

# Visualization and Big Data in Official Statistics

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Statistics  
Netherlands

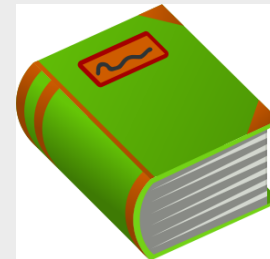
# From a Official Statistics point of view

Three types of data:

1. Survey data = data collected by SN with questionnaires



2. Admin data = administrative (register) data collected by third parties such as the Tax Office



3. Big data = machine generated data of events

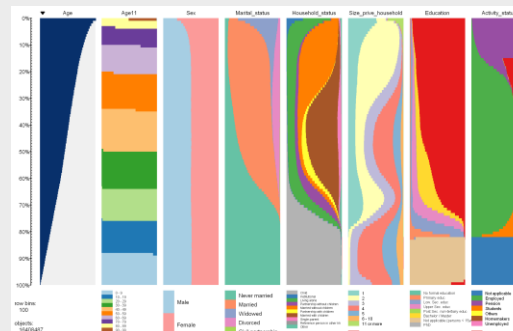


# Big Data case studies

Big data = machine generated data of events

Source	Statistics
Social media	Sentiment (as indicator for business cycle)
Mobile phone metadata	Daytime population, tourism statistics
Road sensors	Traffic index statistics

At the end of this talk:  
Visualization methods for Big Data



# Big data approach

## General Data Science workflow



No privacy issues anymore!

### Data jiu jitsu:

- Editing
- Restructuring
- Transforming
- Combining
- Filtering
- Aggregating
- ...



### Values:

- Totals
- Mean values
- Scores
- ....

### Dimensions:

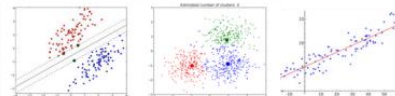
- x, y, time
- from, to, time
- location type, time
- ...



**Data Cubes**



### Modeling and estimating

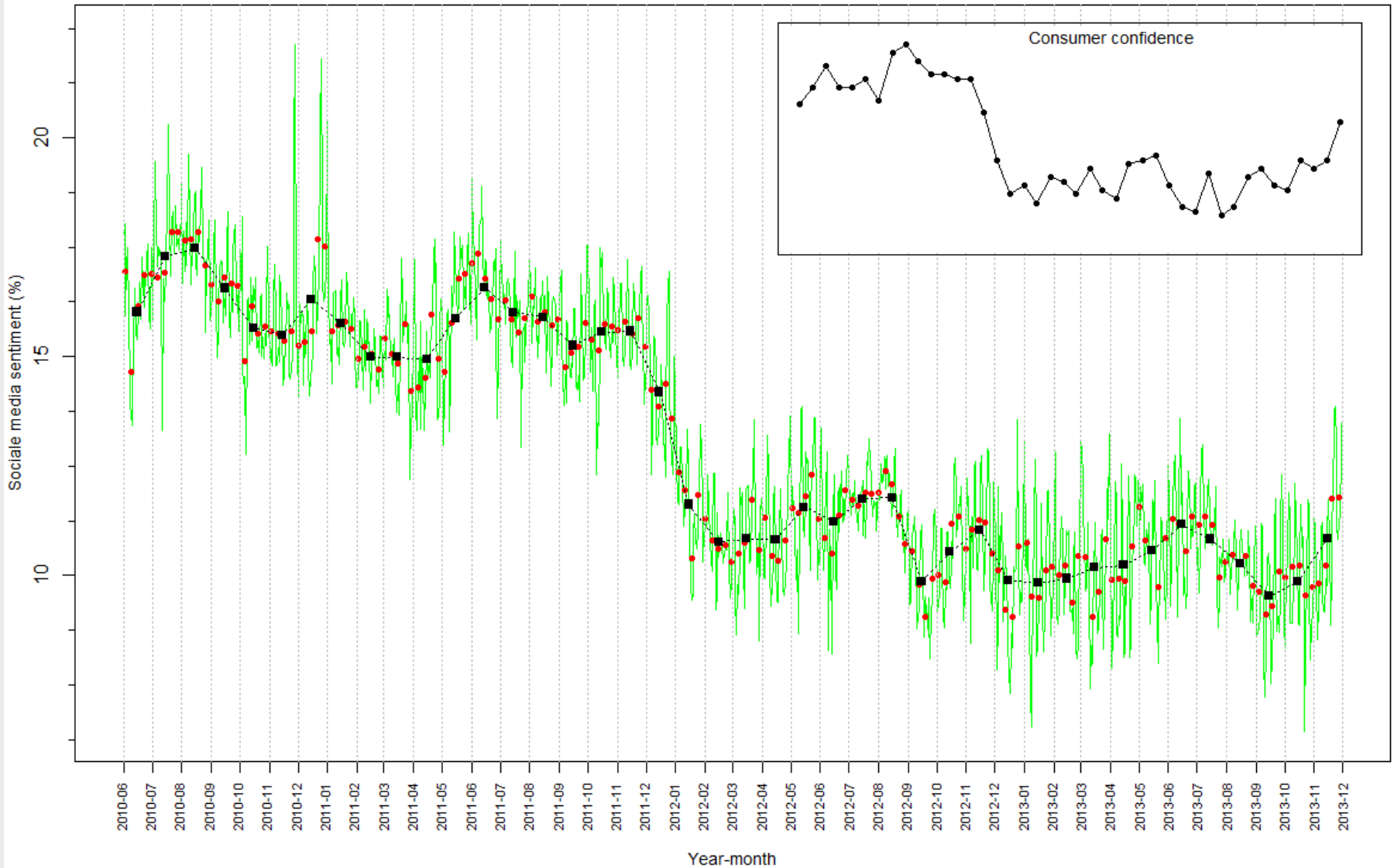


**Estimations**

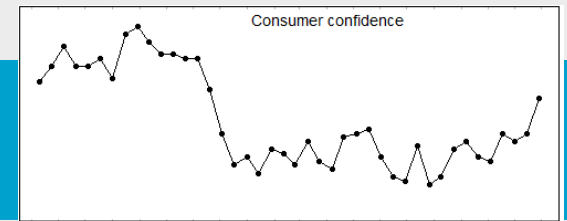
# Case study 1: Social media

- 3 billion messages as of 2009 gathered from Facebook, Twitter, LinkedIn, Google+ by a Dutch intermediate company Coosto.
- Sentiment per message determined by classifying words as negative or positive.
- Could be used as indicator for the business cycle. Could it be fit to the **consumer confidence**, the leading business cycle indicator?

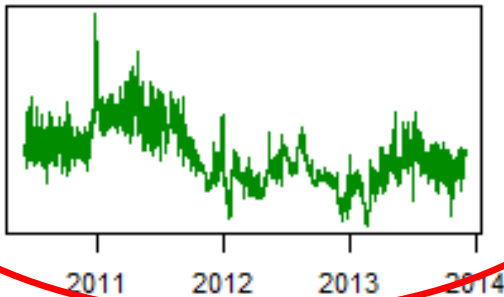
# Sentiment in social media



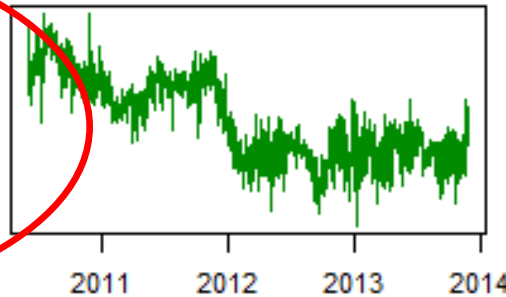
# Platform specific sentiment



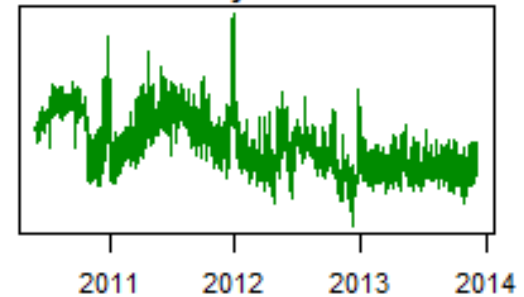
Facebook



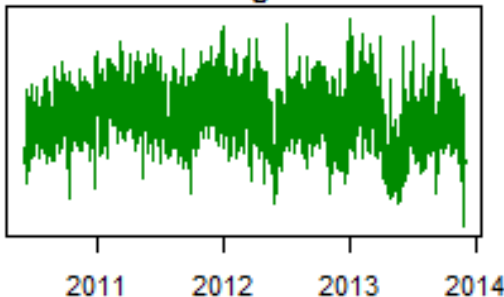
Twitter



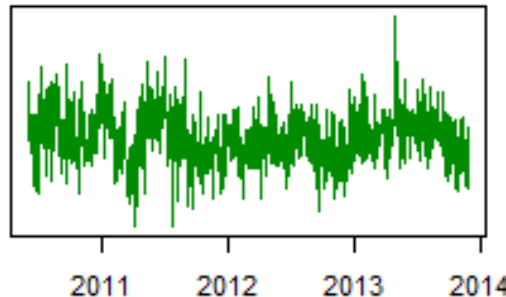
Hyves



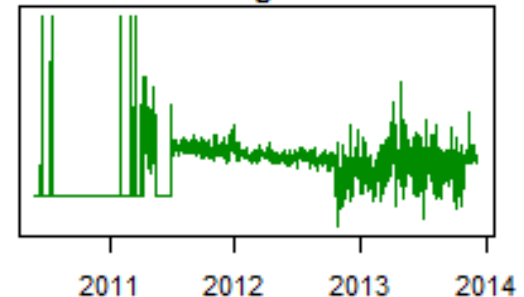
Blogs



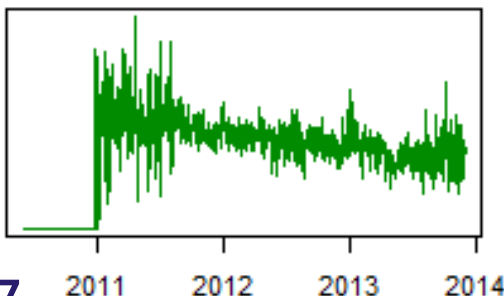
News sites



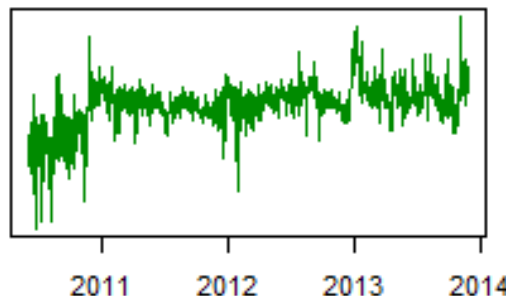
Google+



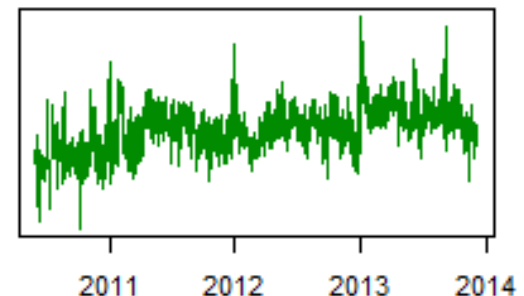
LinkedIn



Youtube



Forums



# Platform specific results

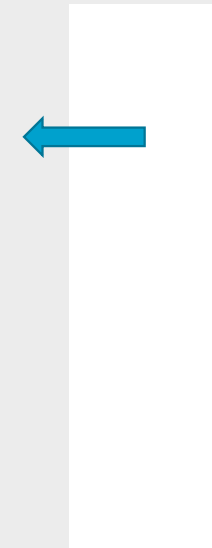
**Table 1.** Social media messages properties for various platforms and their correlation with consumer confidence

Social media platform	Number of social media messages <sup>1</sup>	Number of messages as percentage of total (%)	Correlation coefficient of monthly sentiment index and consumer confidence ( $r$ ) <sup>2</sup>
All platforms combined	3,153,002,327	100	0.75
Facebook	334,854,088	10.6	0.81*
Twitter	2,526,481,479	80.1	0.68
Hyves	45,182,025	1.4	0.50
News sites	56,027,686	1.8	0.37
Blogs	48,600,987	1.5	0.25
Google+	644,039	0.02	-0.04
Linkedin	565,811	0.02	-0.23
Youtube	5,661,274	0.2	-0.37
Forums	134,98,938	4.3	-0.45

<sup>1</sup>period covered June 2010 until November 2013

<sup>2</sup>confirmed by visual inspecting scatterplots and additional checks (see text)

\*cointegrated

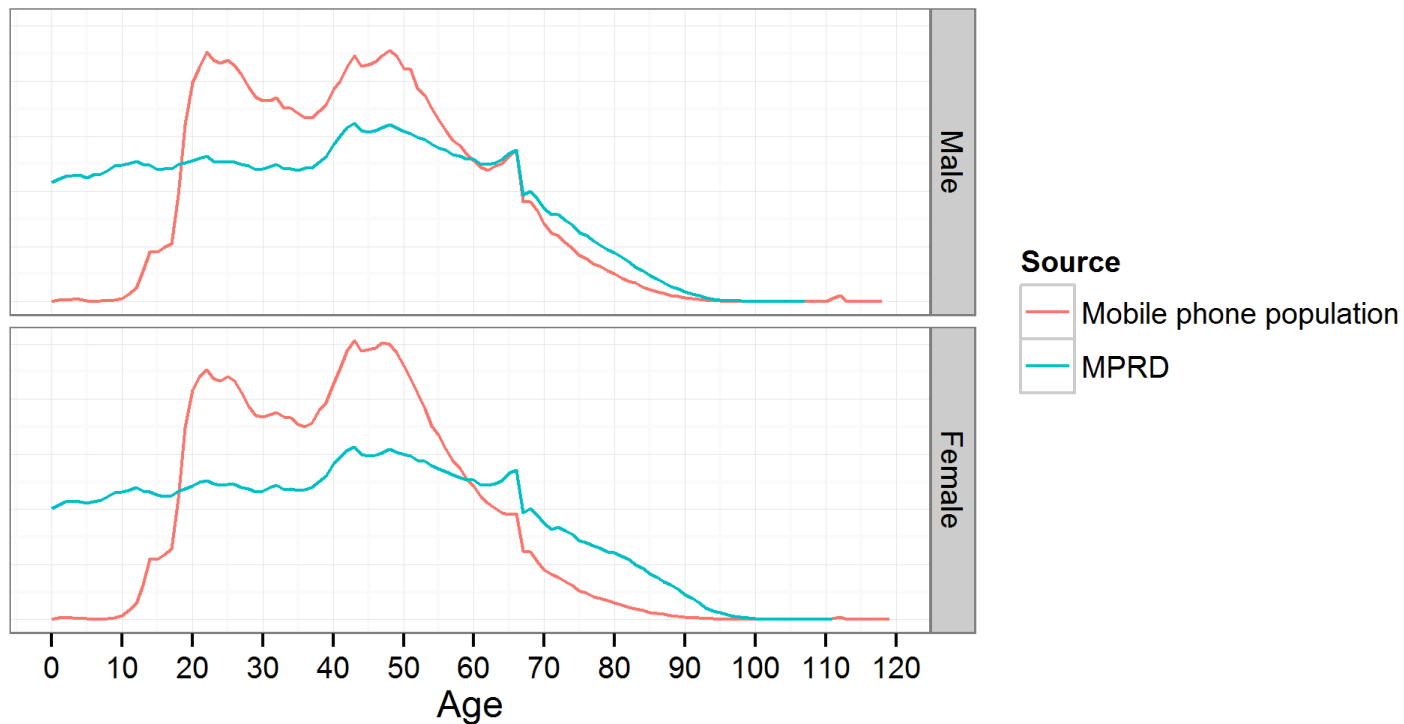




# Case study 2: mobile phone metadata

- Pilot study with Vodafone, a provider with market share of 1/3 in the Netherlands.
- Aggregated data is queried by intermediate company Mezuro and delivered to SN. Privacy is guaranteed!
- Applications: daytime population, tourism statistics, economic activity, mobility studies, etcetera.

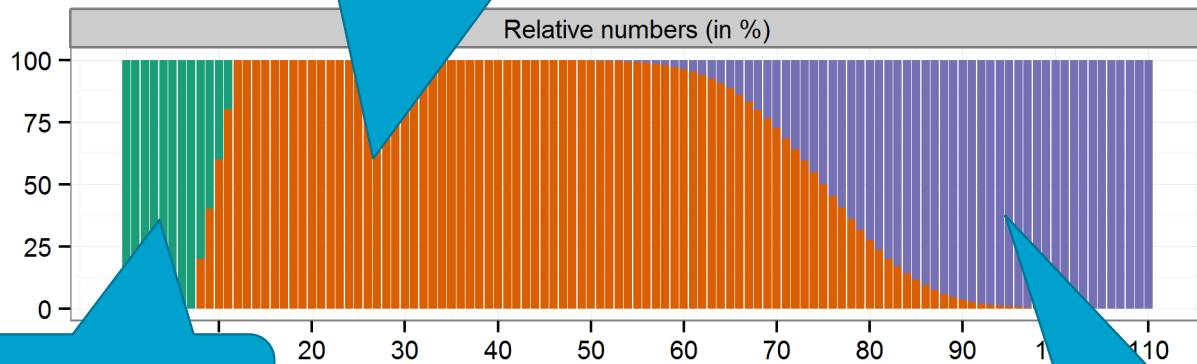
# Mobile phone population



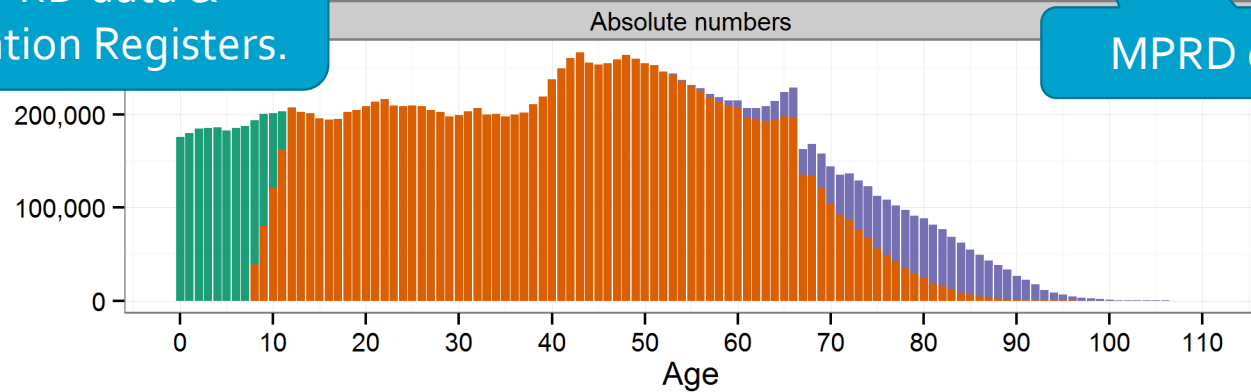
MPRD (Municipal Personal Records Database) = Dutch population

# Subpopulations model

Mobile phone metadata weighted to the MPRD.



MPRD data & Education Registers.



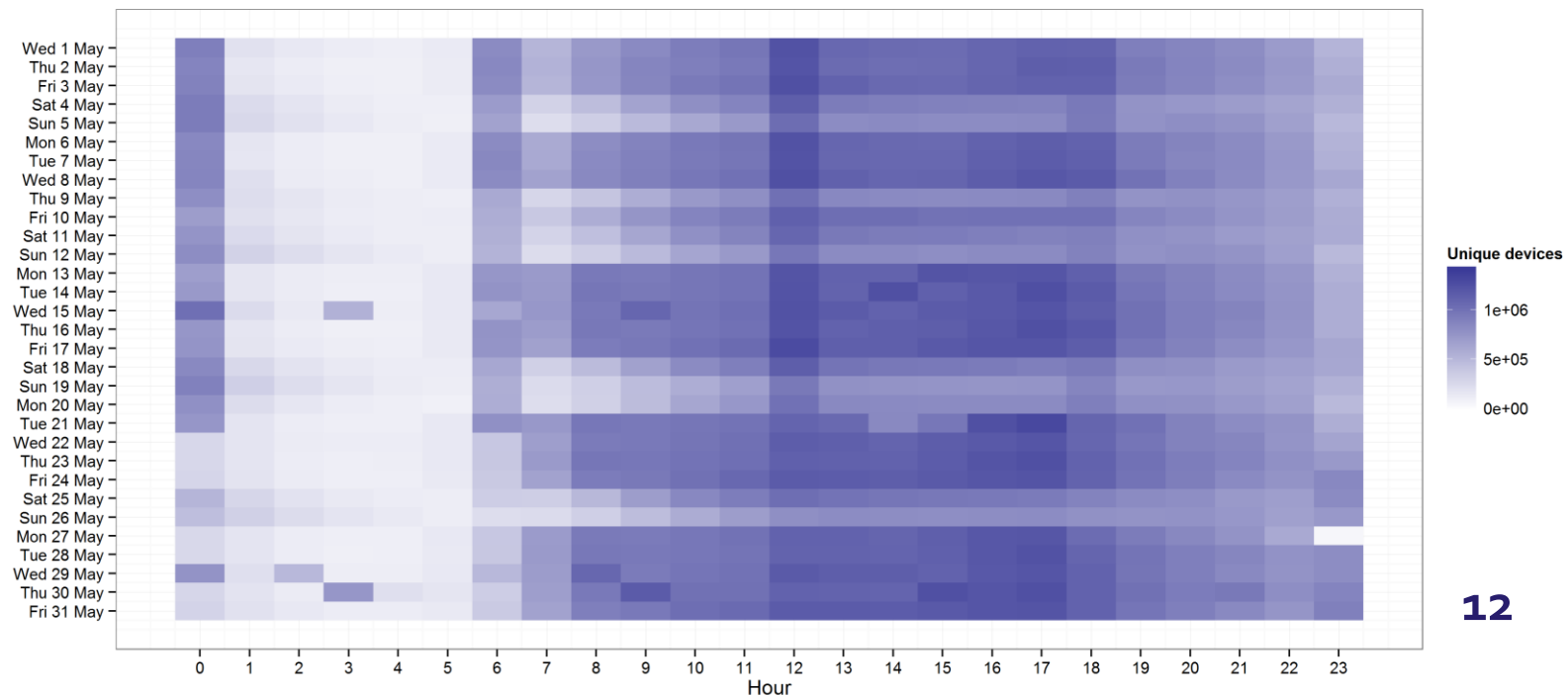
MPRD data only.

Children without mobile phone People with mobile phone Elderly people without mobile phone

# Mobile phone metadata

Event Detail Records (EDR) contain metadata on mobile phone events (i.e. call, SMS or data transfer).

Aggregated table: number of unique devices X time period X current region X residential region.



# Weighting method

Example: suppose there are only 3 regions in the Netherlands: [Amsterdam](#), [Boskoop](#) and [Castricum](#)

	Residence				
		Amsterdam	Boskoop	Castricum	
Current region at time $t$	Amsterdam	199,000	1,000	4,000	
	Boskoop	500	3,500	0	
	Castricum	500	500	16,000	

# Weighting method (2)

Example: suppose there are only 3 regions in the Netherlands: [Amsterdam](#), [Boskoop](#) and [Castricum](#)

	Residence				
		Amsterdam	Boskoop	Castricum	
Current region at time $t$	Amsterdam	199,000	1,000	4,000	
	Boskoop	500	3,500	0	
	Castricum	500	500	16,000	
	<b>MPRD total</b>	<b>800,000</b>	<b>15,000</b>	<b>30,000</b>	

# Weighting method (3)

Example: suppose there are only 3 regions in the Netherlands: [Amsterdam](#), [Boskoop](#) and [Castricum](#)

	Residence				
		Amsterdam	Boskoop	Castricum	
Current region at time $t$	Amsterdam	796,000	3,000	6,000	
	Boskoop	2000	10,500	0	
	Castricum	2000	1,500	24,000	
	<b>MPRD total</b>	<b>800,000</b>	<b>15,000</b>	<b>30,000</b>	

# Weighting method (4)

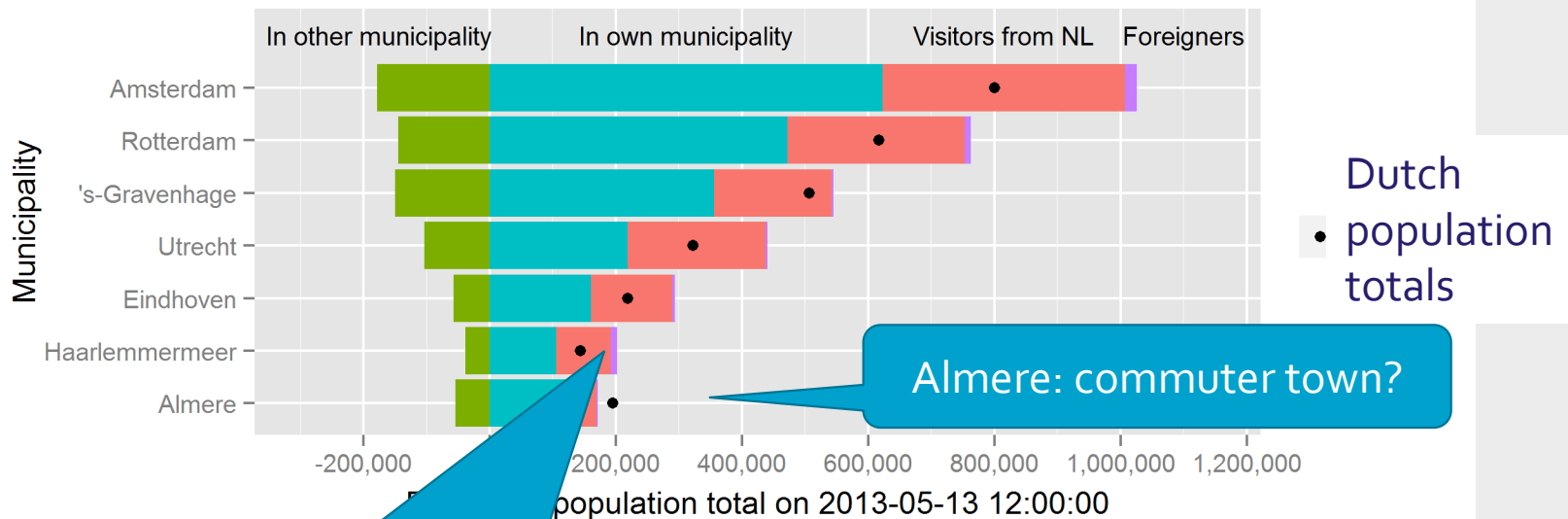
Example: suppose there are only 3 regions in the Netherlands: [Amsterdam](#), [Boskoop](#) and [Castricum](#)

	Residence				
		Amsterdam	Boskoop	Castricum	DTP total
Current region at time $t$	Amsterdam	796,000	3,000	6,000	805,000
	Boskoop	2000	10,500	0	12,500
	Castricum	2000	1,500	24,000	27,500
	MPRD total	800,000	15,000	30,000	





# Daytime population results



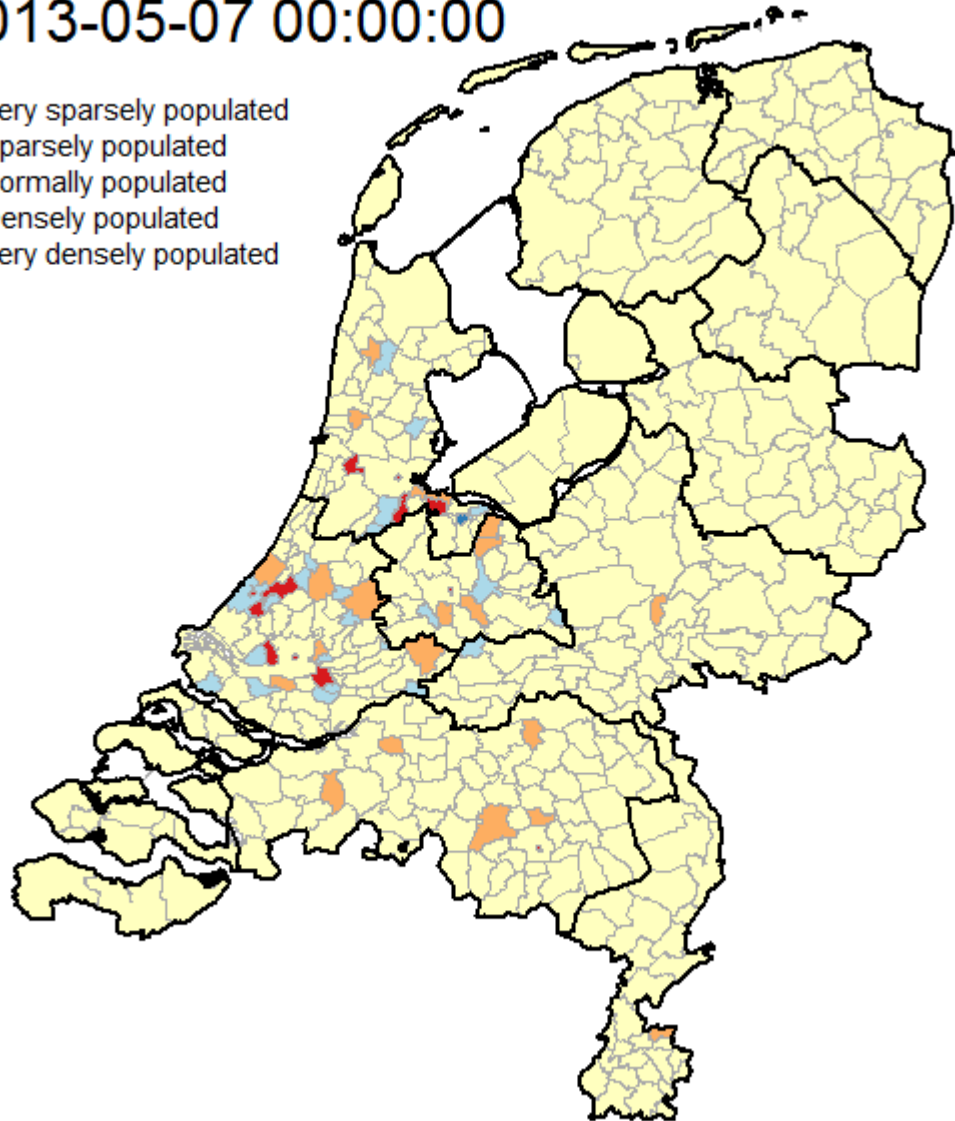
Foreigners at Schiphol Airport



# Day time population (relative)

2013-05-07 00:00:00

- Very sparsely populated
- Sparsely populated
- Normally populated
- Densely populated
- Very densely populated



# Day time population (relative)

0:00



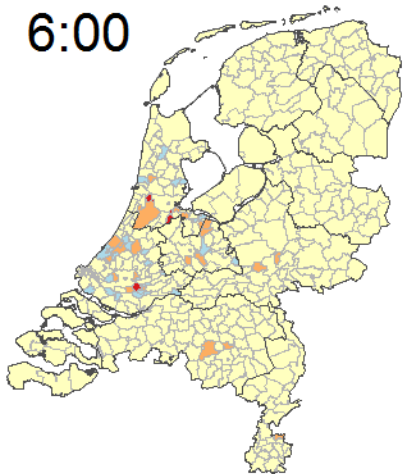
2:00



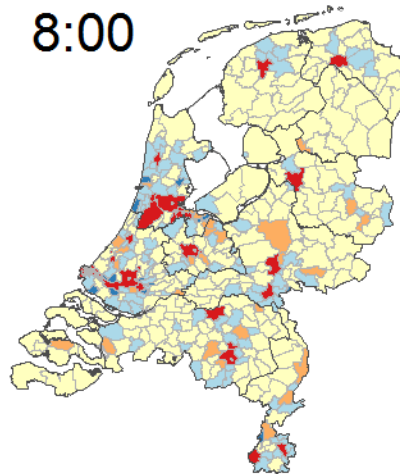
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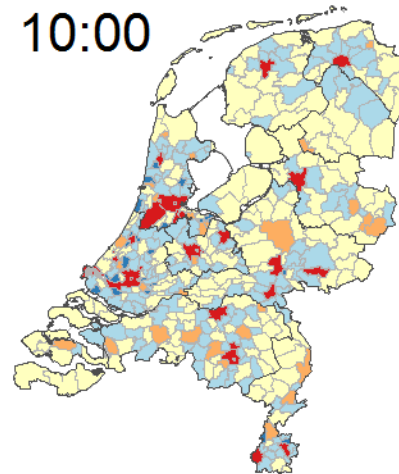
6:00



8:00



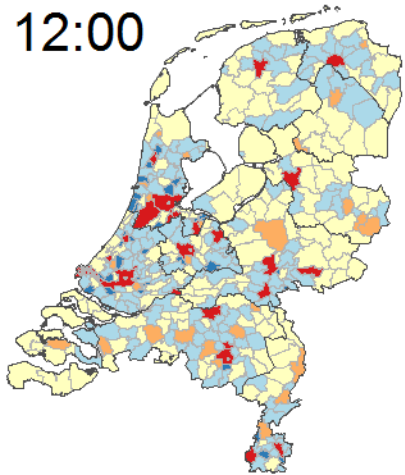
10:00



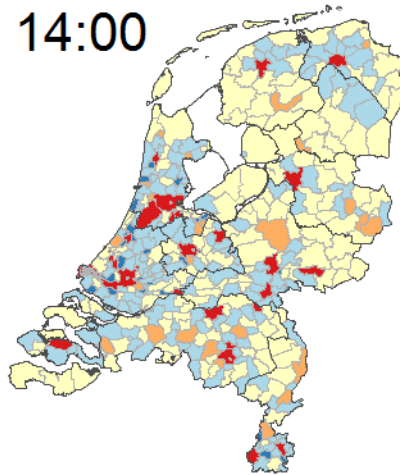
- Very sparsely populated
- Sparsely populated
- Normally populated
- Densely populated
- Very densely populated

# Day time population (relative)

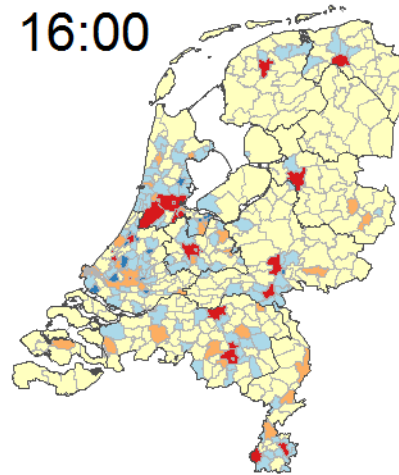
12:00



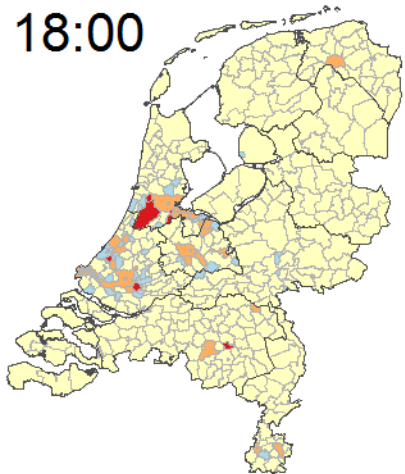
14:00



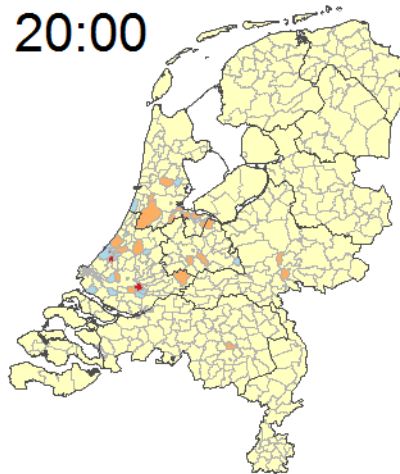
16:00



18:00



20:00



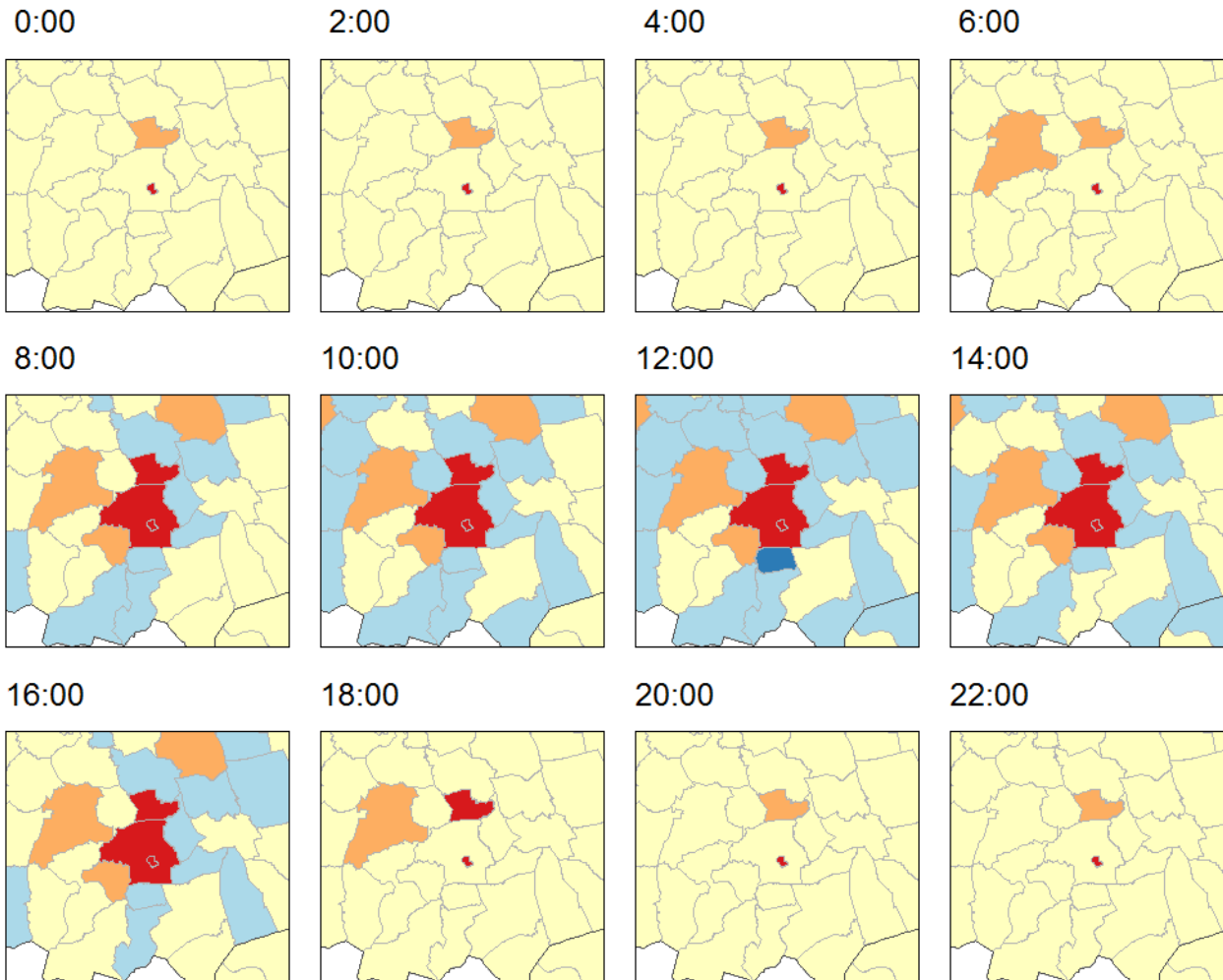
22:00



- Very sparsely populated
- Sparsely populated
- Normally populated
- Densely populated
- Very densely populated

# Day time population (relative)

## City of Eindhoven and surrounding towns



- Very sparsely populated
- Sparsely populated
- Normally populated
- Densely populated
- Very densely populated

# Day time population – Region profile

## K-means clustering

**Work** = daytime vs. night-time during working weeks

**Weekend** = weekends activity

**Holiday** = May holiday activity



# Case study 3: Road sensors



## Road sensors data

- Each minute (24/7) the number of passing vehicles is counted in around 20.000 'loops' in the Netherlands (100 million records a day)
- Nice data source for transport and traffic statistics (and more)



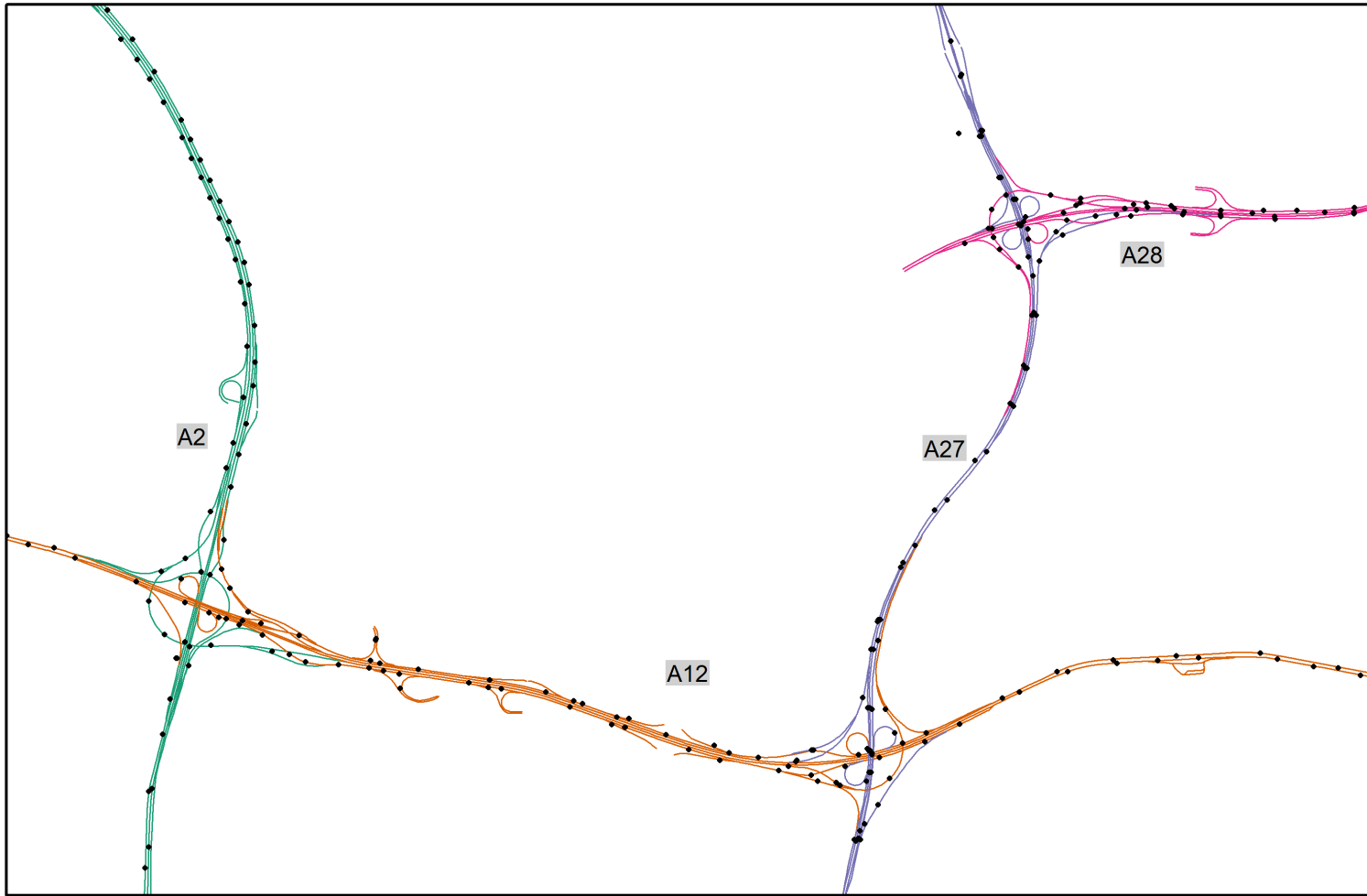
# Road sensors on main roads



A close look at the highways around Utrecht

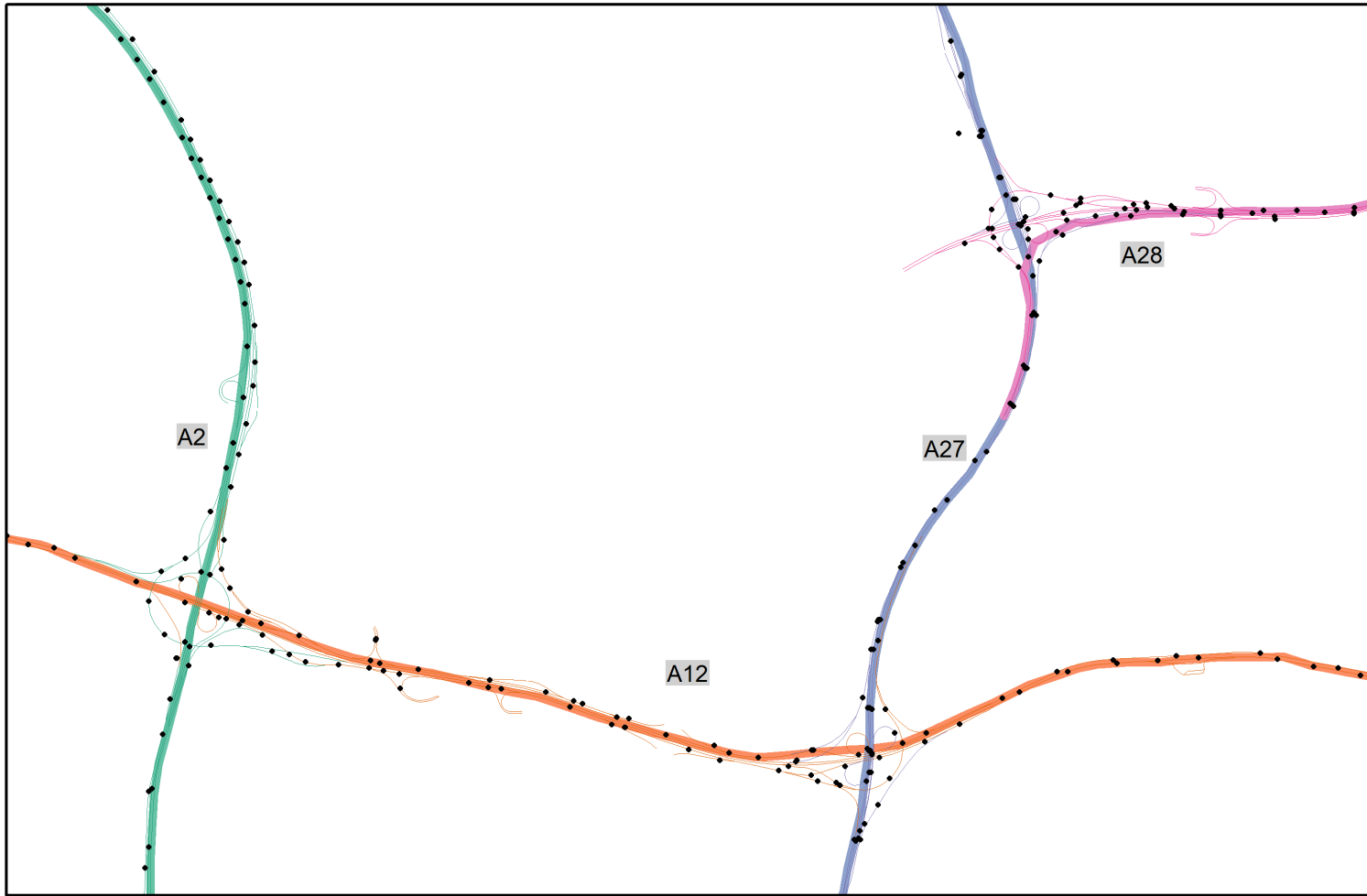


# Road sensors on main roads (2)



Traffic loops everywhere...

# Road sensors on main roads (3)



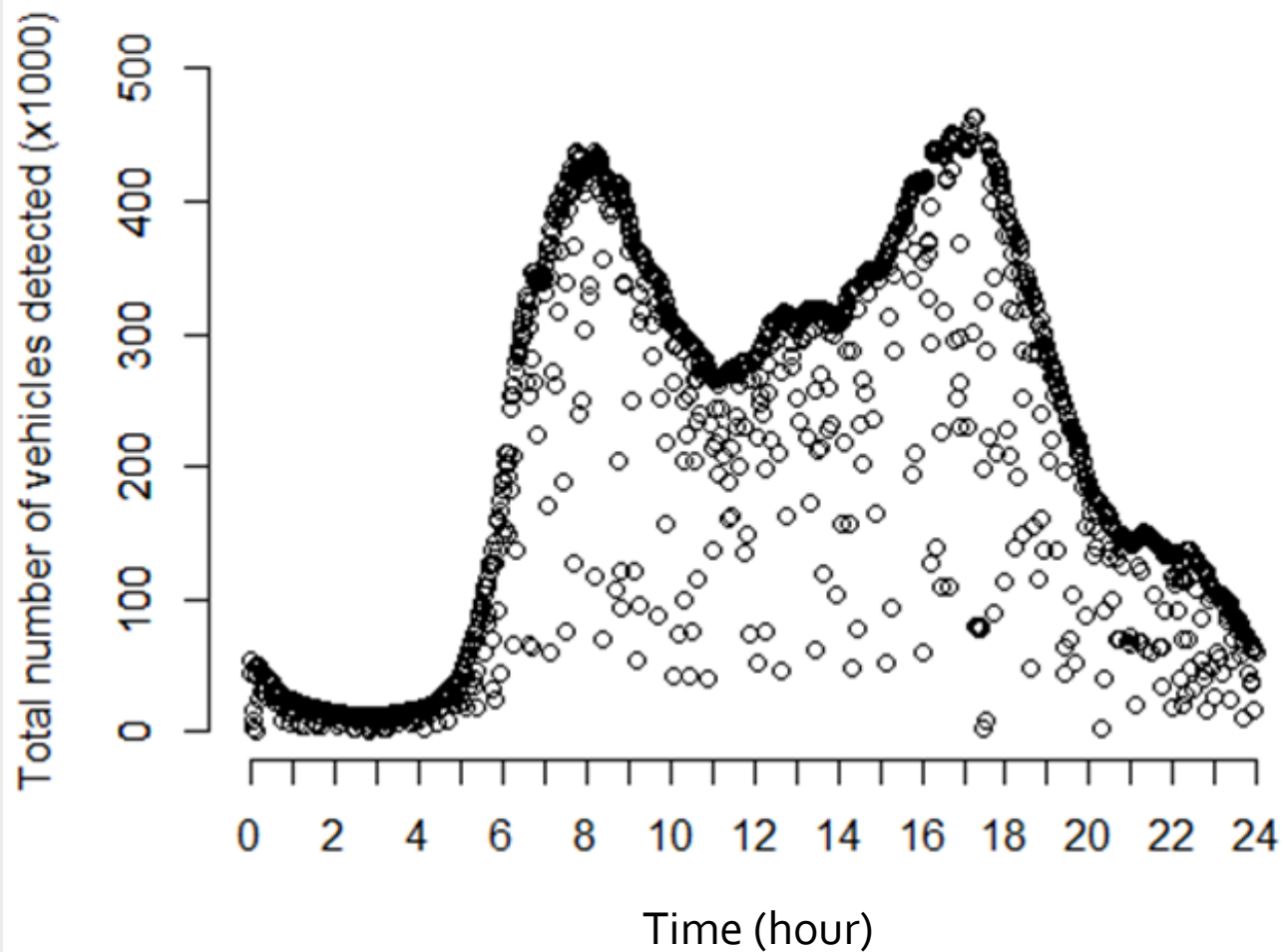
Highways simplified for analysis

# Road sensors on main roads (4)



Dutch highways by COROP region

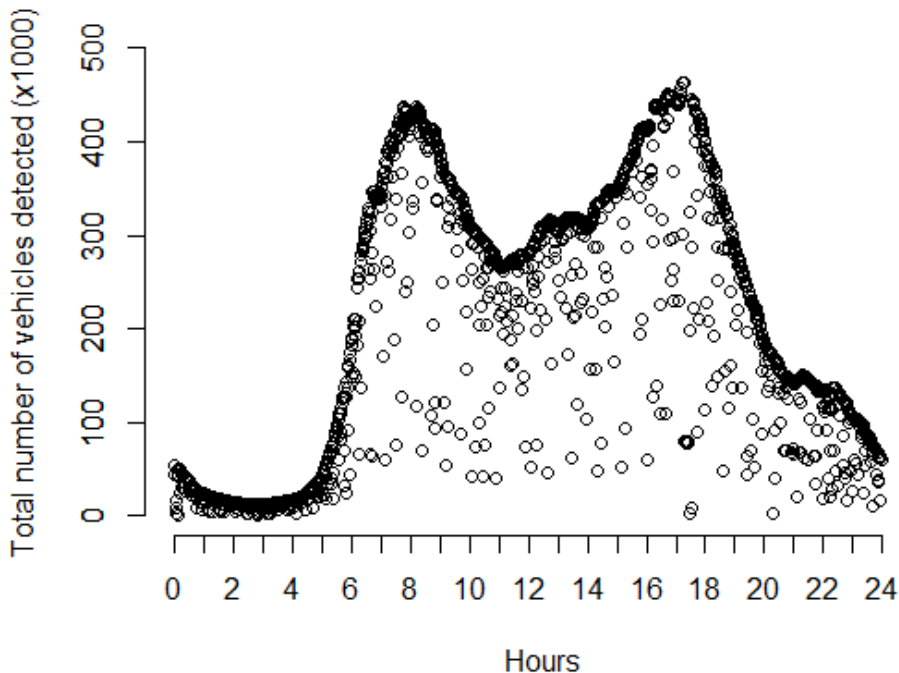
# Raw data: Total number of vehicles a day



# Correct for missing data: macro level

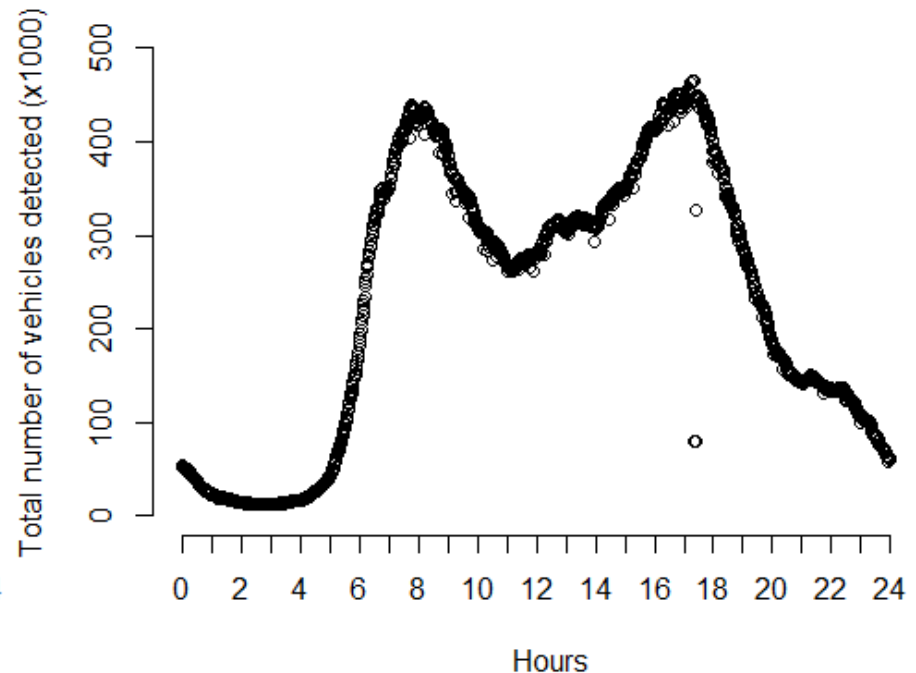
Sliding window of 5 min. Impute missing data.

Before



Total = ~ 295 million detected vehicles

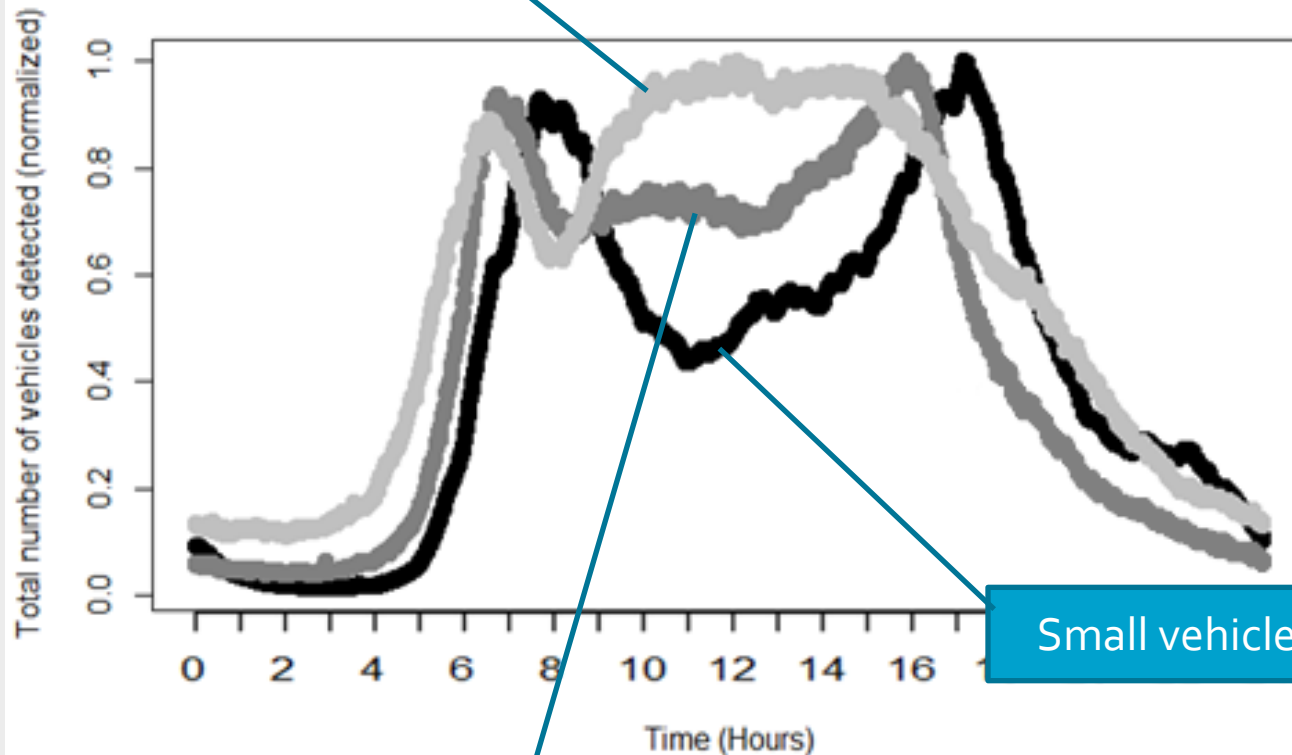
After



Total = ~ 330 million (+ 12%)  
detected vehicles

# Data by type of vehicle

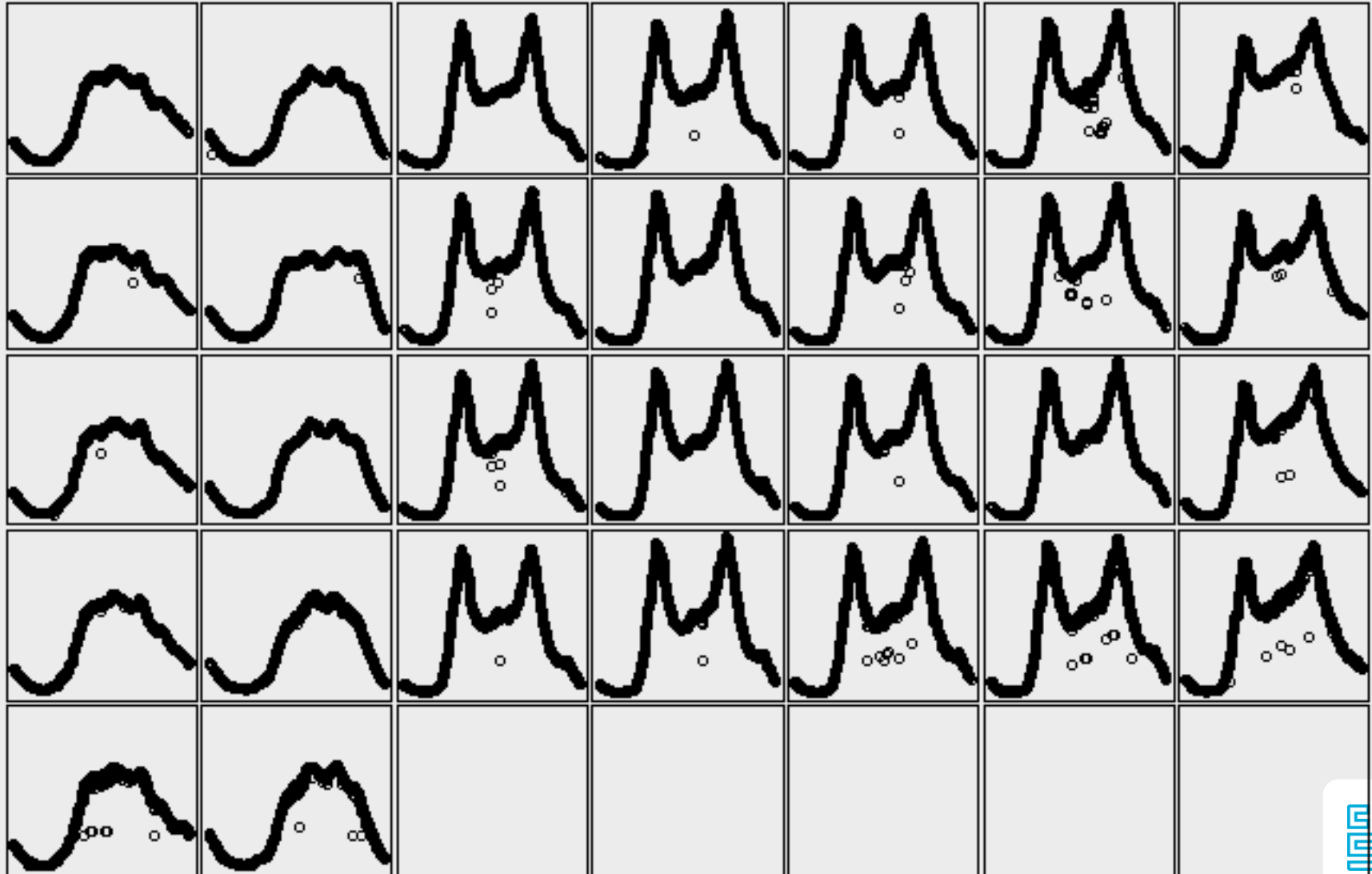
Long vehicles (> 12.2 meter)



Small vehicles (<= 5.6 meter)

Medium vehicles (> 5.6 & <= 12.2 meter)

# All Dutch vehicles in September



# Selectivity of big data

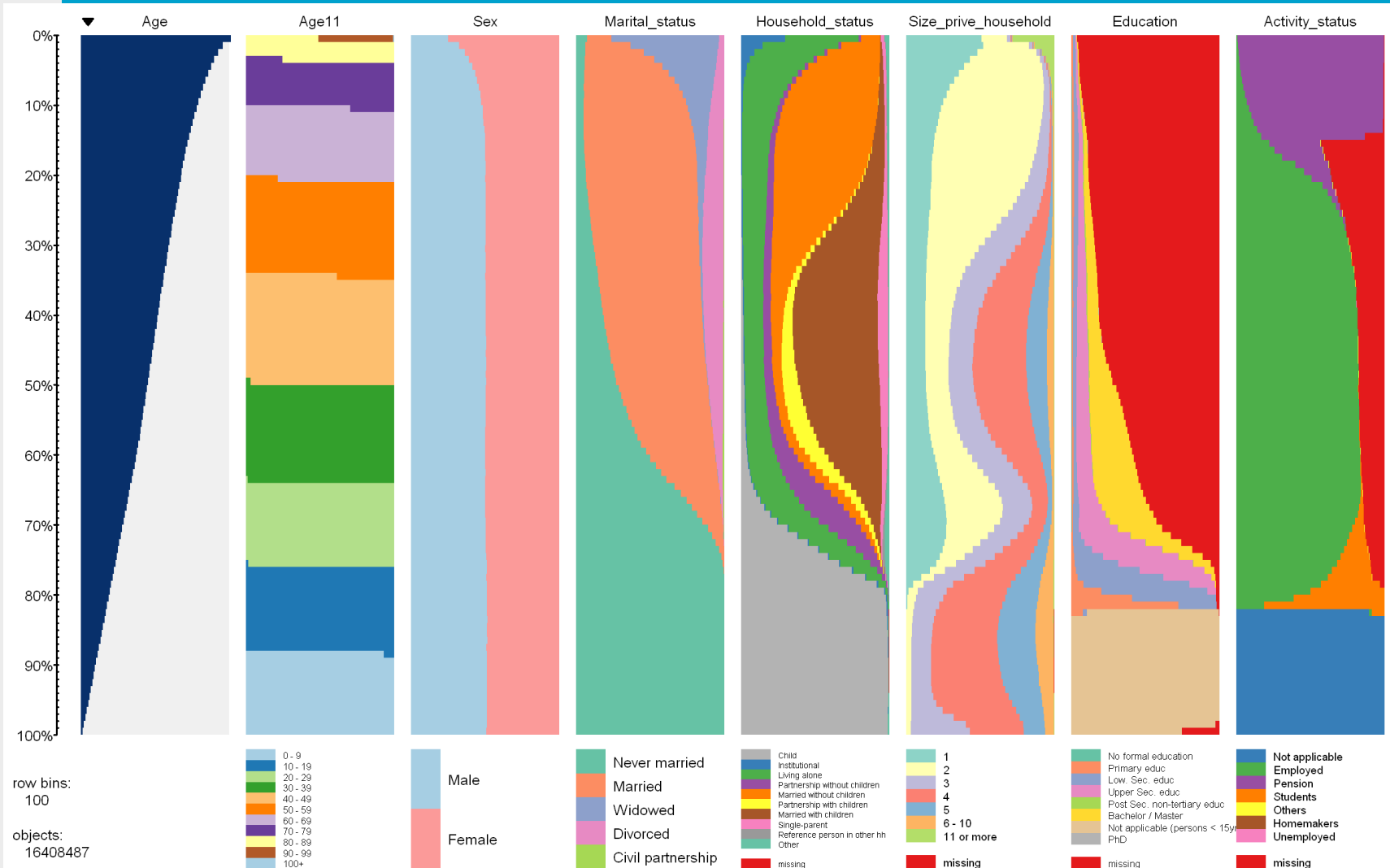
- Big Data sources may be selective when
  - Only **part of the population** contributes to the data set (e.g. **mobile phone owners**)
  - The **measurement** mechanism is **selective** (e.g. **traffic loops placement on Dutch highways is not random**)
- Many Big Data sources contain events
  - How to **associate** events with **units**?
  - Number of events per unit may vary.
- Correcting for selectivity
  - Background characteristics – or *features* – are needed (linking with registers; profiling)
  - Use predictive modeling / machine learning to produce population estimates



# Visualization of Big Data

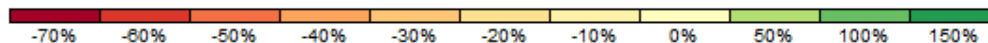
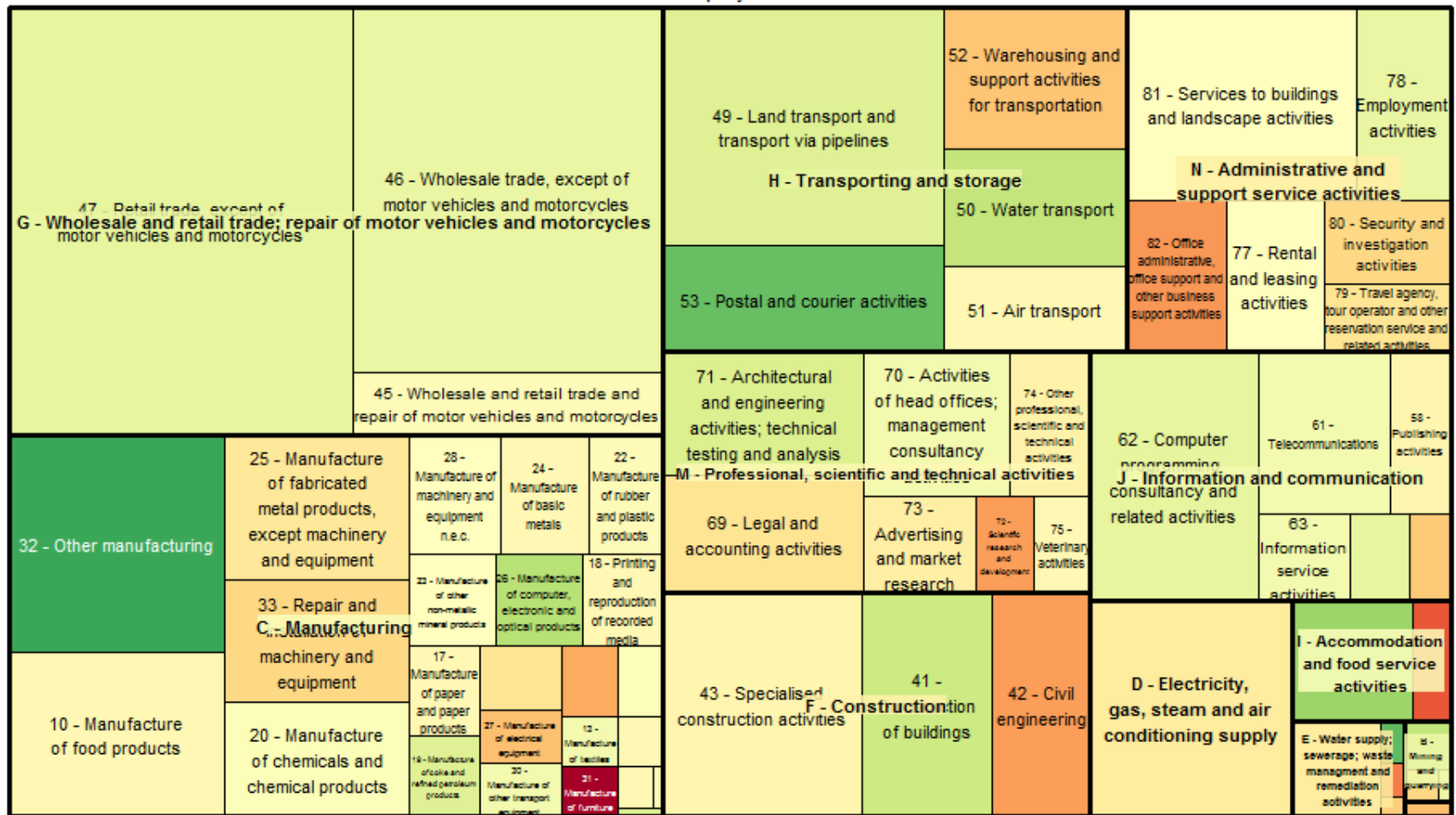
- Large **volume**:
  - Data binning or aggregation
- High **velocity**:
  - Animations
  - Dashboard / small multiples
- Large **variety**:
  - Interactive interface
  - Advanced visualization methods

# Tableplot: Dutch (Virtual) Census



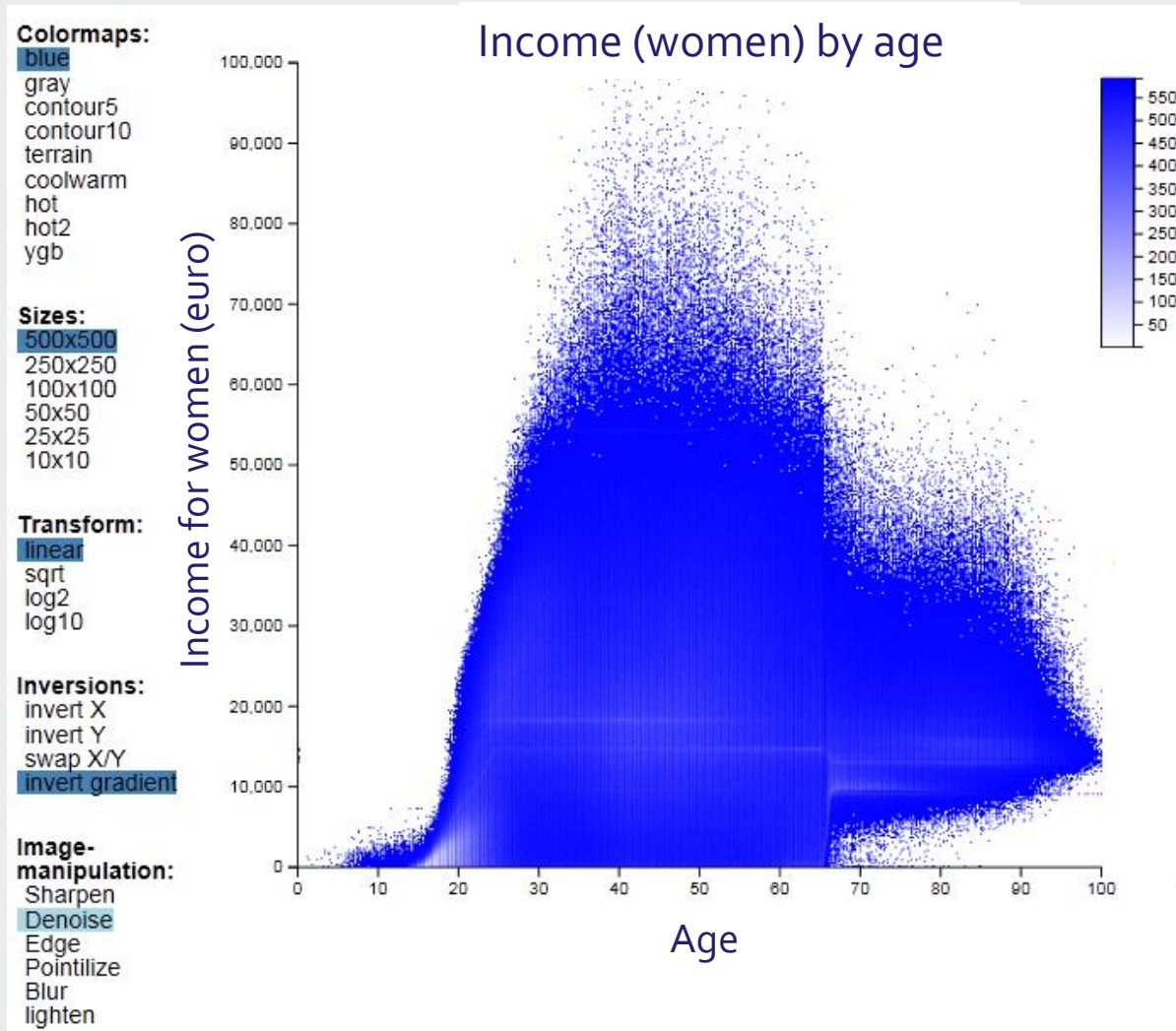
# Treemap: Structural Business Statistics

employees



employees.prev

# Heatmap: Income statistics



# References

Topic	Links
Social Media	Daas, P.J.H., Puts, M.J.H. (2014) Sociale Media Sentiment and Consumer Confidence. Paper for the Workshop on using Big Data for Forecasting and Statistics, Frankfurt, Germany. <a href="http://www.ecb.europa.eu/events/pdf/conferences/140407/Daas_Puts_Sociale_media_cons_conf_Stat_Neth.pdf?409d61b733fc259971ee5beec7cedc61">http://www.ecb.europa.eu/events/pdf/conferences/140407/Daas_Puts_Sociale_media_cons_conf_Stat_Neth.pdf?409d61b733fc259971ee5beec7cedc61</a>
Mobile phone metadata	Paper in progress...
Road sensors	Paper in progress...
Big Data for Official Statistics	Buelenes, B. et al. (2014) Selectivity of Big Data <a href="http://www.cbs.nl/nl-NL/menu/methoden/onderzoek-methoden/discussionpapers/archief/2014/2014-selectivity-of-big-data-pub.htm">http://www.cbs.nl/nl-NL/menu/methoden/onderzoek-methoden/discussionpapers/archief/2014/2014-selectivity-of-big-data-pub.htm</a>
Visualization	Tennekes, M., Jonge, E. de, Daas, P.J.H. (2013) Visualizing and Inspecting Large Datasets with Tableplots, Journal of Data Science 11 (1), 43-58. <a href="http://www.jds-online.com/file_download/379/JDS-1108.pdf">http://www.jds-online.com/file_download/379/JDS-1108.pdf</a>  Tennekes, M., Jonge, E. de, Daas, P.J.H. (2012) Innovative visual tools for data editing. Paper presented at the United Nations Economic Commission for Europe (UNECE) Work Session on Statistical Data Editing, 2012, Oslo, Norway. <a href="http://www.unece.org/fileadmin/DAM/stats/documents/ece/ces/ge.44/2012/30_Netherlands.pdf">http://www.unece.org/fileadmin/DAM/stats/documents/ece/ces/ge.44/2012/30_Netherlands.pdf</a>
R packages by Statistics Netherlands (all on CRAN)	Visualization: tabplot, tabplotd3, treemap, geo (in development only) Data editing: editrules, deducorrect, rspa Large data processing: ffbase, LaF Other: extremevalues, stringdist, whisker

