

# MOBILITY IN THE AGE OF BIG DATA

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DATA ANALYTICS 2018, The Seventh International Conference on Data Analytics,  
November 18 - 22, 2018 - Athens, Greece

# INTRODUCTION

The Seventh International Conference on Data Analytics DATA ANALYTICS 2018 November 18-22, 2018 - Athens, Greece  
TTASC: Transport and Traffic Analytics in Smart Cities  
Keynote: Ivana Semanjski - Mobility in the age of big data



# ADDING **VALUE** TO SOCIETY

Innovation Centre For Intelligent Information Processing

WHAT WE DO



HoGent





# BY ASSIGNING **VALUE** TO DATA

Innovation Centre For Intelligent Information Processing

WHAT WE DO



HoGent



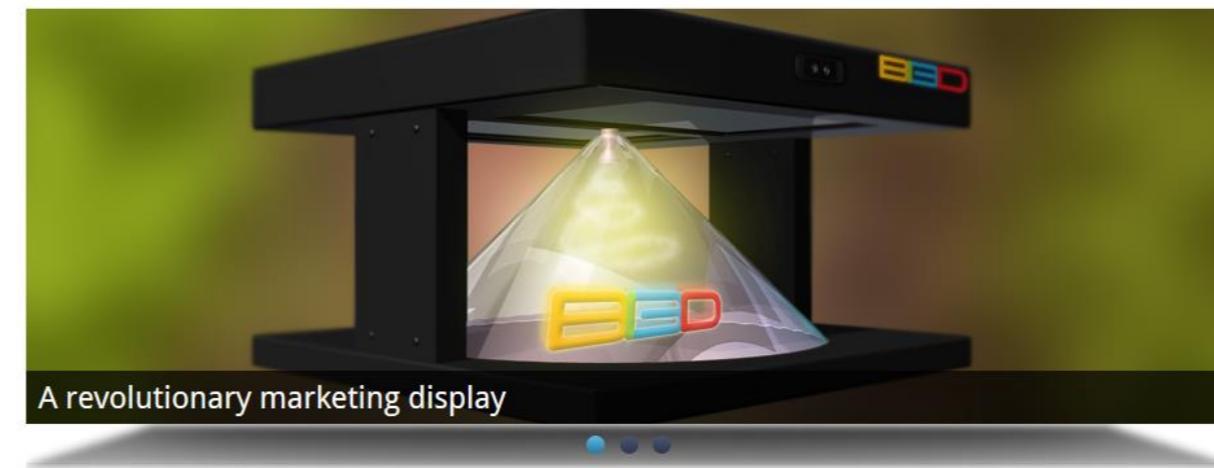
# Spin off companies



# Spin off companies



Home Contact



# Industrial cooperation and partnerships



# DATA COLLECTION

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# Transportation planning



prediction of usage demand in future travel

ensuring necessary facilities and services to cater to that demand

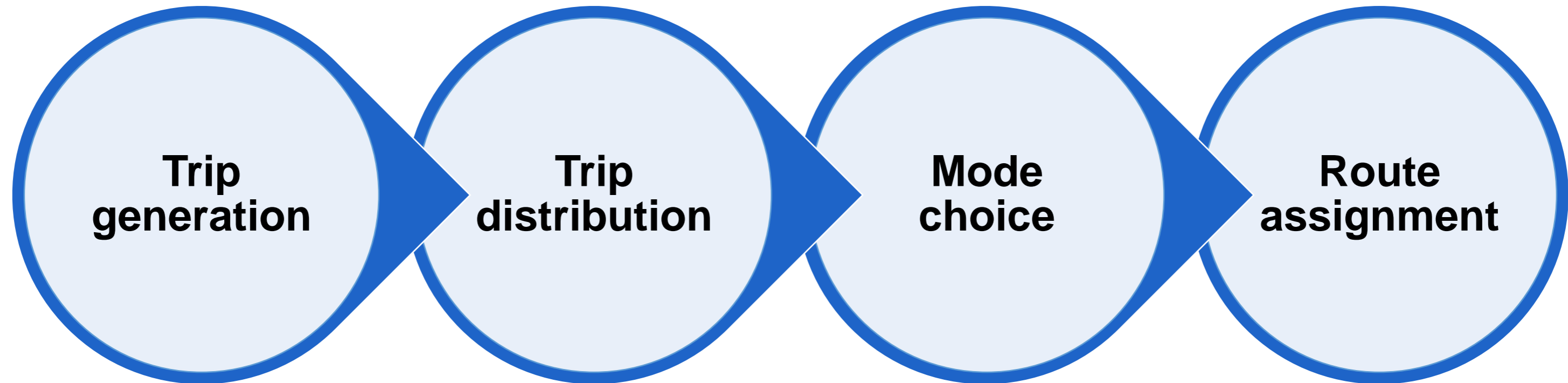
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# HOW DO WE DO IT?

## Four-step models



# How do we collect data for transportation planning

Travel diaries

Interviews



78% of European citizens live in cities

2014 54% of the total global population lived in urban areas

average each person produces 1 GB of content daily

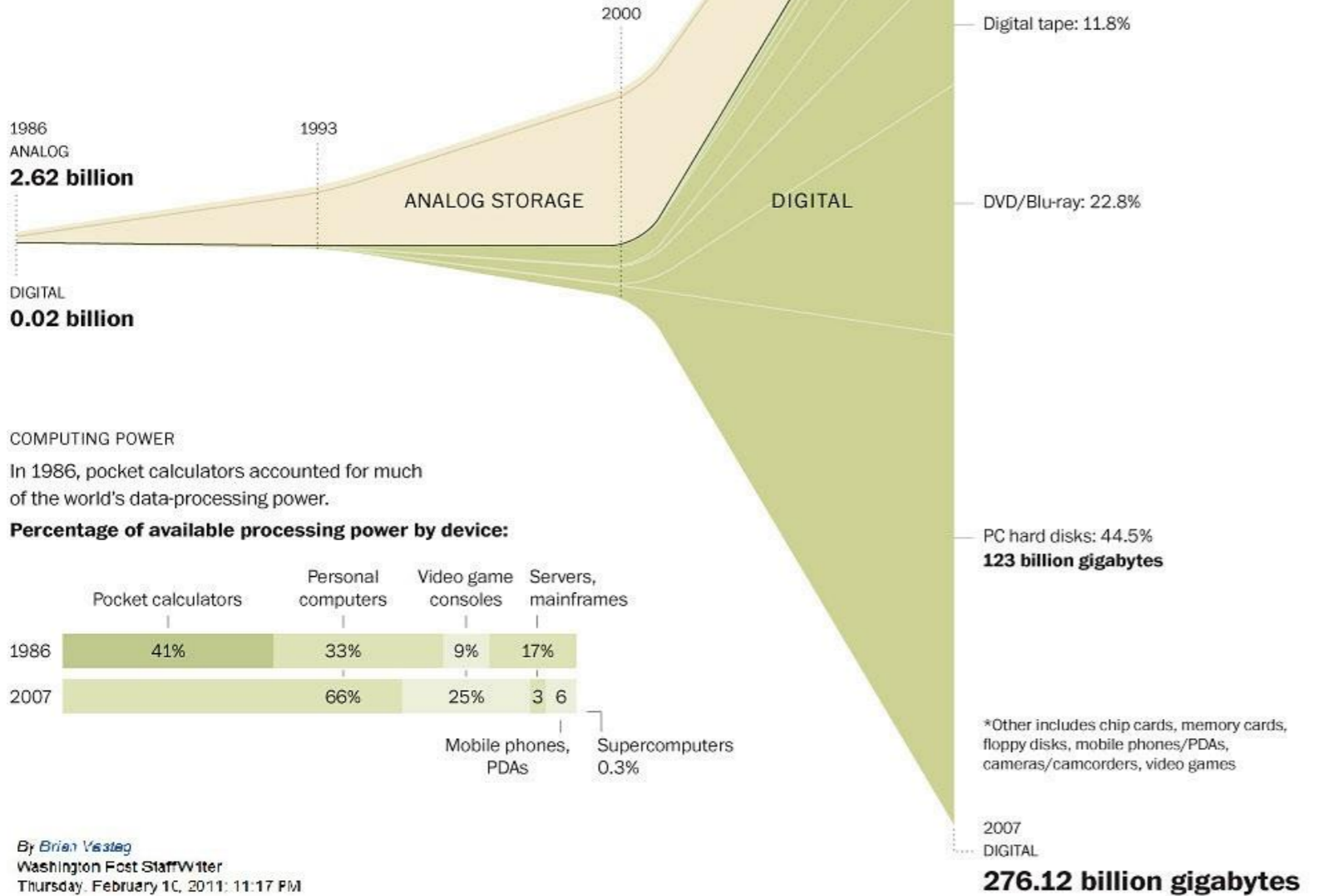
# The Washington Post

## Exabytes: Documenting the 'digital age' and huge growth in computing capacity

### THE WORLD'S CAPACITY TO STORE INFORMATION

This chart shows the world's growth in storage capacity for both analog data (books, newspapers, videotapes, etc.) and digital (CDs, DVDs, computer hard drives, smartphone drives, etc.)

In gigabytes or estimated equivalent



By Brian Vesteg  
 Washington Post Staff Writer  
 Thursday, February 10, 2011: 11:17 PM



volume

variety

velocity

value



# Big data sources for mobility studiers

CDR & signalisation

GNSS

Mobile phone

# CDR & NETWORK SIGNALIZATION DATA



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# Global Navigation Satellite Systems

A satellite is visible in the upper right corner of the slide, with its solar panels and antenna extending outwards. The background is a view of Earth from space, showing the curvature of the planet and the blue atmosphere. The landmasses are visible in shades of brown and green, with some white clouds scattered across the surface.

GPS

Galileo

Glonass...

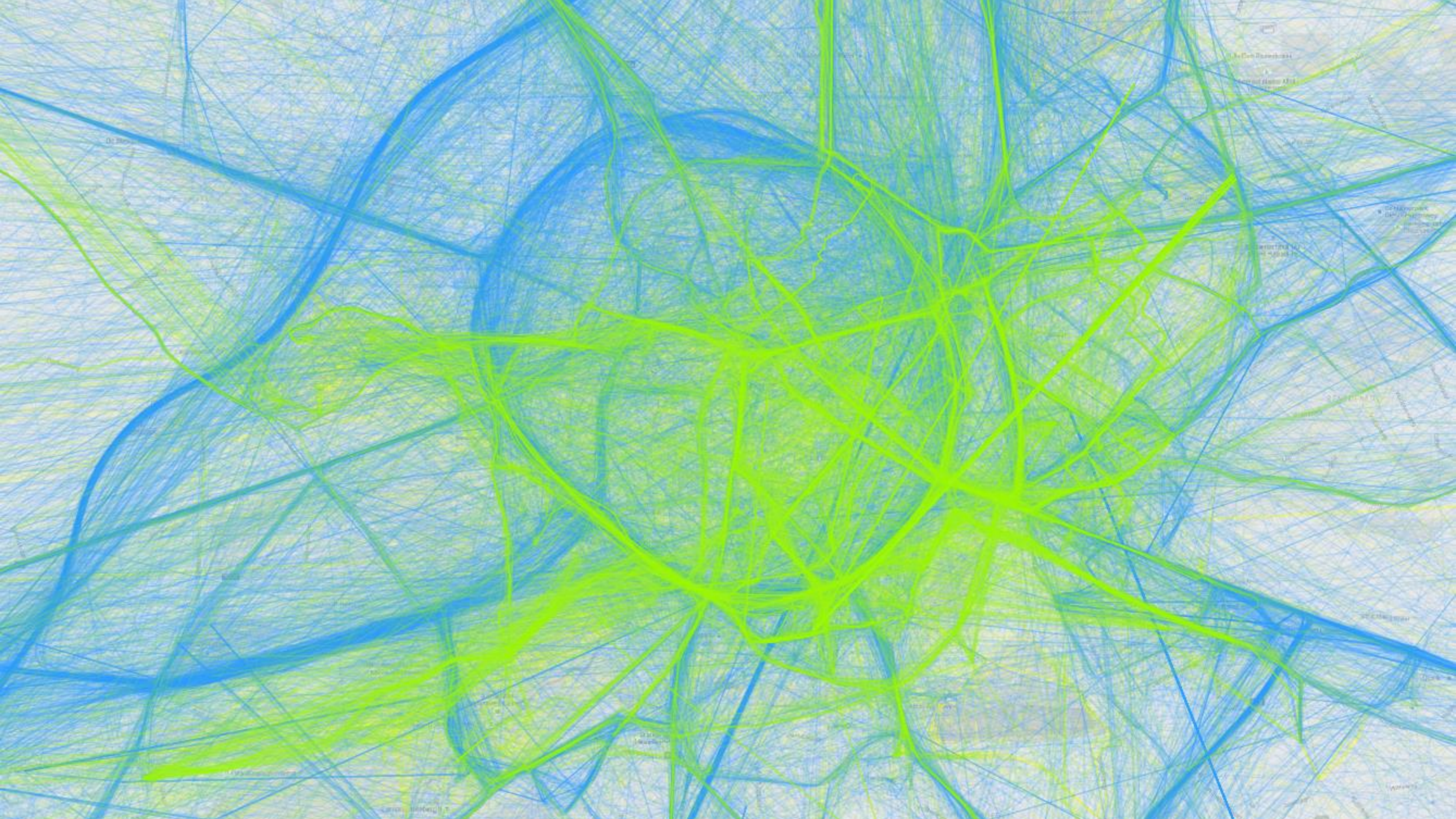


# Mobile phone data



Passive

Active



# TRIP GENERATION

Literature	OD estimation	OD Accuracy	Vehicle occupancy rates	Occupancy rate Accuracy	Duration of test data	Number of users	Dataset	User Validated
(Bohte and Kees, 2009)	No	-	No	-	1 week	1104	GNSS, GIS	Yes
(Munizaga and Palma, 2012)	Yes	<b>82 %</b>	No	-	2 weeks	N/A (74 million observations)	Smartcard, GNSS (PT only)	No
(Shen and Stopher, 2013)	No	-	-	-	3 days	2059	GNSS	No
(Xiao et al., 2016)	No	-	-	-	1 week	321	Smartphone	Yes
(Ge and Fukuda, 2016)	Yes	77%	No	-	1 day	N/A (650,000 observations)	Smartphone	No
(Dong et al., 2015).	Yes	-	No	-	1 day	N/A	CDR	No
(Wolf et al., 2001)	Yes	37%	No	-	3 day	13	GNSS, GIS	Yes
(Lu et al., 2012)	No	-	No	-	<b>13 weeks</b>	N/A (3188 trips)	GNSS, GIS	Yes
(Feng and Timmermans, 2011)	No	-	No	-		329		

# TRIP DISTRIBUTION

Literature	Duration of test data	Number of users	Dataset	Accuracy
(Fanhas and Saptawati, 2017)	2 months	N/A (16,337 records)	GNSS, taximeter	N/A
(Ge and Fukuda, 2016)	1 day	N/A	GNSS	<b>88%</b>
(Moreira-Matias et al., 2016)	<b>9 months</b>	441	GNSS, taximeter	79%
(Li et al., 2017)	1 month	12,000	GNSS	80%
(Ma et al., 2013)	1 month	128000	CDR and network signalization data	78%
(Bahoken and Raimond, 2013)	6 weeks	<b>10 millions</b>	CDR	45%
(Larijani et al., 2015)	1 day	1.4 millions	CDR and network signalization data	N/A
(Bonnel et al., 2015)	10 days	4,1 millions	network signalization data	82%
(Alexander et al., 2015)	2 months	2 million	CDR and network signalization data	65%
(Gundlegård et al., 2016)	2 weeks	300000	CDR	N/A

# MODE CHOICE

Literature	Number of modes	Data	Duration of test data	Number of users	Accuracy	User validated
(Reddy et al., 2008)	3	Mobile sensed GNSS, accelerometer	240 min	6	90	No
(Bohte and Kees, 2009)	4	GNSS, GIS	1 week	1104	70	Yes
(Wang et al., 2010)	5	CDR	12 hours	<b>56,715</b>	70	No
(Reddy et al., 2010)	3	Mobile sensed GNSS, accelerometer	24 hours	16	93	No
(Manzoni et al., 2010)	<b>7</b>	Mobile sensed GNSS, accelerometer	1 day	4	82	Yes
(Hemminki et al., 2013)	4	Mobile sensed accelerometer	150 hours	16	60-85	Yes
(Biljecki et al., 2013)	7	GNSS, GIS	1 week	1104	92	No
(Xiao et al., 2015)	5	GNSS,	5 days	202	86	Yes
(Zhou et al., 2016)	3	Mobile sensed GNSS, accelerometer	6 days	12	<b>94</b>	Yes
(Semanjski et al., 2017)		Mobile sensed GNSS, GIS	<b>4 months</b>	8000	<b>94</b>	Yes

# ROUTE ASSIGNMENT

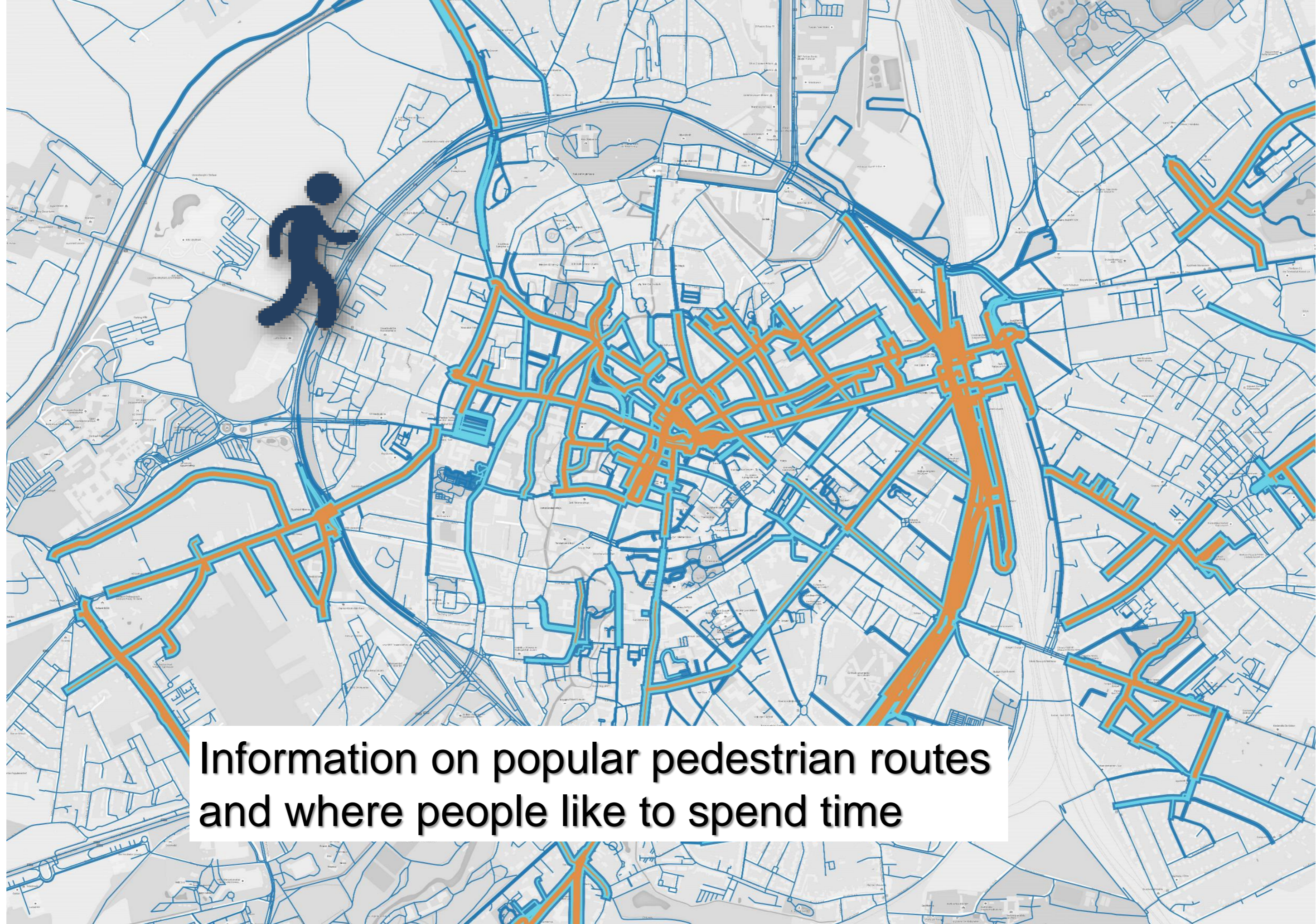
Literature	Data type	Data Size	Time span	Transport mode	Success rate
(Marchal et al. 2005)	GNSS, low and high-resolution road data	3 cars (2.5 million points)	3 weeks	Car	Up to 10 m/pt
(Yang et al. 2005)	GNSS, road network	1 car (9500 points)	3 h	Car	<b>100%</b>
(Krumm et al. 2007)	GNSS, road network (NAVTEQ)	187 cars (1,351,669 points)	2 weeks	Car	N/A
(Chen et al. 2014)	GNSS, road network	<b>10,245 taxis (172,154 points)</b>	15 minutes	Car	90-97 %
(Lou et al. 2009)	synthetic GNSS, GeoLife trajectories, road network	unknown for synthetic data, 28 trajectories for real data	N/A	N/A	Real data: up to 85%, Synthetic data: > 95%
(Yuan et al. 2010)	GeoLife trajectories, road network	26 trajectories	30 hours	N/A	66- 84%
(Li et al. 2013)	GNSS, OpenStreetMap road network data	121737 trajectories	<b>1 month</b>	Car	85%

# WHERE BIG DATA CAN ASSIST TRANSPORT PLANNING

- Data collection
- Complementing traditional models by:
  - Identifying TAZs properties as population, trip generation/attraction
  - Detecting trip purpose
  - Matching OD pairs
  - Detecting transport mode
  - Reconstructing trip paths.....







**Information on popular pedestrian routes  
and where people like to spend time**

# CONSUMER PROFILING

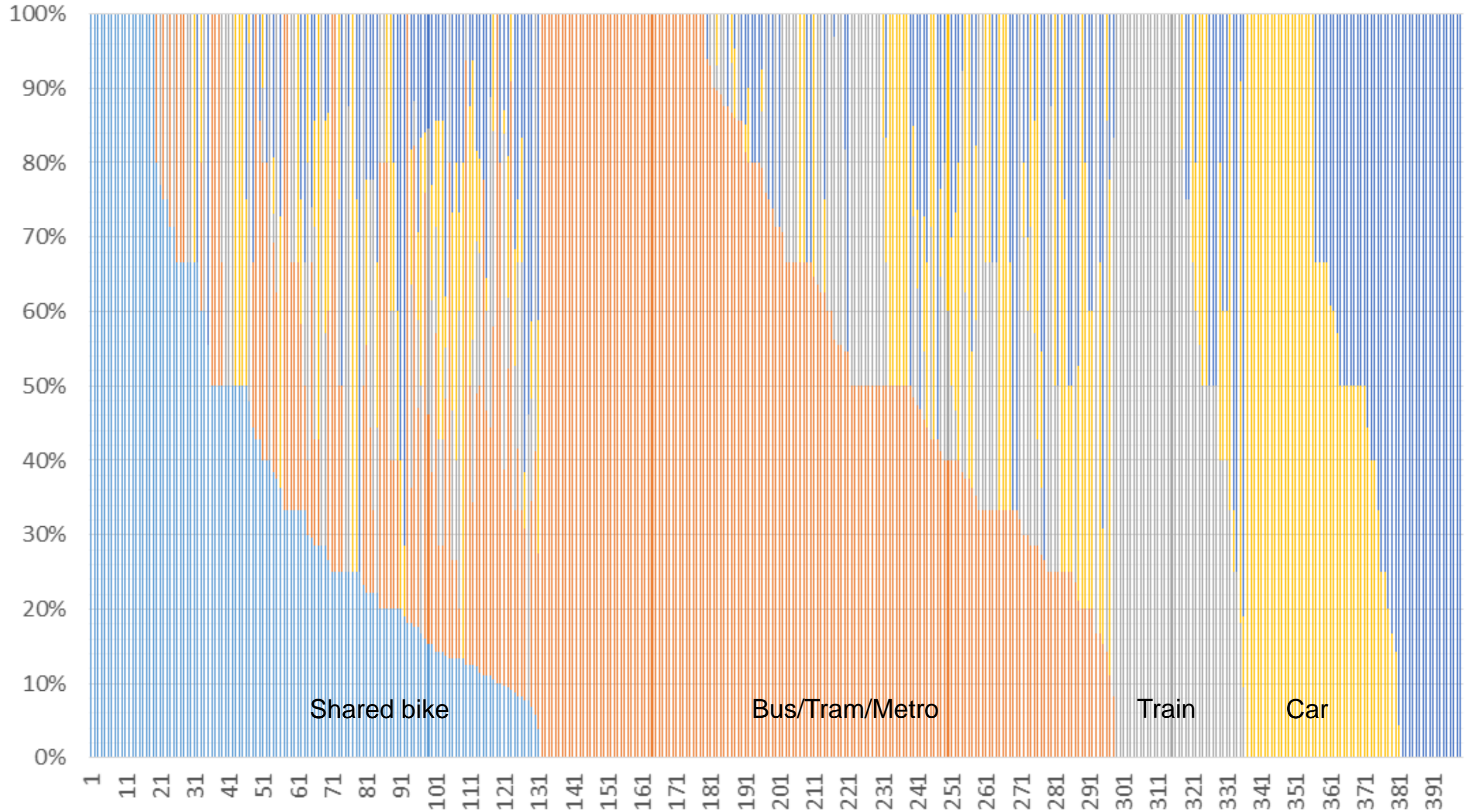


## TARGET GROUP INSIGHT

- **Mobility behavior**  
*mode, shift, travel time, ...*
- **Customer metrics**  
*recency, frequency, duration, ...*
- **Spatial insight**  
*O/D, visitor flow, POI, ...*

# Mobility Profiles

■ VELO ■ BTM ■ TREIN ■ PARK\_GARAGE ■ PARK\_STRAAT



Thank you for your attention!!!