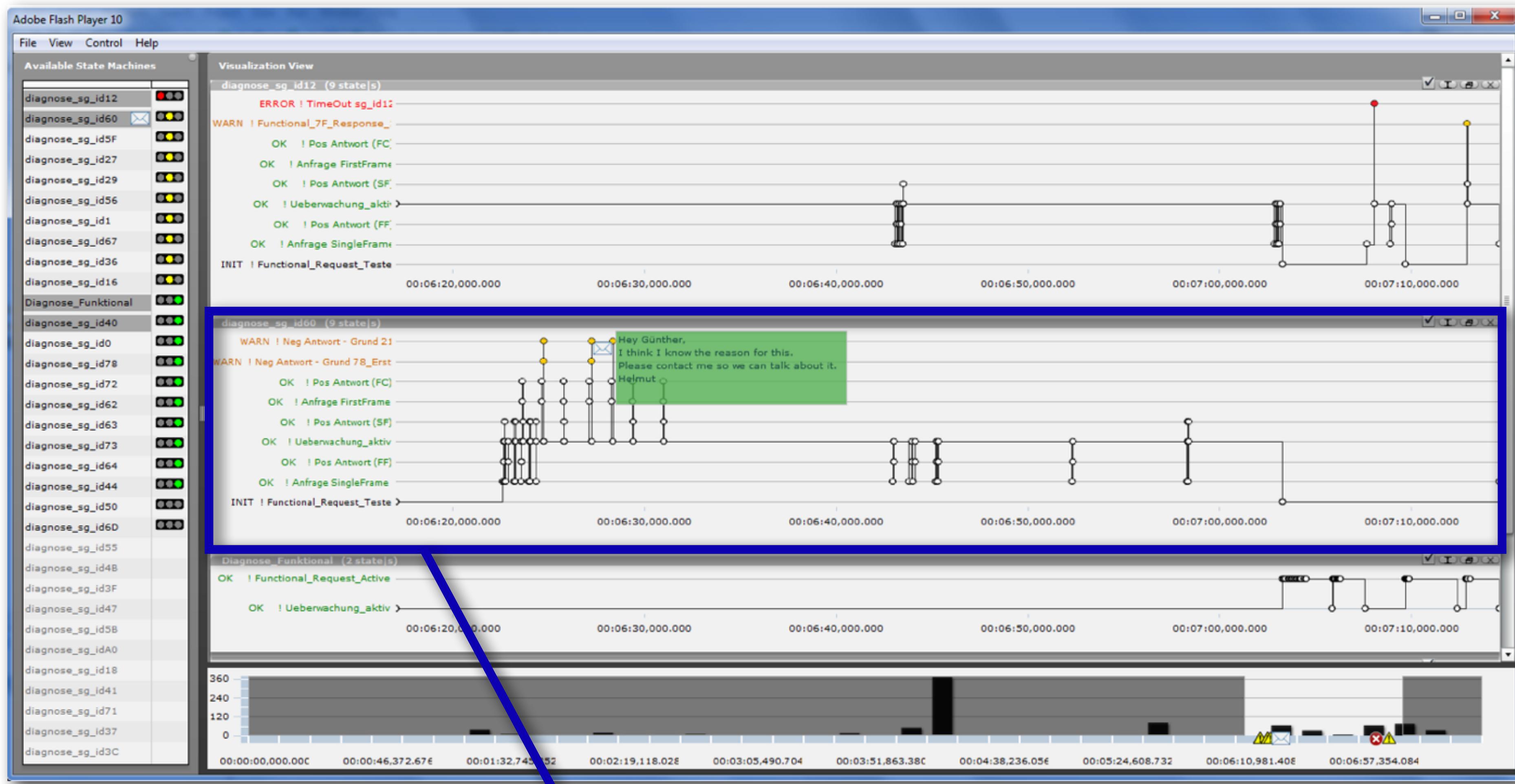


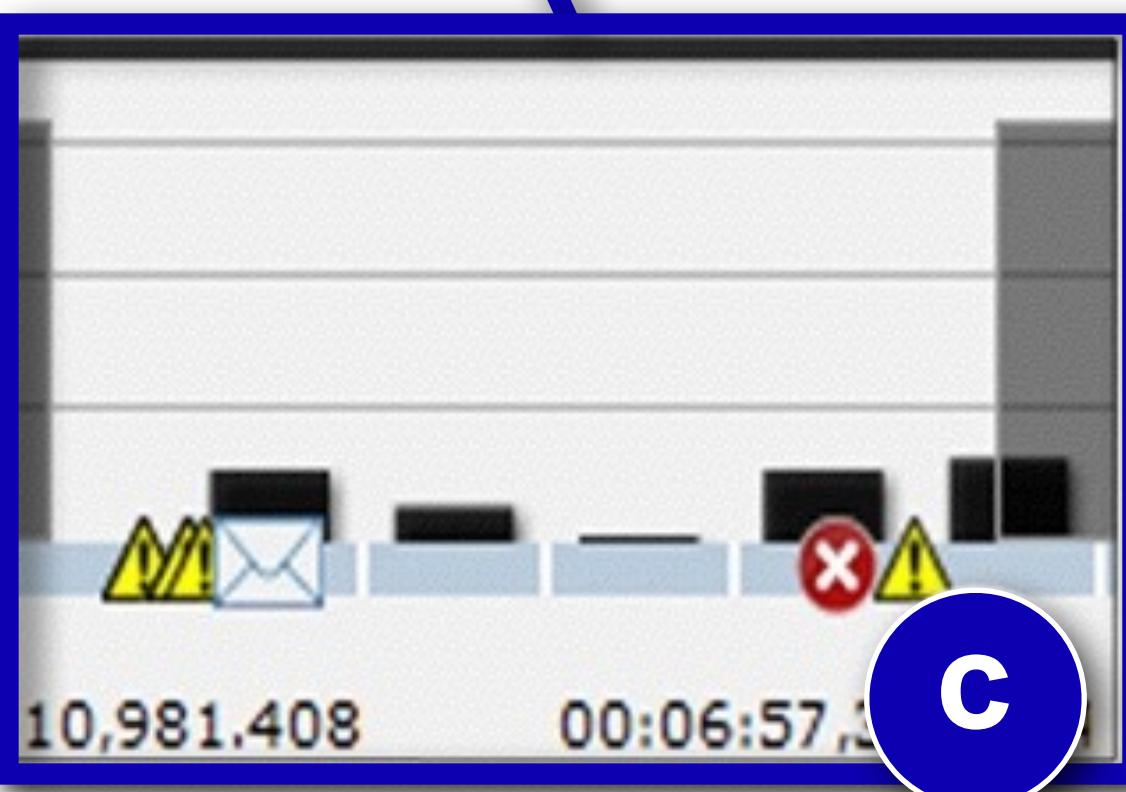
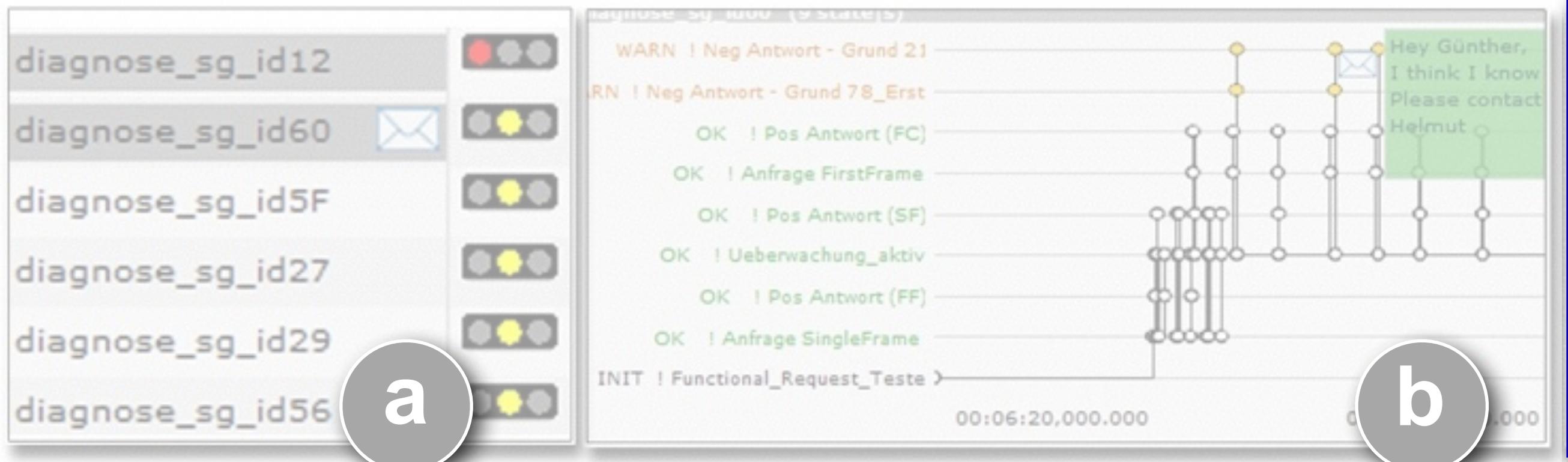
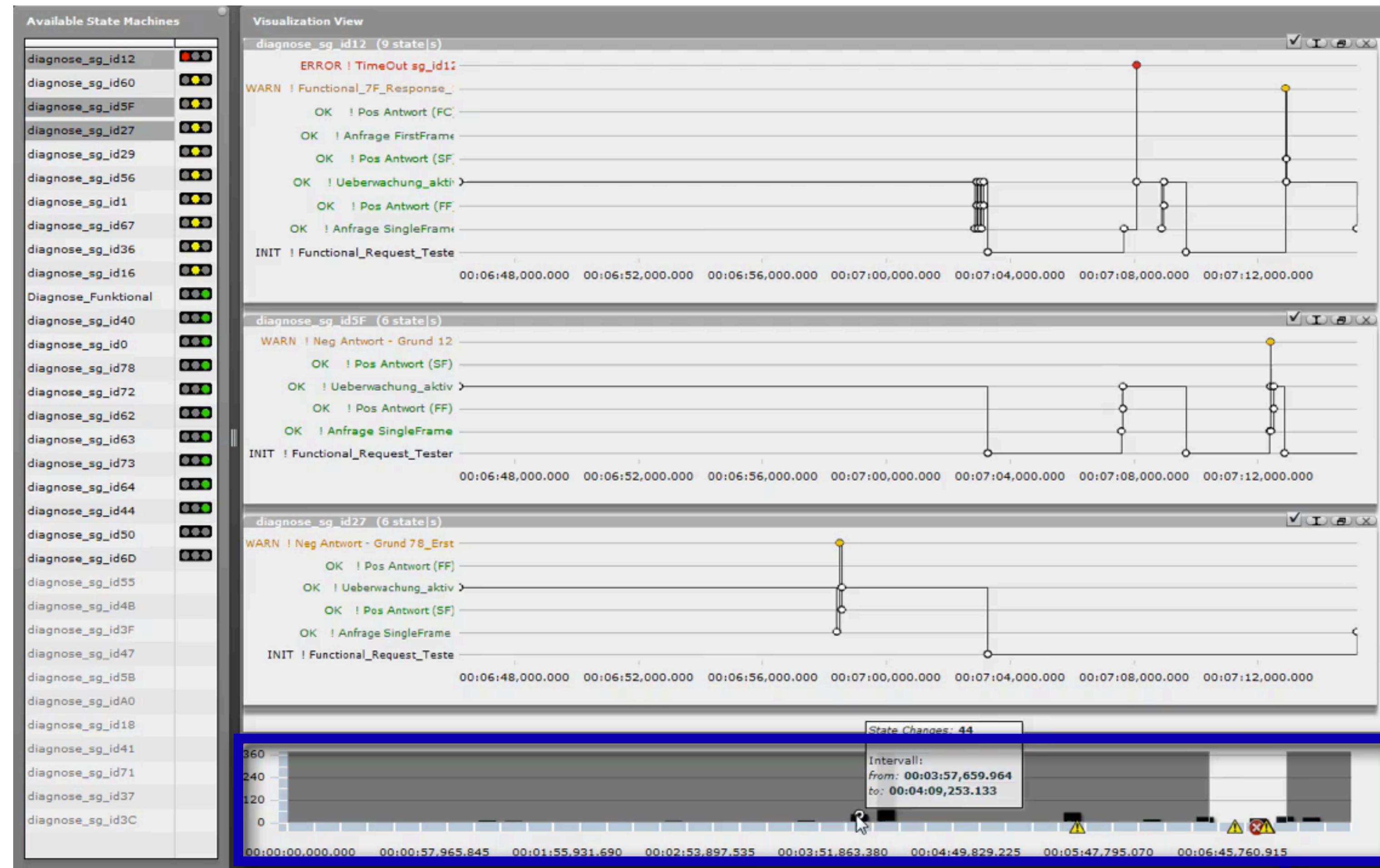
Data abstraction/reduction









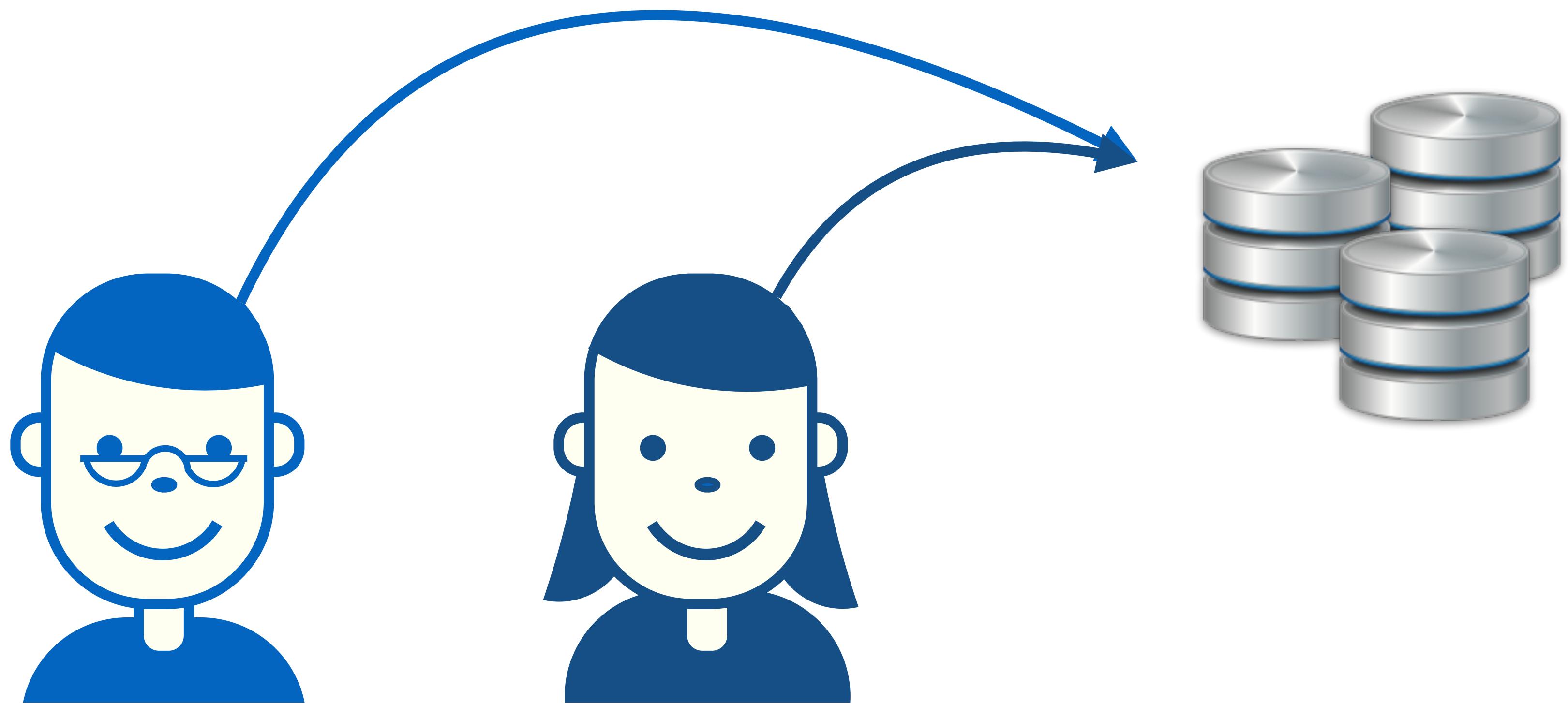


Longitudinal field study

- 1 year, 15 engineers

Results

- externalization and sharing of expert knowledge



Results

- externalization and sharing of expert knowledge
 - complete coverage of traces vs. sparse sampling

```

    75836... F.. 55 Rx 16 $8 f7 f8 27 00 20 00 20 00 01 00 0...
    75836... F.. 56 Rx 16 ee f8 27 00 20 00 20 00 20 00 00 0...
    75836... F.. 57 Rx 16 ed 77 10 28 00 20 00 20 00 00 0...
    75836... F.. 58 Rx 16 5e ff 1f 00 02 20 ff 21 22 22 07 f...
    75851... F.. 5a Rx 16 33 08 77 22 75 62 6f e2 70 f2 03 0...
    75851... 1 1al V_VEH Rx 5 cl f7 00 00 08

    ~ V_VEH_COG 0 km/h [ 0 ]
    ~ ST_ECU_V_VEH Signal ungültig [ f ]
    ~ QU_V_VEH_COG Signalwert ist gültig, Zustand/S [ a ]
    ~ DVCO_VEH Fahrzeug steht [ 0 ]
    ~ CRC_V_VEH 193 [ c1 ]
    ~ ALIV_V_VEH 7 [ 7 ]

    75851... F.. 5c Rx 8 00 00 00 00 ff ff 00 10
    75851... 1 104 Rx 6 00 00 00 ff tt tt
    75851... F.. 12t Rx 72 00 00 00 00 00 ff 59 87 21 4c 0...
    75851... 1 105 Rx 6 02 00 04 00 ff tt
    75836... F.. 1 Rx 16 10 7d 18 28 20 20 00 20 20 00 0...
    75836... F.. 2 Rx 16 ca 76 f9 27 20 00 20 20 00 00 0...
    75836... F.. 3 Rx 16 6a 7d fa 27 20 00 20 20 00 00 0...
    75836... F.. 4 Rx 16 a3 76 0e 28 20 00 20 20 00 00 0...
    75836... F.. 5 Rx 16 2c f7 1f 00 02 20 ff 21 22 22 07 f...
    75851... F.. 7 Rx 16 6d fa 76 12 75 62 ff 70 f2 01 0...
    75851... F.. 12 Rx 0

    75851... 2 301 AVL_STEA_DV Rx 7 $1 15 f8 7e ff 7e 11
    75851... 3 301 AVL_STEA_DV Rx 7 $1 15 ff 7e ff 7e 11
    75851... 4 137 Rx 2 fd 00
    75851... 3 d9 AMG_ACDF Rx 8 9b 99 00 c0 00 e0 7f f0
    75849... 1 299 Rx 5 ff ff ff ff
    75851... F.. 21 Rx 0
    75851... 4 45 TORQ_CSRH_1 Rx 8 45 f5 48 27 7e 00 00 ff
    75851... 3 45 TORQ_CSRH_1 Rx 8 fe ff 48 27 7e 00 00 ff
    75836... F.. 23 Rx 16 7c 10 05 05 ea f3 53 20 74 10 20 f...
    75851... F.. 26 Rx 0

```

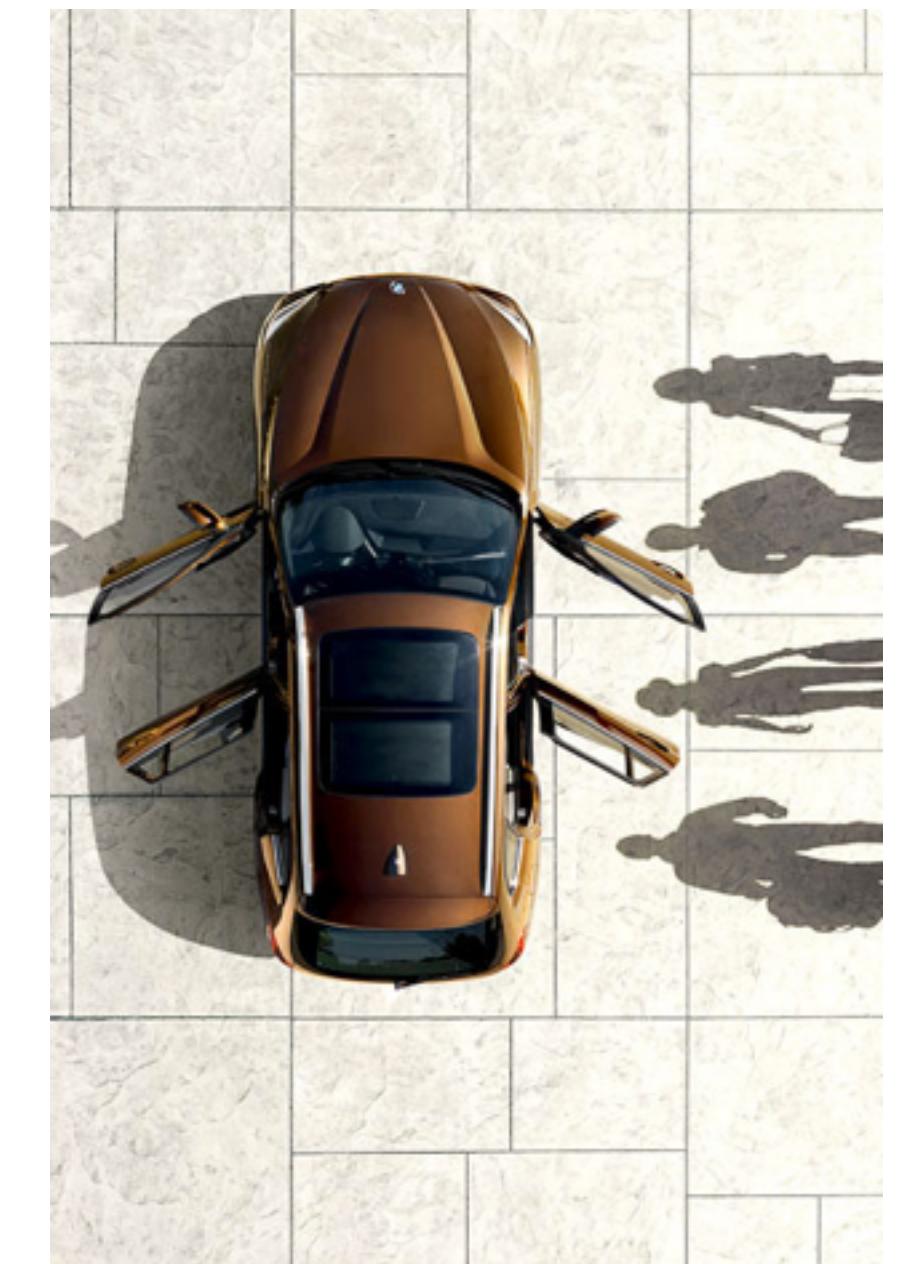
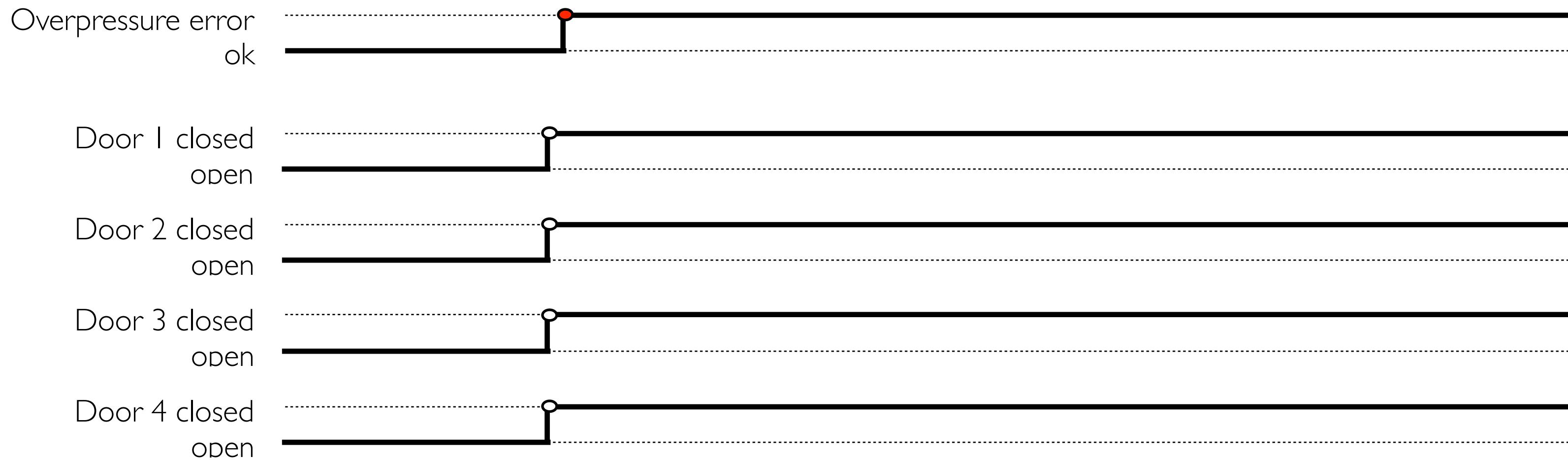
```

... F... 55 Rx 16 9a 77 10 a7 00 20 20 20 20 01 00 0...
... F... 56 Rx 16 ee f8 27 00 20 00 20 20 00 00 0...
... F... 57 Rx 16 ed 77 10 28 00 20 00 20 20 00 0...
... F... 58 Rx 16 5e f8 ff 00 02 20 ff c1 22 22 07 f...
... F... 5a Rx 16 33 08 77 22 75 d6 e7 20 f2 03 0...
... l 1al V_VEH Rx 5 c1 07 00 08a
EH_COG 0 km/h [ 0 ]
ECU_V_VEH Signal ungültig [ f ]
V_VER_COG Signalthwert ist gültig, Zustand/S [ a ]
_V_VEH Fahrtzeug steht [ 0 ]
_V_VEH 193 [ c1 ]
_V_VEH ? [ 7 ]
... F... 5c Rx 8 00 00 00 00 ff ff 00 10
... l 1c4 Rx 6 00 00 00 ff ff
... F... 12f Rx 72 00 00 00 00 00 ff 59 87 21 4c 0...
... l 1c5 Rx 6 02 00 04 00 ff ff
... F... 1 Rx 16 10 7d 18 28 00 20 00 20 20 02 00 0...
... F... 2 Rx 16 ca fe 79 27 00 20 00 20 20 02 00 0...
... F... 3 Rx 16 6d 7d fe 27 00 20 00 20 20 02 00 0...
... F... 4 Rx 16 a3 76 0e 28 00 20 00 20 20 02 00 0...
... F... 5 Rx 16 2c 77 10 02 00 20 ff 21 22 22 07 f...
... F... 7 Rx 16 6d f9 76 12 05 2d 6f 20 f2 01 0...
... F... 12 Rx 0
... 2 301 AVL_STEA_DV Rx 7 51 15 ff 87 ff ff 11
... 3 301 AVL_STEA_DV Rx 7 51 15 ff 87 ff ff 11
... 4 137 Rx 2 fd 00
... 3 49 ANG_ACPD Rx 8 9b 99 00 00 e0 0f f0
... 1 299 Rx 5 ff ff ff ff ff
... F... 21 Rx 0
... 4 a5 TORQ_CRSH_1 Rx 8 45 e5 48 f7 7f 00 00 fc
... 3 a5 TORQ_CRSH_1 Rx 8 fe f9 48 f7 7f 00 00 fc
... F... 23 Rx 16 7c 10 05 05 ea f3 53 20 74 10 20 f...
... 26 Rx 0

```

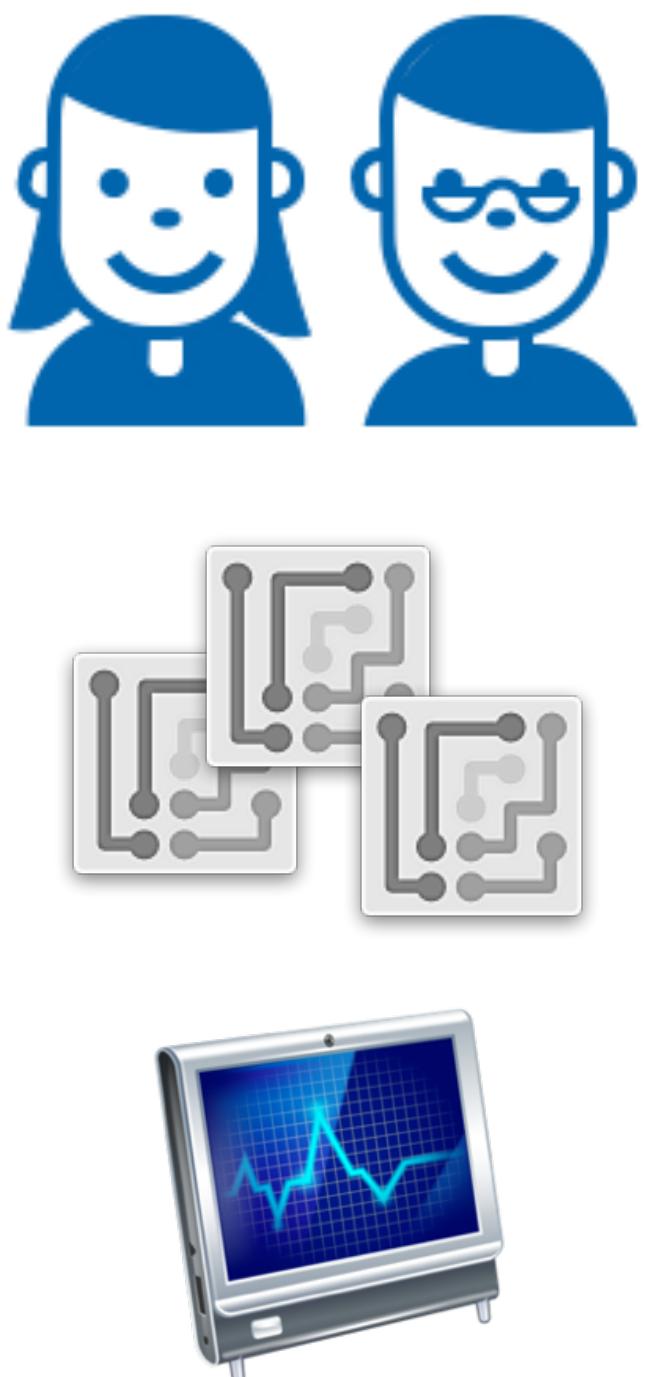
Results

- externalization and sharing of expert knowledge
- complete coverage of traces vs. sparse sampling
- understand behavioral correlations



<http://www.automotiveaddicts.com/5276/bmw-x1-officially-released-images-first-drive-videos/bmw-x1-top-doors-open>

Example of: **human-centered data science**

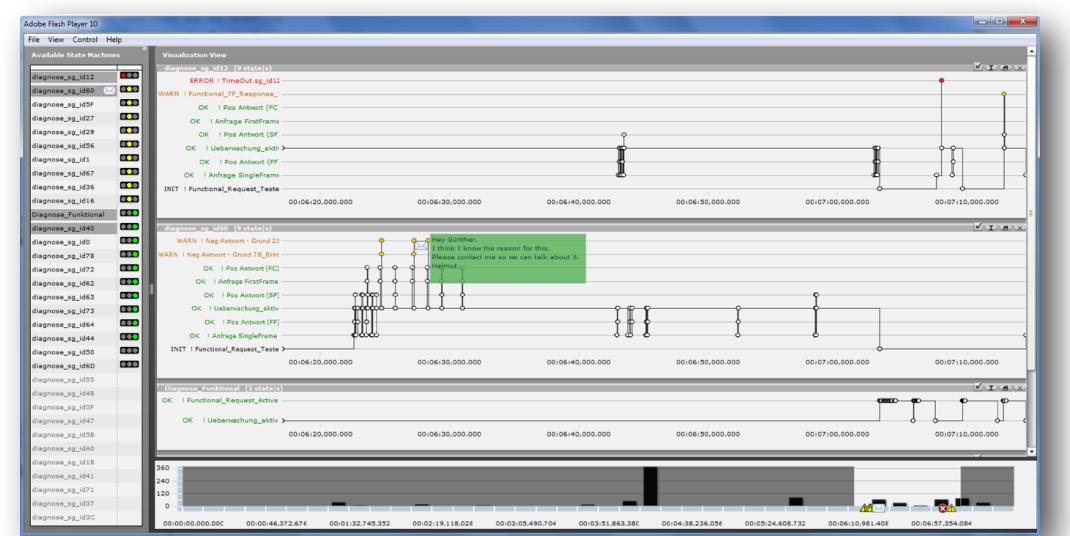
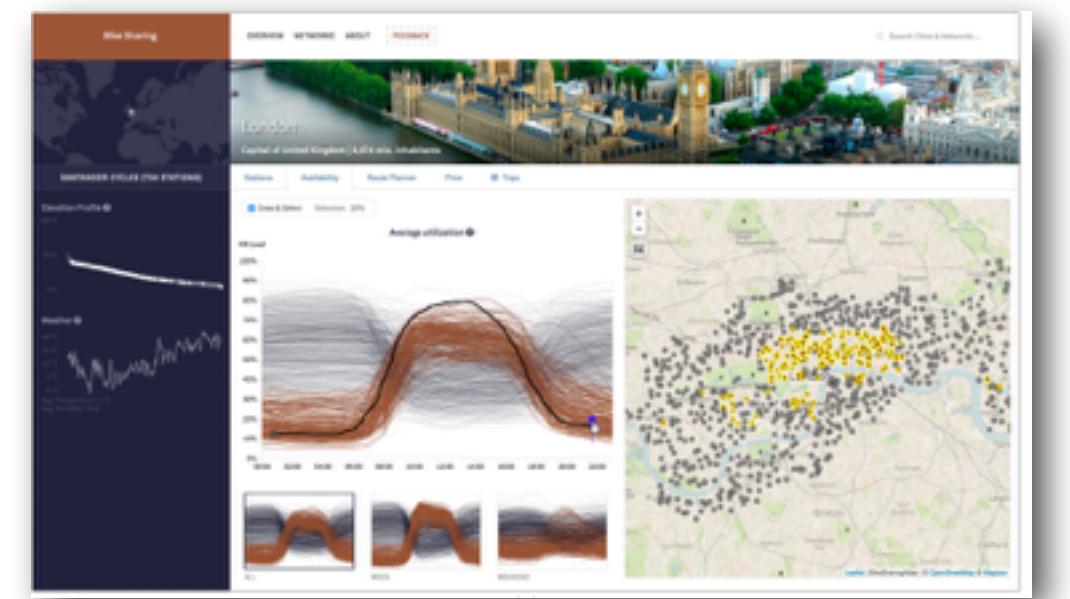
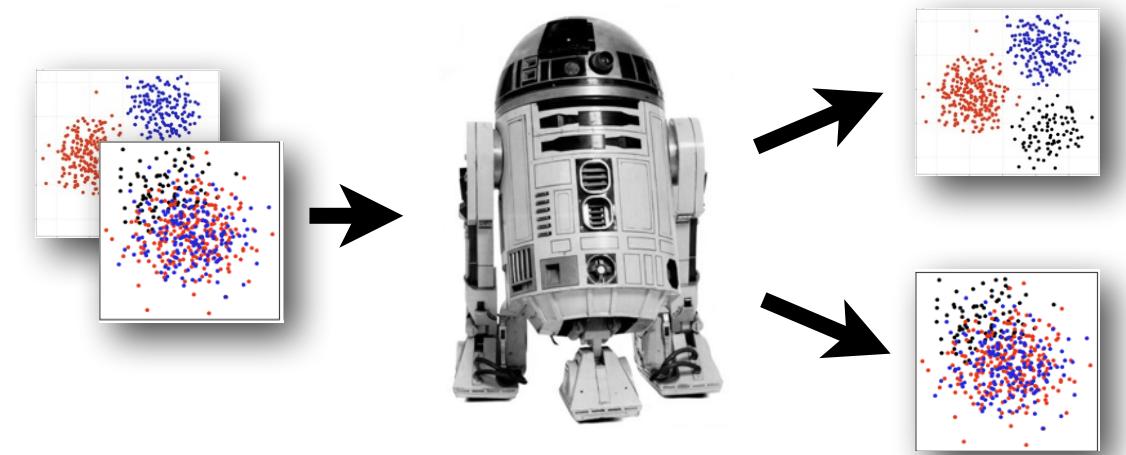


**deployed
and
adopted**

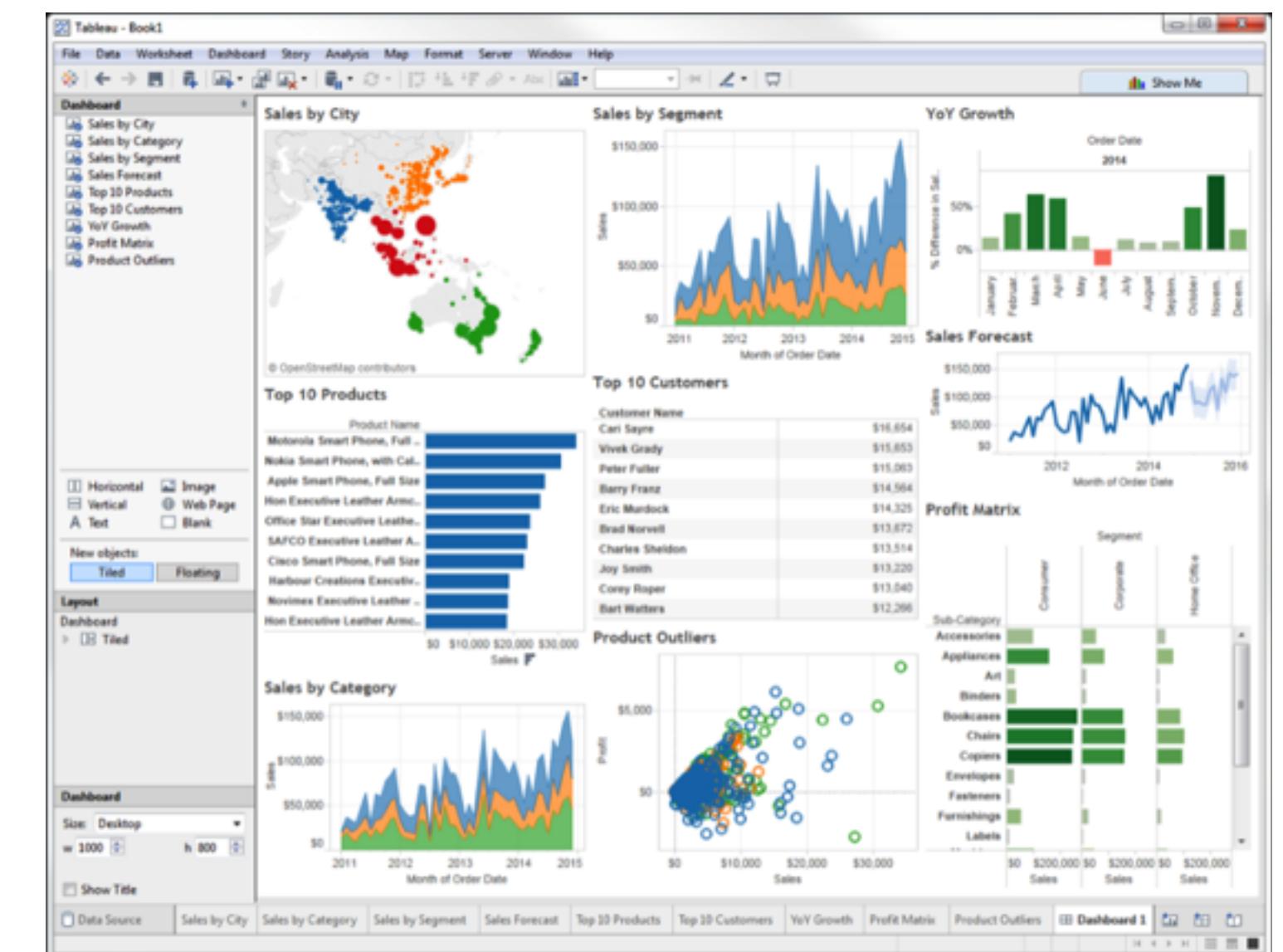
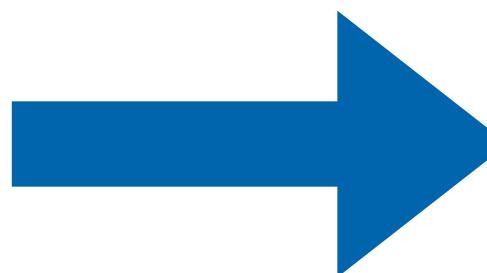
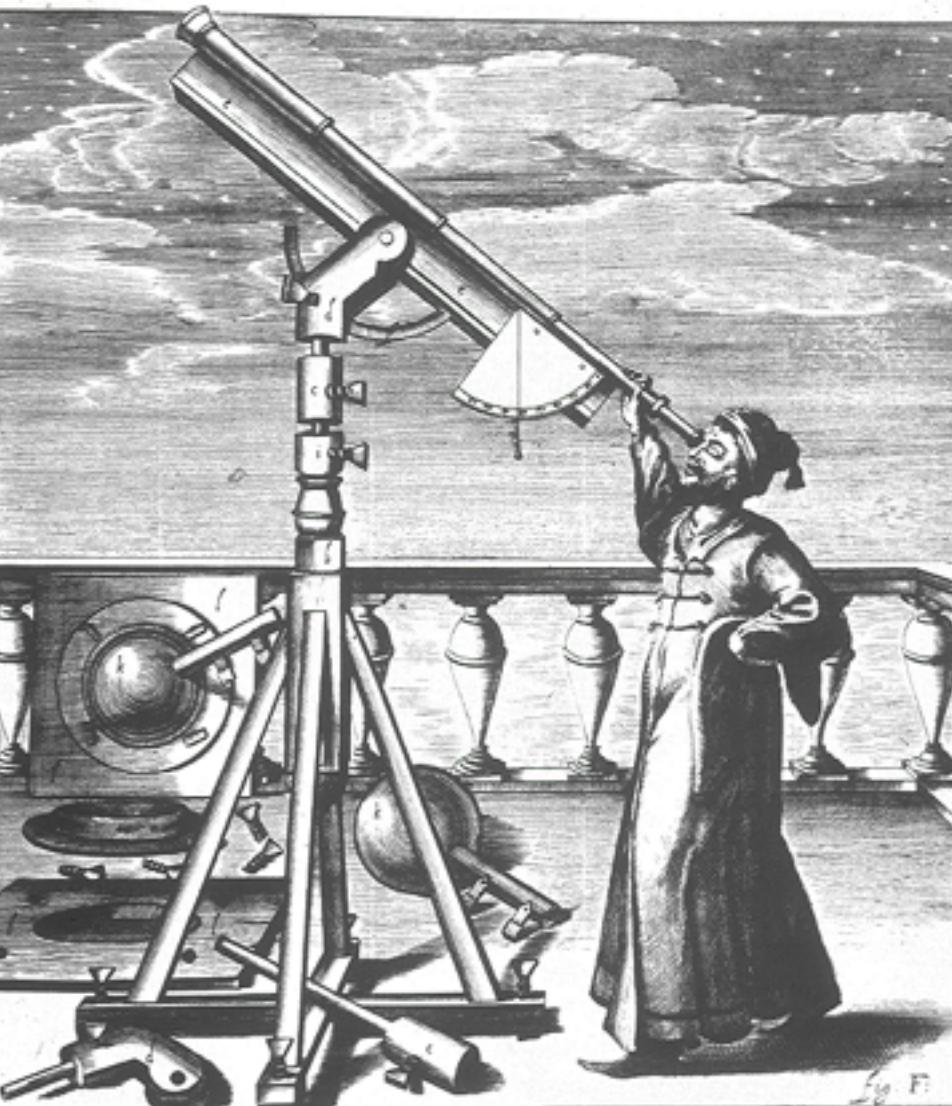
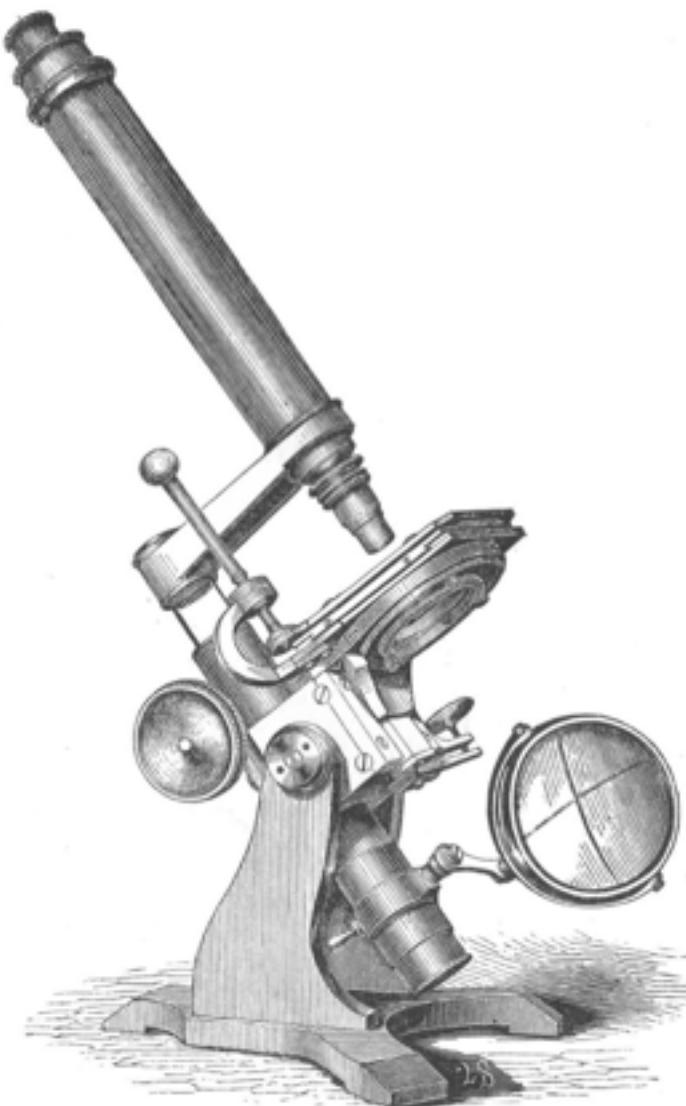
Summary

Summary: How to visualize data?

- Leverage perception
- Make it interactive
- Combine computation and visualization



Visualization: The Human Lens to Data



Tableau



Thank you!

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michael.sedlmair@univie.ac.at
slides: <http://tinyurl.com/siat2017>



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