## **Big Data**

#### **Case studies in Official Statistics**

#### **Martijn Tennekes**

Special thanks to Piet Daas, Marco Puts, May Offermans, Alex Priem, Edwin de Jonge



#### From a Official Statistics point of view

#### Three types of data:

Survey data = data collected by SN with questionnaires



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



3. Big data = machine generated data of events





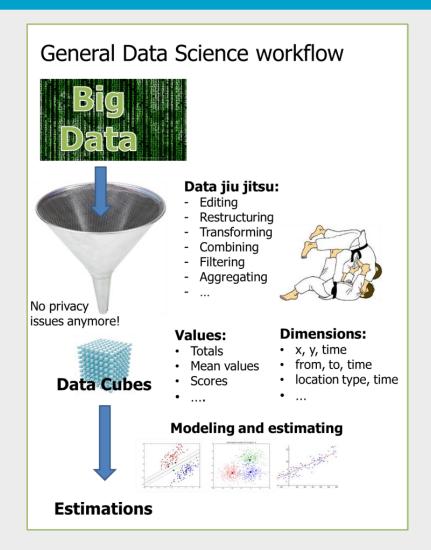
### Big Data case studies

#### Big data = machine generated data of events

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



### Big data approach

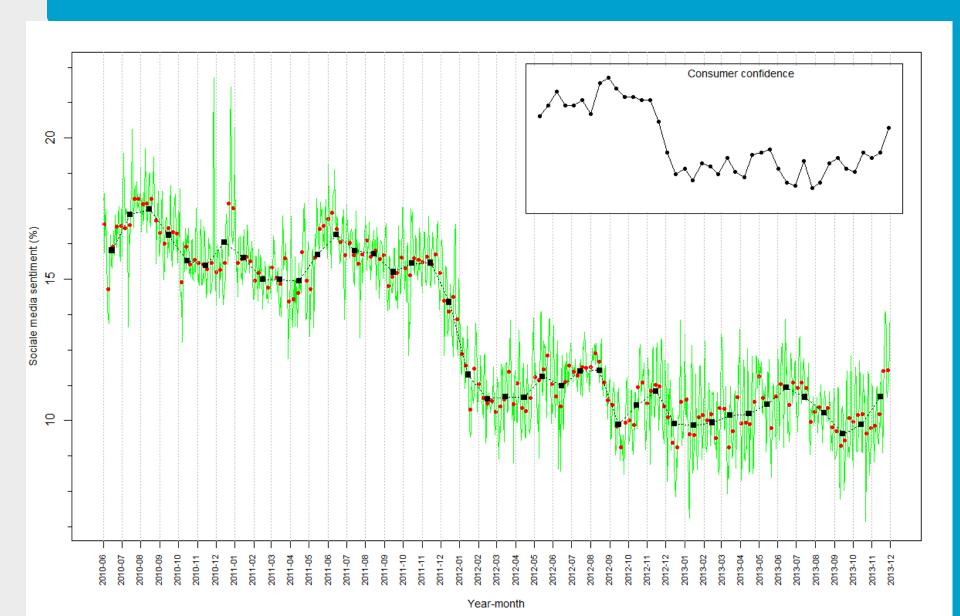




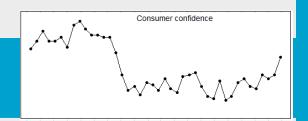
#### Case study 1: Social media

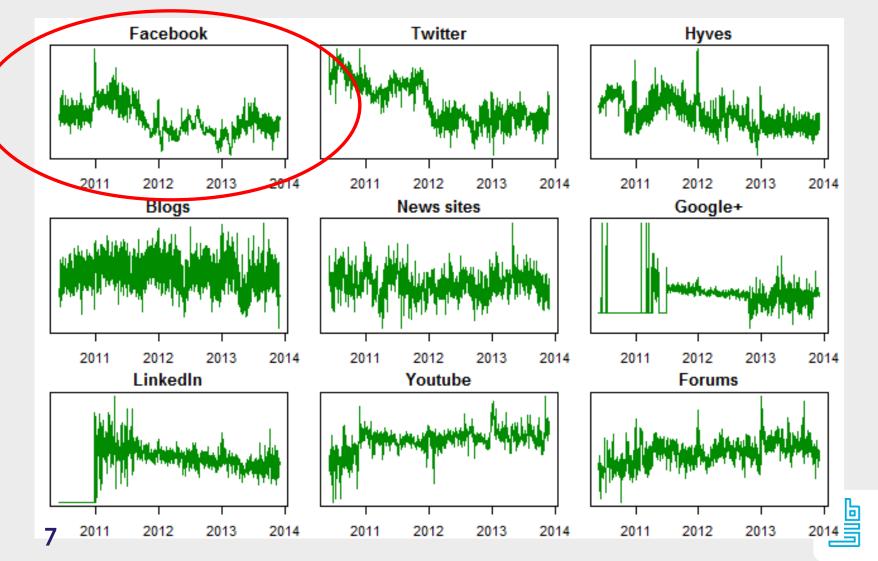
- 3 billion messages as of 2009 gathered from Facebook,
  Twitter, LinkedIn, Google+ by a Dutch intermediate
  company Coosto.
- Sentiment per message determined by classifying words as negative or positive.
- Could be used as indicator for the business cycle. Could it be fit to the consumer confidence,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

#### Sentiment in social media



#### Platform specific sentiment





### 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 $(r)^2$		
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>&</sup>lt;sup>1</sup>period covered June 2010 untill November 2013

Granger causality reveals that Consumer Confidence precedes Facebook sentiment! (p-value < 0.001)



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

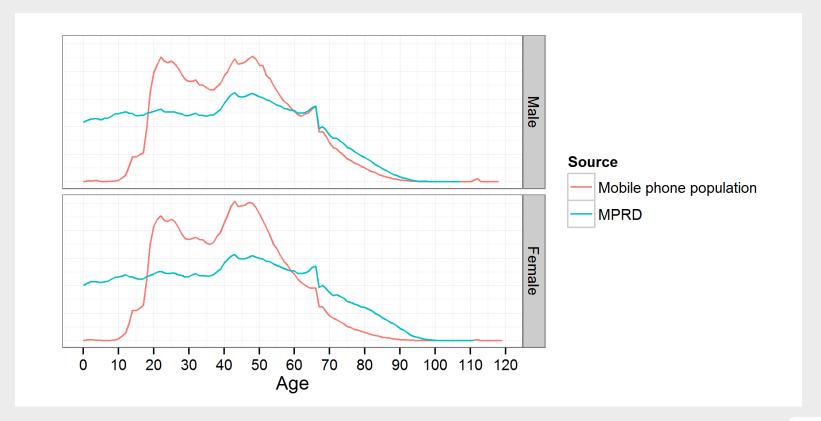
<sup>\*</sup>cointegrated

#### 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.

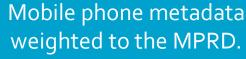


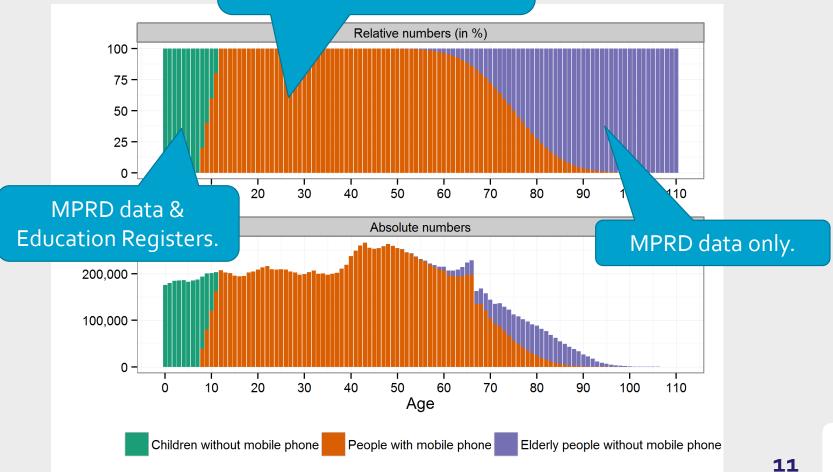
### Mobile phone population





### Subpopulations model



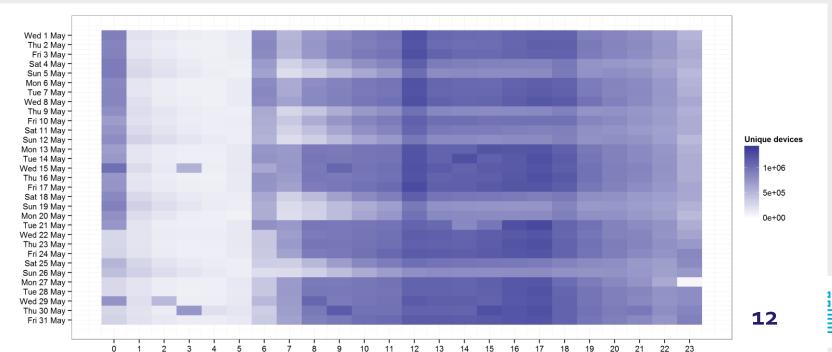




#### Mobile phone metadata

Event Datail 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.



Hour



### Weighting method

Example: suppose there are only 3 regions in the

	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

	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
	MPRD total	800,000	15,000	30,000



## Weighting method (3)

Example: suppose there are only 3 regions in the

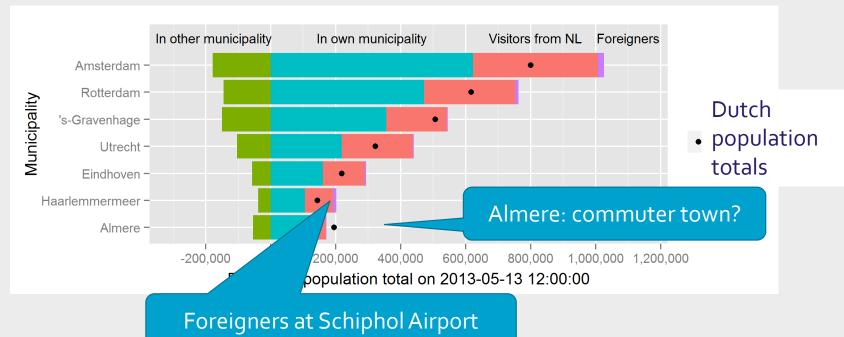
	Residence			
		Amsterdam	Boskoop	Castricum
Current region at time t	Amsterdam	796,000	3,000	6,000
	Boskoop	2000	10,500	0
	Castricum	2000	1,500	24,000
	MPRD total	800,000	15,000	30,000

## Weighting method (4)

Example: suppose there are only 3 regions in the

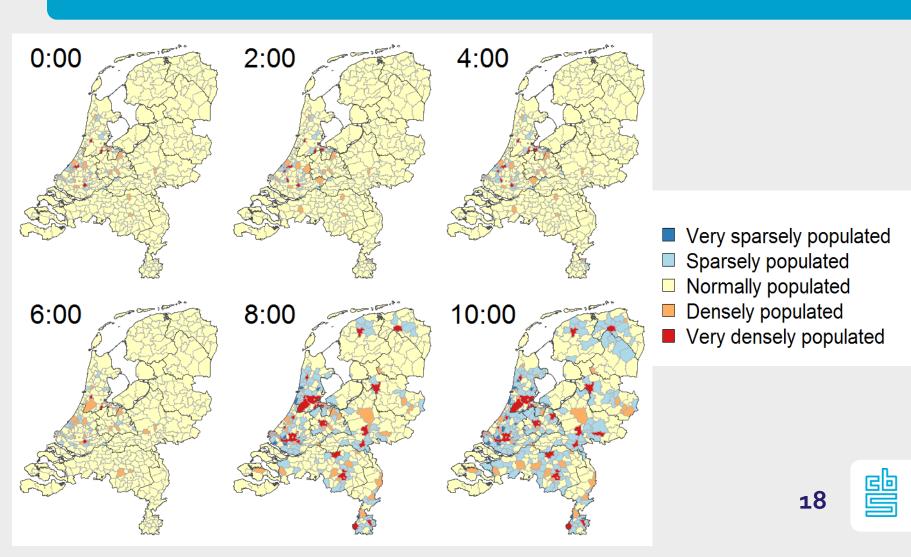
	Residence				
Current region at time t		Amsterdam	Boskoop	Castricum	DTP total
	Amsterdam	796,000	3,000	6,000	805,000
	Boskoop	2000	10,500	0	12,500
	Castricum	2000	1,500	24,000	27,500
	MPRD total	800,000	15,000	30,000	

### Daytime population results

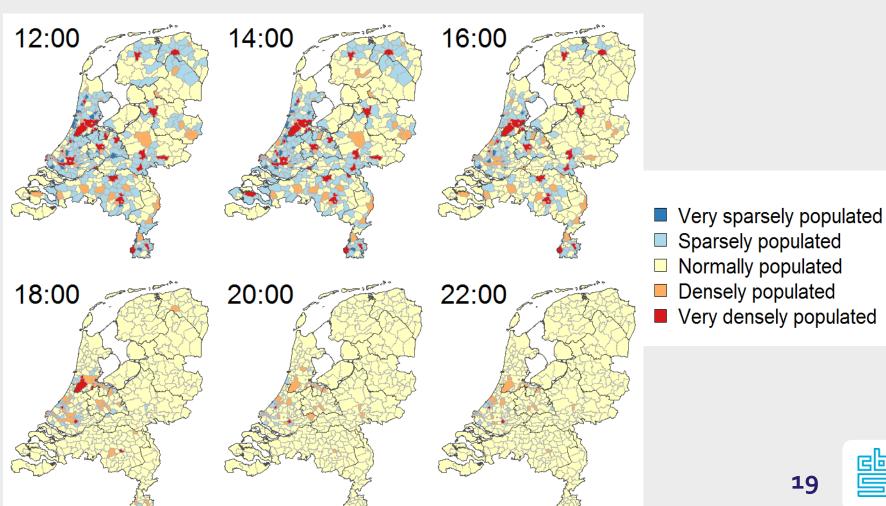




## Day time population (relative)

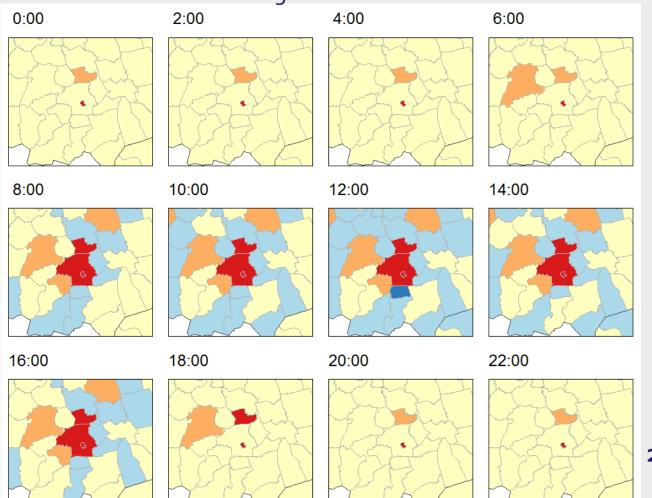


### Day time population (relative)



## Day time population (relative)

#### City of Eindhoven and surrounding towns





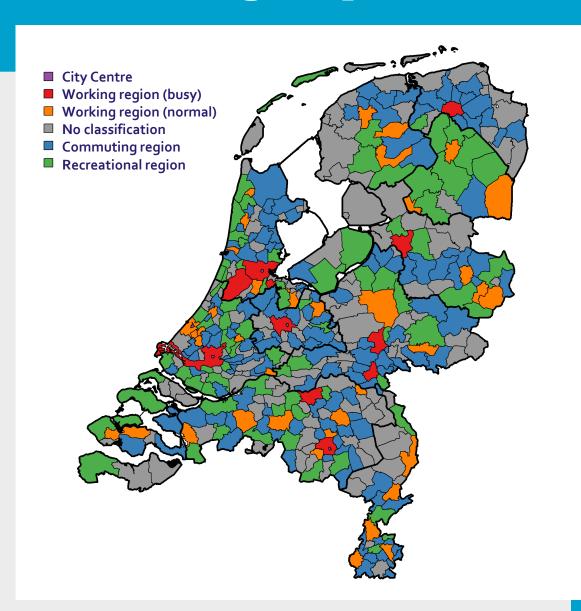
## Day time population - Region profile

#### K-means clustering

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

Weekend = weekends activity

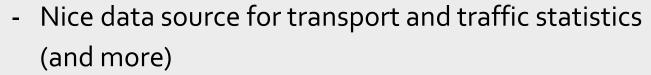
**Holiday** = May holiday activity

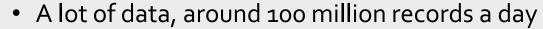


### Case study 3: Traffic loops

#### Traffic loop data

- Each minute (24/7) the number of passing vehicles is counted in around 20.000 'loops' in the Netherlands
  - Total and in different length classes



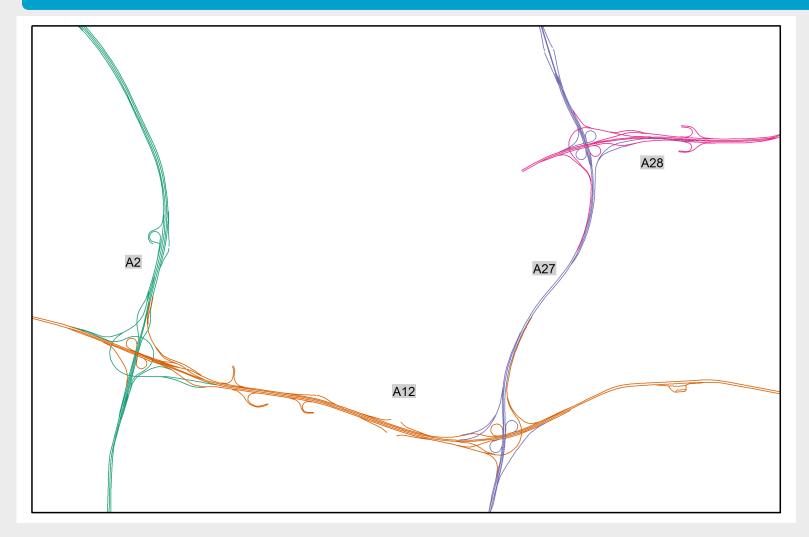






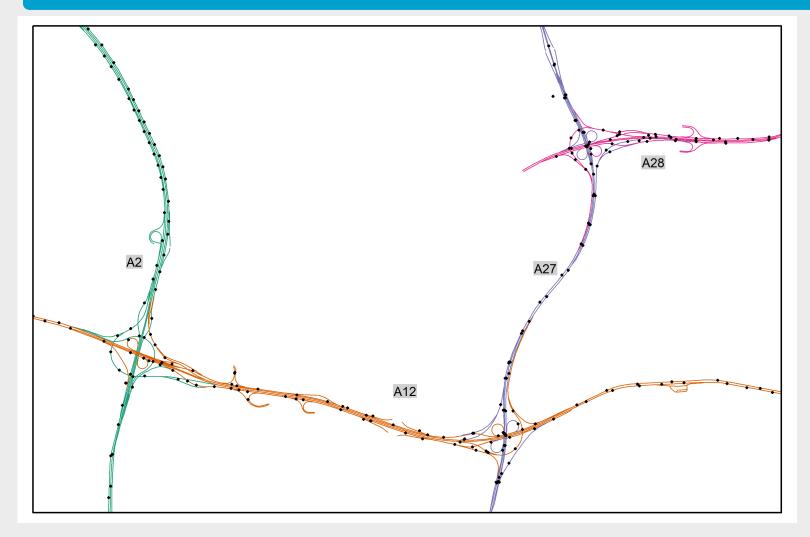


## Traffic loops on main roads



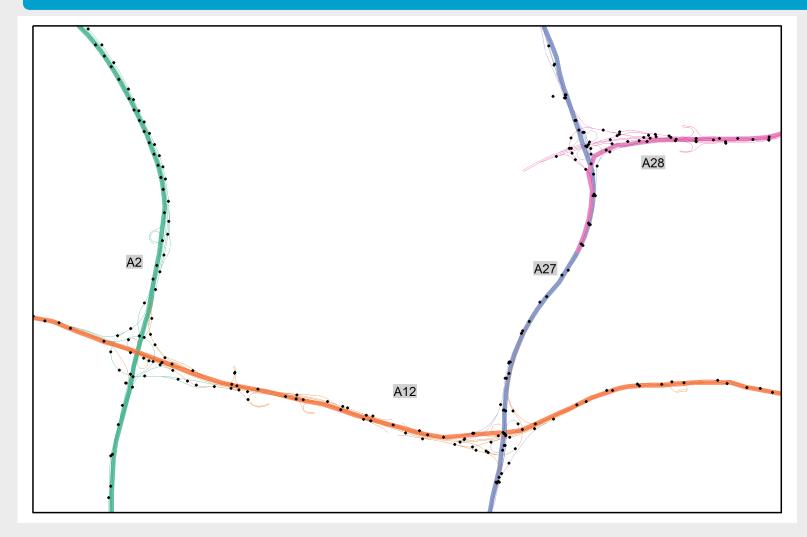


## Traffic loops on main roads (2)





## Traffic loops on main roads (3)



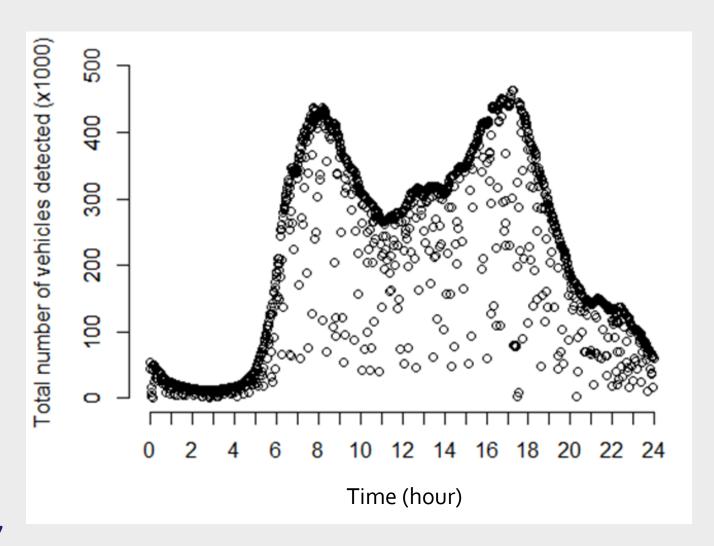


## Traffic loops on main roads (4)





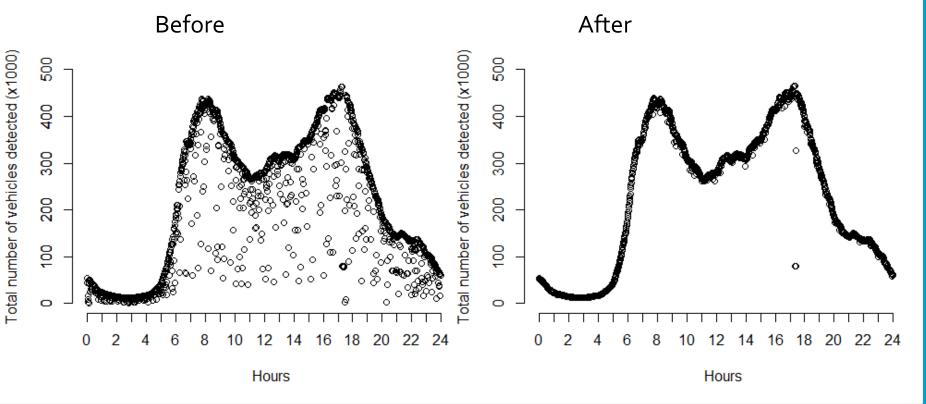
## Raw data: Total number of vehicles a day





#### Correct for missing data: macro level

Sliding window of 5 min. Impute missing data.



Total = ~ 295 million detected vehicles

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



# 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



# **Questions?**

