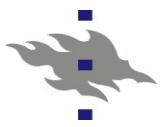




A close-up photograph of a woman's face. She is looking down at a smartphone held in her hands. Her expression is focused and slightly smiling. The background is blurred, showing some colorful lights and what might be a city street at night.

Big Data Applications in Programmatic Marketing Ecosystem

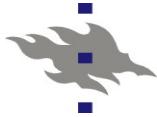
Ilkka O. Lavas
Helsingin Yliopisto
Tietojenkäsittelytieteen laitos



Big Data Applications

Deep Dive to Programmatic Marketing Ecosystem

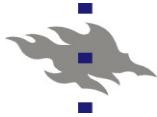
- Why this is interesting
- Summary
- How this relates to Big Data
- Summary and Conclusion



Big Data Applications

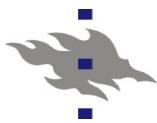
Big data have many applications in different areas

- science and research
- public health
- customer relation management
- machine and device performance analysis
- optimizing cities and countries
- finance and banking
- Advertising and data driven marketing



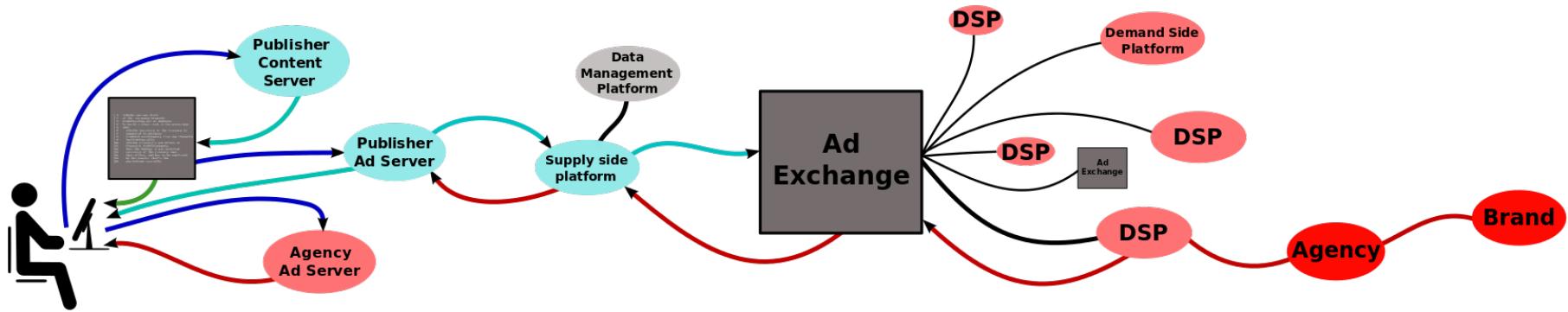
Programmatic Marketing Ecosystem examples

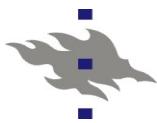
- Programmatic = Automating simple tasks
 - Marketing = telling stories to existing and new potential customers
 - Ecosystem = distributing of data between systems
-
- Hotels.com example (retargeting, email reminder)
 - Zalando example



Advertising

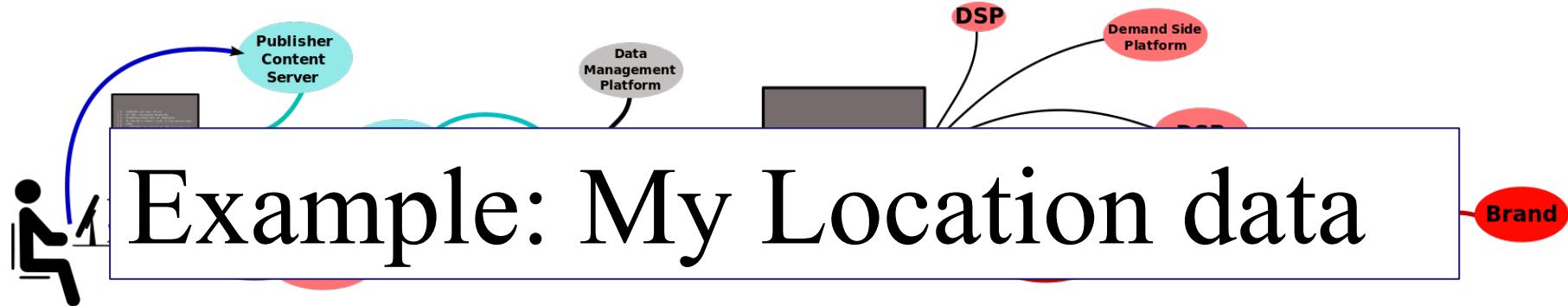
Programmatic marketing ecosystem

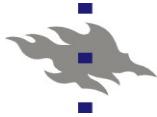




Advertising

Programmatic marketing ecosystem

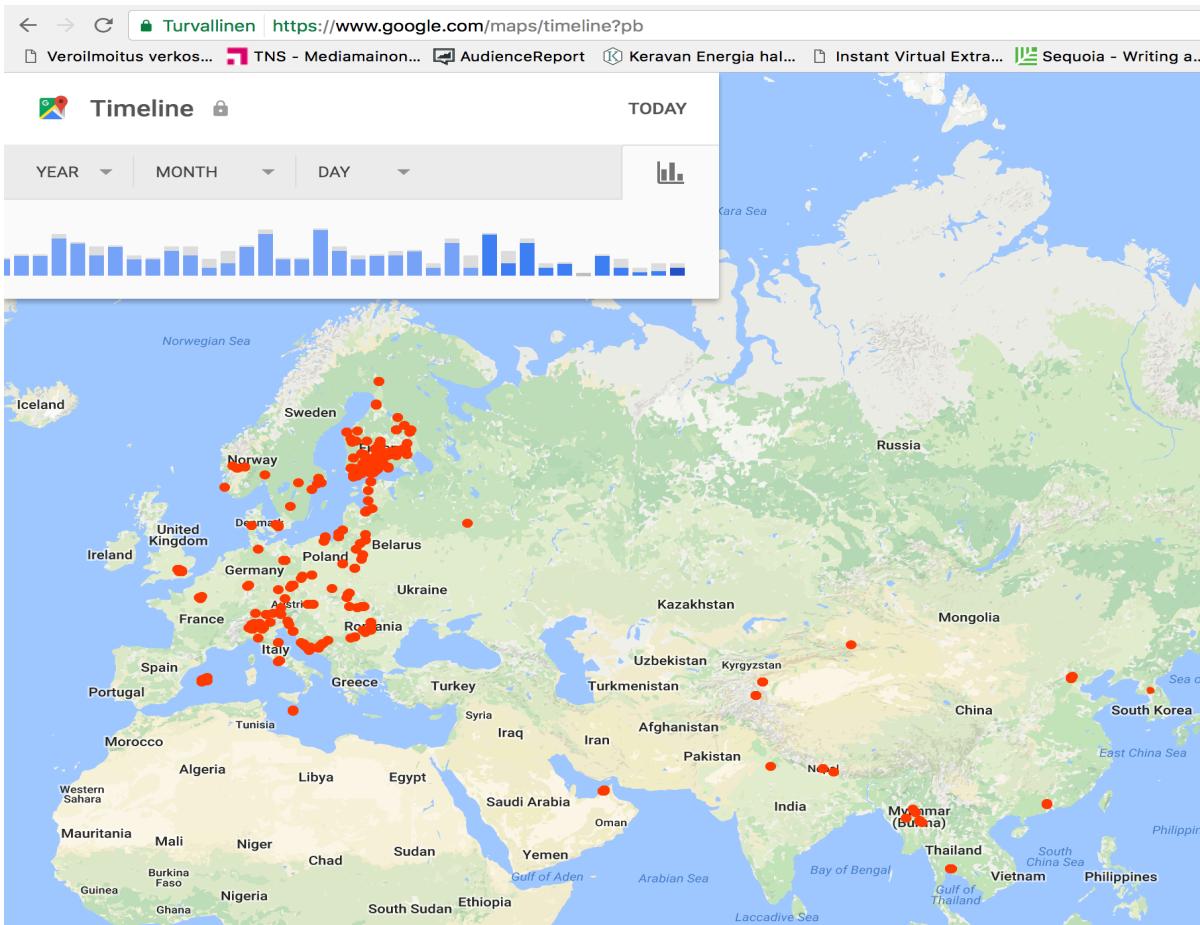
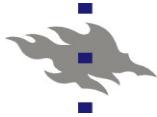




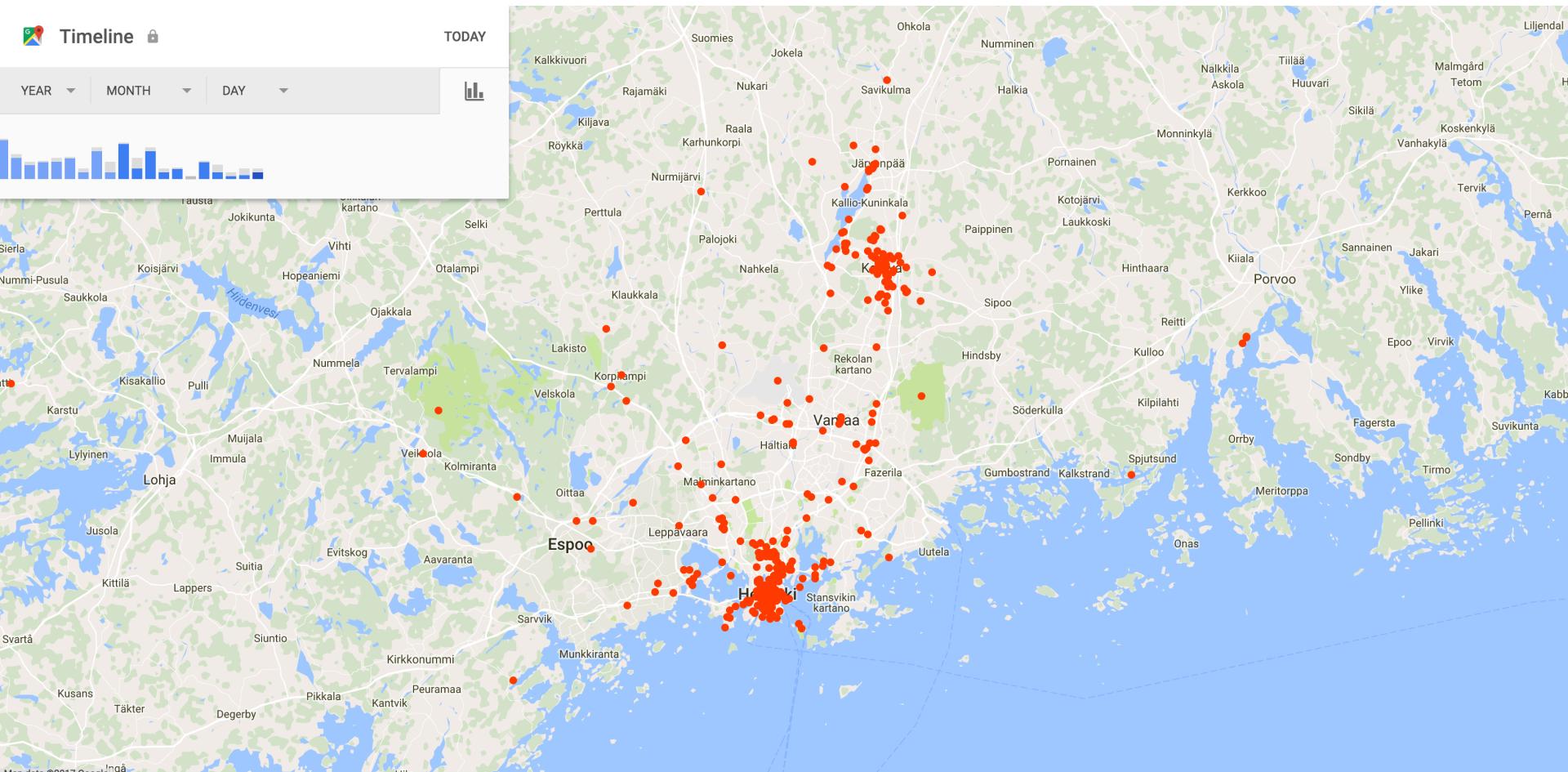
Example:

**Geospatial data
combined with location data**

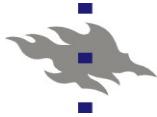
**is one form of data in programmatic marketing
ecosystem**



Ilkka Lavas, Helsingin Yliopisto, Tietojenkäsittelytieteen laitos

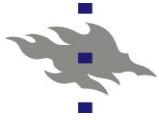


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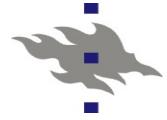
Why this is interesting?

- geospatial big data and location data can give benefit to
 - save fuel
 - Save time
 - Increase revenues
 - Plan urban areas
 - Improve health care
 - Better marketing



Applications

- Forum example (from mobile screenshot)



Ilmoitukset

Pikavalikot

Aika

12.36 13. HELMIKUUTA
maanantai

12....

Arkistoitu.

KUMOA

12....



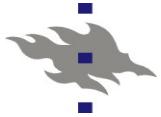
Forum

Ostoskeskuksen hakemisto



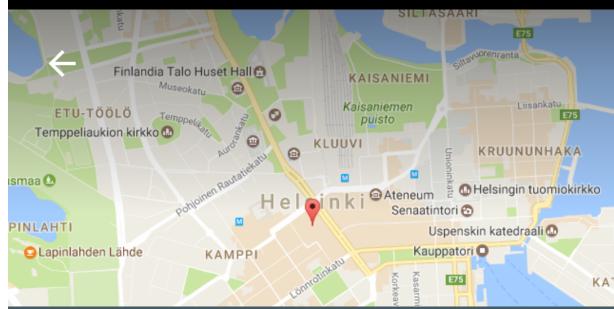
DNA





DNA

NFC 4G 29 % 12.42



Forum

Ostoskeskuksen hakemisto

Specsavers Optikko
Helsinki Forum

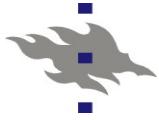
Dressmann Forum
Miestenvaateliike

KIC
Optikko

Ostoskeskuksen hakemisto

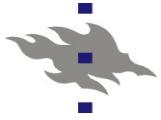
Hakemisto – Forum





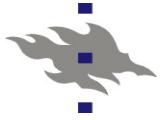
Case Netflix

- Netflix generated the data on the top-50rentals in 2009 in each zip code
 - Such patterns are very useful for recommending movies to their users



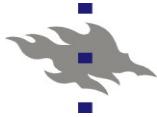
Marketers & brands want to talk to you personally and big data can help them to make you feel

“Oh how nice from you
to think of me!”



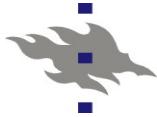
Applying Big Data

1) Collecting data



Amount of data is increasing

- McKinsey Global Institute says that the pool of personal location data was in the level of 1 PB in 2009 and is growing at a rate of 20% per year
- + + + New data from internet of things data stored to internal archives
- in Google, about 25 PB of data is being generated per day, and a significant portion of the data falls into the realm of spatio-temporal data



Rise of mobile

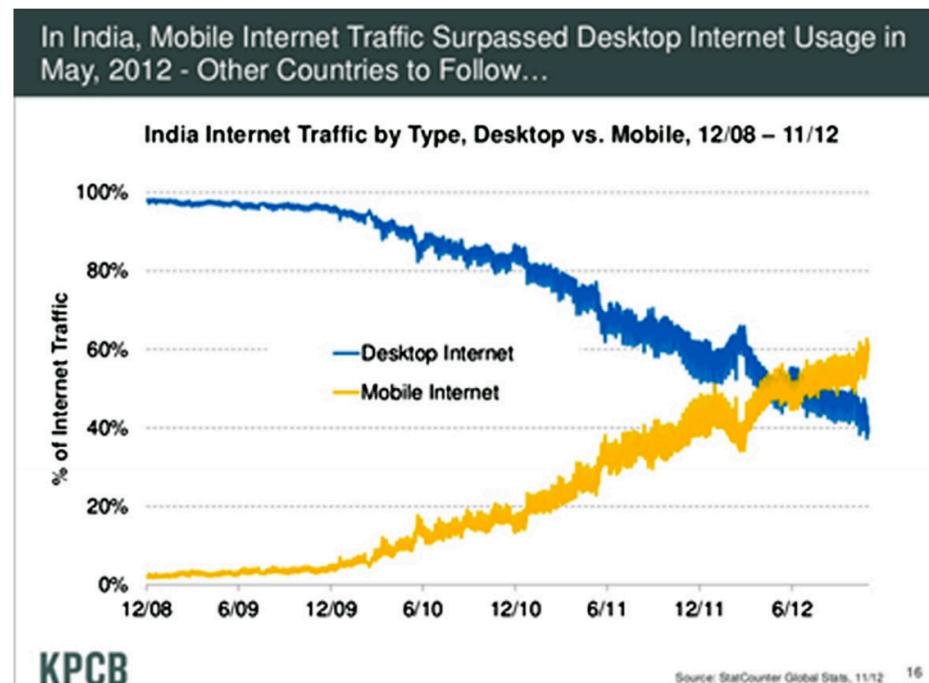
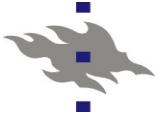
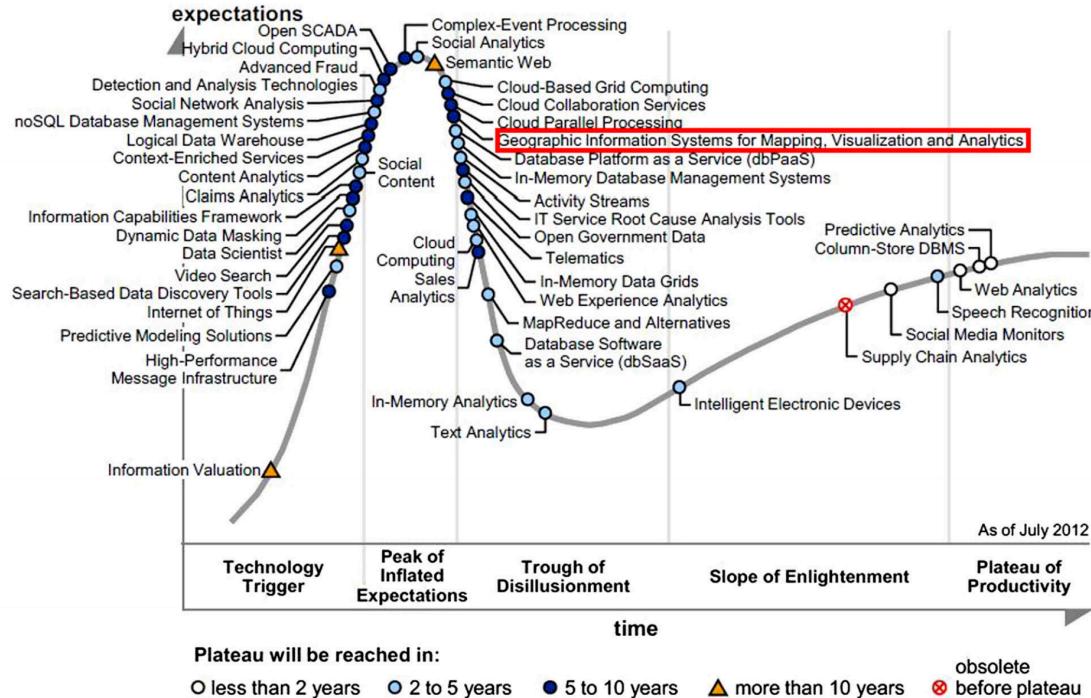


Fig. 1. Mobile internet traffic in India [3].

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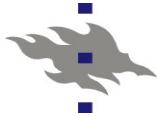


Not only hype anymore



Source: Gartner (July 2012)

Fig. 2. Gartner's hype cycle (source: Gartner) [10].

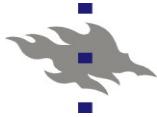


Big Data is important for marketing

Example Location data, subset of Big Dta:

**“Location targeting is holy grail
for marketers.”**

- Sir Martin Sorrell, the CEO of WPP Group



Applications

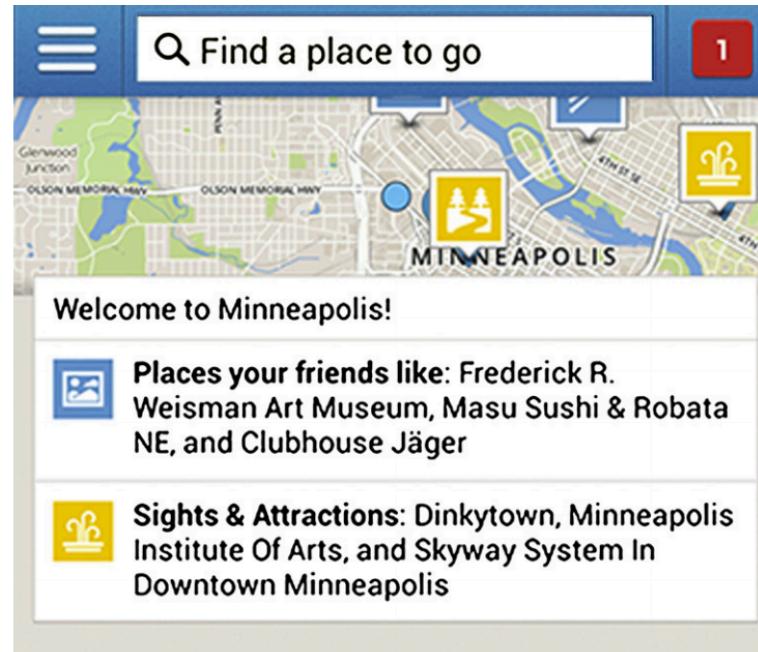
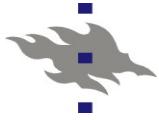


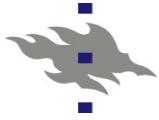
Fig. 6. Foursquare Android app with recommendations.

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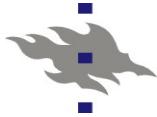
How to Apply big data for marketing

2) Analysing



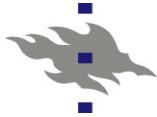
Prediction

- The future location of a human can be predicted by analyzing his/her records of previous traces
- Analysis of human mobility can boost many applications ranging from epidemic modeling to traffic prediction and urban planning.



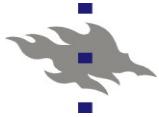
Case Walmart and Hurricane

- The executives at Walmart decided to adopt one of big data technologies—predictive analytics
- Linda M. Dillman, Walmart's chief information officer, asked her staff to predict what would happen soon based on what had happened when Hurricane Charley landed several weeks ago.
- By analyzing the transaction records stored in Walmart's data warehouse, the company could predict which items were bought just before or after an event (i.e., a hurricane) at a specific region.



Case Walmart and Hurricane

- People who had lived in Florida's Atlantic coast did not increasingly buy some products **directly related to hurricanes, e.g., water and flash lights.**
- Surprisingly, strawberry PopTarts increased in sales, by seven times compared with their usual sale rate, just before a hurricane. In addition, the top-selling item immediately before the hurricane was instead ...



Top selling product before Hurricane was beer



How Target Figured Out A Teen Girl Was Pregnant Before Her Father Did



Kashmir Hill, FORBES STAFF

Welcome to *The Not-So Private Parts* where technology & privacy collide [FULL BIO ▾](#)

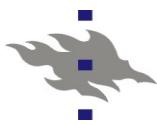
Every time you go shopping, you share intimate details about your consumption patterns with retailers. And many of those retailers are studying those details to figure out what you like, what you need, and which coupons are most likely to make you happy. Target , for example, has figured out how to data-mine its way into your womb, to figure out whether you have a baby on the way long before you need to start buying diapers.



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Microsoft Cloud

lower the cost of borrowing by 90%.

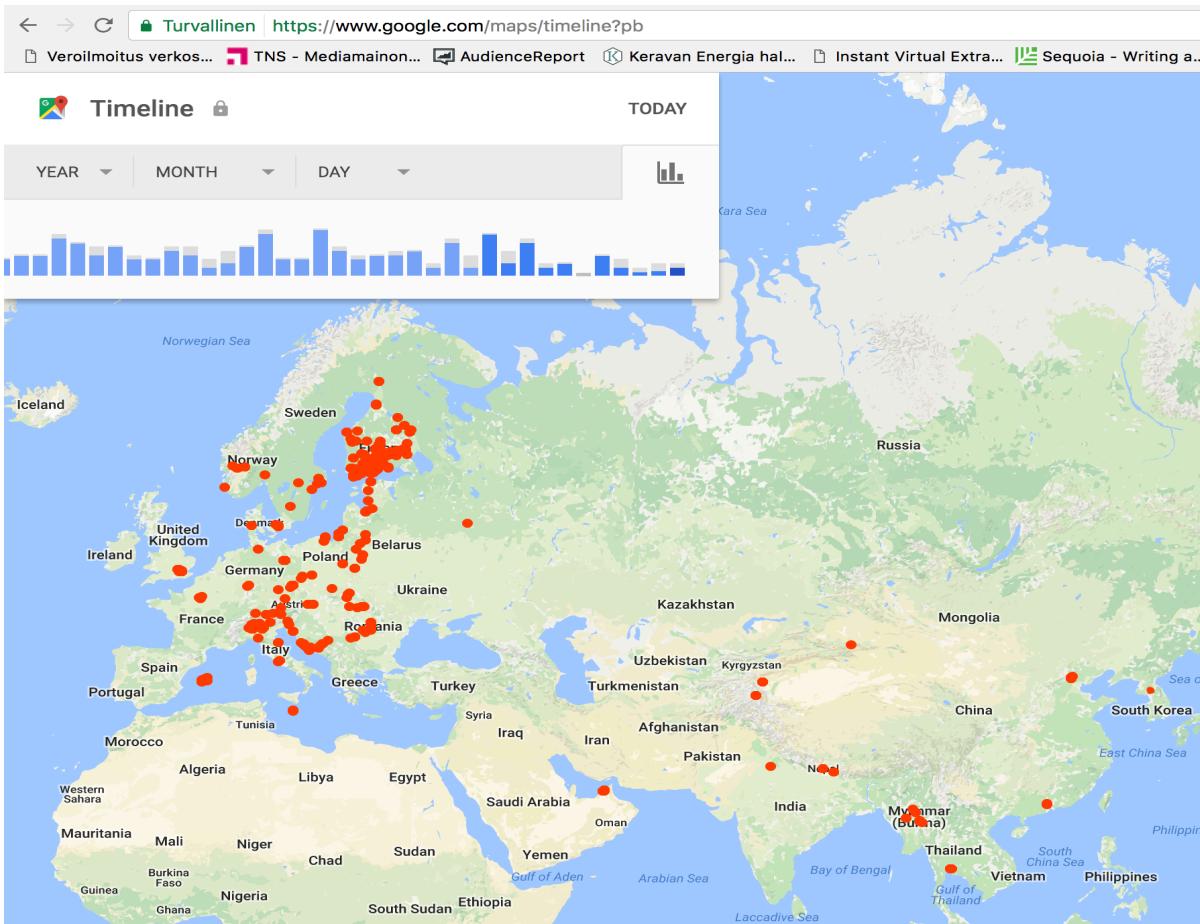
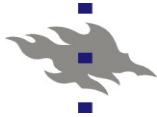


How we predict relevancy: example nearbyness

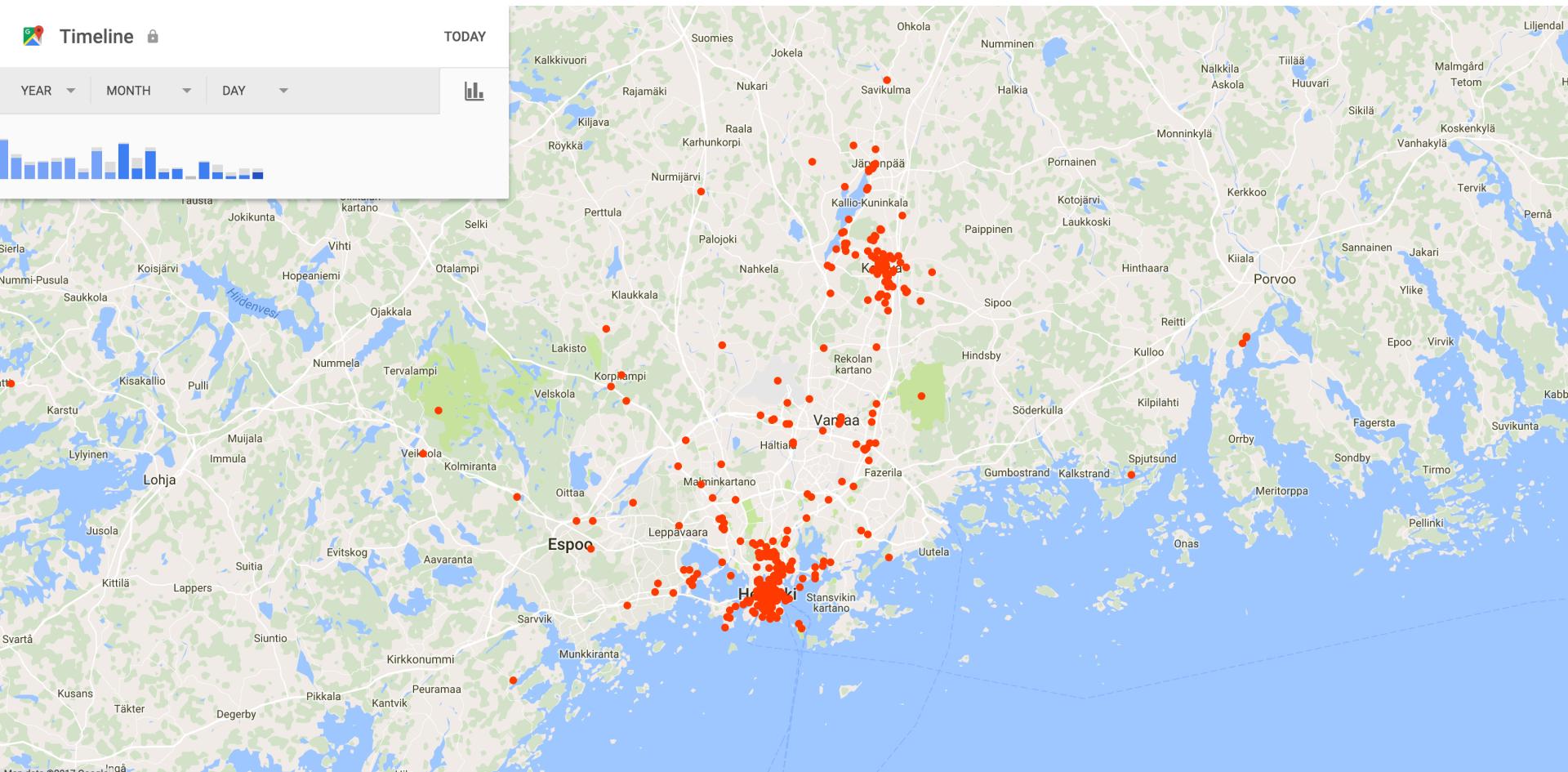
- Direct marketing.
 - This field is supported by Tobler's first law of geography

“Everything is related to everything else,
but near things are more related than distant things.”

- suggesting the services or stores close to the current location of a user should be more effective than suggesting those far-away



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Ikkka Lavas, Helsingin Yliopisto, Tietojenkäsittelytieteen laitos

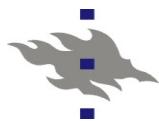


How data is collected?

Ilkka-MacBook-Pro-Isos:~\$ netstat

Active Internet connections											
Proto	Recv-Q	Send-Q	Local Address	Foreign Address	(state)	Proto	Recv-Q	Internet connections	Foreign Address	(state)	
tcp4	0	0	192.168.43.39.55465	arpn09\$11-in-xe0..https	ESTABLISHED	Active LOCAL (UNIX) domain sockets	0	0	kb9b945eccd872d stream	0	0
tcp4	0	0	192.168.43.39.55465	189.169.56.99.100	CLOSE_WAIT	Address	Type	Recv-Q	Conn	Refs	
tcp4	0	0	dt#43h716z7qgbv.555442	2001:14:b8:1:8000:4.https	ESTABLISHED	b0b945eccd872d stream	0	0	b0b945eccdca8c5	0	0
tcp4	0	0	192.168.43.39.55439	40.71.176.https	ESTABLISHED	b0b945eccdca8c5 stream	0	0	b0b945eccdca8c5	0	0
tcp4	0	0	192.168.43.39.55399	a23-43-149-72..https	ESTABLISHED	b0b945eccdcb1cd stream	0	0	b0b945eccdcb1cd	0	0
tcp4	0	0	192.168.43.39.55388	62-165-164-293.c..http	ESTABLISHED	b0b945eccdcb1cd stream	0	0	b0b945eccd98b5	0	0
tcp4	0	0	192.168.43.39.55387	62-165-164-293..http	ESTABLISHED	b0b945eccd98b5 stream	0	0	b0b945eccd98b5	0	0
tcp4	0	0	192.168.43.39.55384	62-165-164-293..http	ESTABLISHED	b0b945eccd98b5 stream	0	0	b0b945eccd98b5	0	0
tcp4	0	0	192.168.43.39.55383	62-165-164-293..http	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	192.168.43.39.55382	62-165-164-293.c..http	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	192.168.43.39.55381	62-165-164-293.c..http	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	192.168.43.39.55380	62-183-179-19..co..http	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	dt#43h716z7qgbv.55321	arn#09\$11-in-xe0..https	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	192.168.43.39.55321	arn#09\$11-in-f61..https	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	dt#43h716z7qgbv.55322	arn#09\$10-in-xe02..https	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	192.168.43.39.55315	a23-43-149-72..de..http	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	192.168.43.39.55314	a23-43-149-72..de..http	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	192.168.43.39.55313	a23-43-149-72..de..http	ESTABLISHED	b0b945eccd98d9 stream	0	0	b0b945eccd98d9	0	0
tcp4	0	0	dt#43h716z7qgbv.55311	pbtng-new-mugdad..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43.39.55310	112.245.178.197..https	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43.39.55301	a23-43-149-72..de..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43.39.55299	62-183-179-26..de..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43.39.55298	62-183-179-26..de..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43.39.55295	151.181.184.134..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43.39.55294	151.181.184.134..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55273	arn#09\$11-in-xe08..https	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43..36.55272	a23-43-149-72..de..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43..36.55271	a23-43-149-72..de..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43..36.55268	a23-43-149-72..de..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43..36.55267	a23-43-149-72..de..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55266	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55265	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43..36.55268	151.181.184.134..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43..36.55265	151.181.184.134..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55259	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55258	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55257	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
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tcp4	0	0	dt#43h716z7qgbv.55255	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55254	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
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tcp4	0	0	dt#43h716z7qgbv.55252	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43..36.55251	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
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tcp4	0	0	dt#43h716z7qgbv.55194	cdu-vip47..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168.43..36.55129	cdu-vip17..hardw..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168..43..36.55024	104.41.267.73..https	FIN_WAIT_1	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	31	0	192.168..43..36.55025	162.125.66.3..https	CLOSE_WAIT	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168..43..36.55196	a23-24-47..dep..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55195	89.167.158.148..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55184	189.169.148..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.55183	189.169.148..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	794	0	192.168..43..36.55129	164.215.116.208..https	FIN_WAIT_1	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168..43..36.55078	138.68..127..https	LAST_ACK	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	31	0	192.168..43..36.55059	d.v..pppppp..com..http	CLOSE_WAIT	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168..43..36.55058	162.125.118..http	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168..43..36.54664	126.128..132..https	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.54668	arn#09\$10-in-xe0..https	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	dt#43h716z7qgbv.54662	442..168..132..https	ESTABLISHED	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168..43..36.54417	ec2-52-45-68-77..https	CLOSE_WAIT	b0b945eccd98f5 stream	0	0	b0b945eccd98f5	0	0
tcp4	0	0	192.168..43..36.54376	do-5.lastpass.co..https	ESTABLISHED	kctl	0	0	1	12	com.apple.network.advisory

Ilkka Lavas, Helsingin Yliopisto, Tietojenkäsittelytieteen laitos

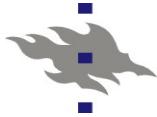


How data is collected?

We all leave traces all the time from
our laptops and mobiles
Spooky or good service?

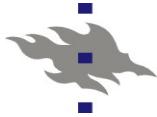
Spooky or good service?

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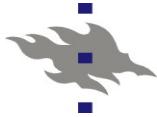
How geospatial data is saved?

- 3 forms
 - Raster data
 - vector data
 - graph data



Data types 1/3: Raster data

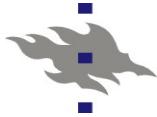
- include geoimages typically obtained by unmanned aerial vehicles, security cameras, and satellites. Recently, the military is collecting huge amounts of raster data by utilizing drones, and the satellites keep providing us with the remote sensing data of the Earth
- The raster data is being provided by digital map services, e.g., Google Earth. Data analysts extract the tracks of moving objects or useful features from these raster data.
- Representative use cases include life pattern mining and change detection.



Data types 2/3: Vector data

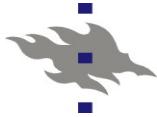
- vector data

- consists of points, lines, and polygons.
- For example, points can be collected through checkin's on
 - Foursquare /
 - Facebook /
 - Google Maps, and
 - lines and polygons correspond to roads in OpenStreetMap

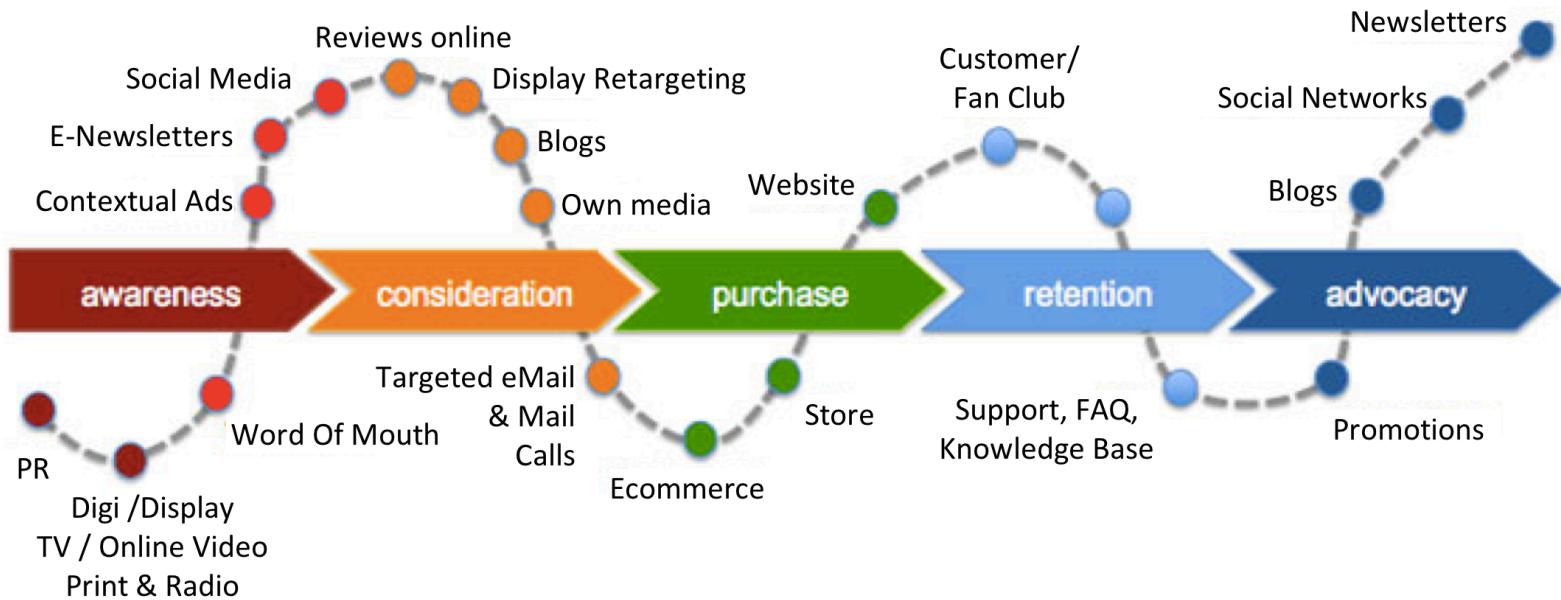


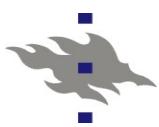
Data types 3/3: Graph Data

- graph data
 - mainly appears in the form of road networks.
 - Here, an edge represents a road segment, and a node represents an intersection or a landmark.
 - The trajectories of vehicles on the road network are
 - represented by sequences of road segments (edges)



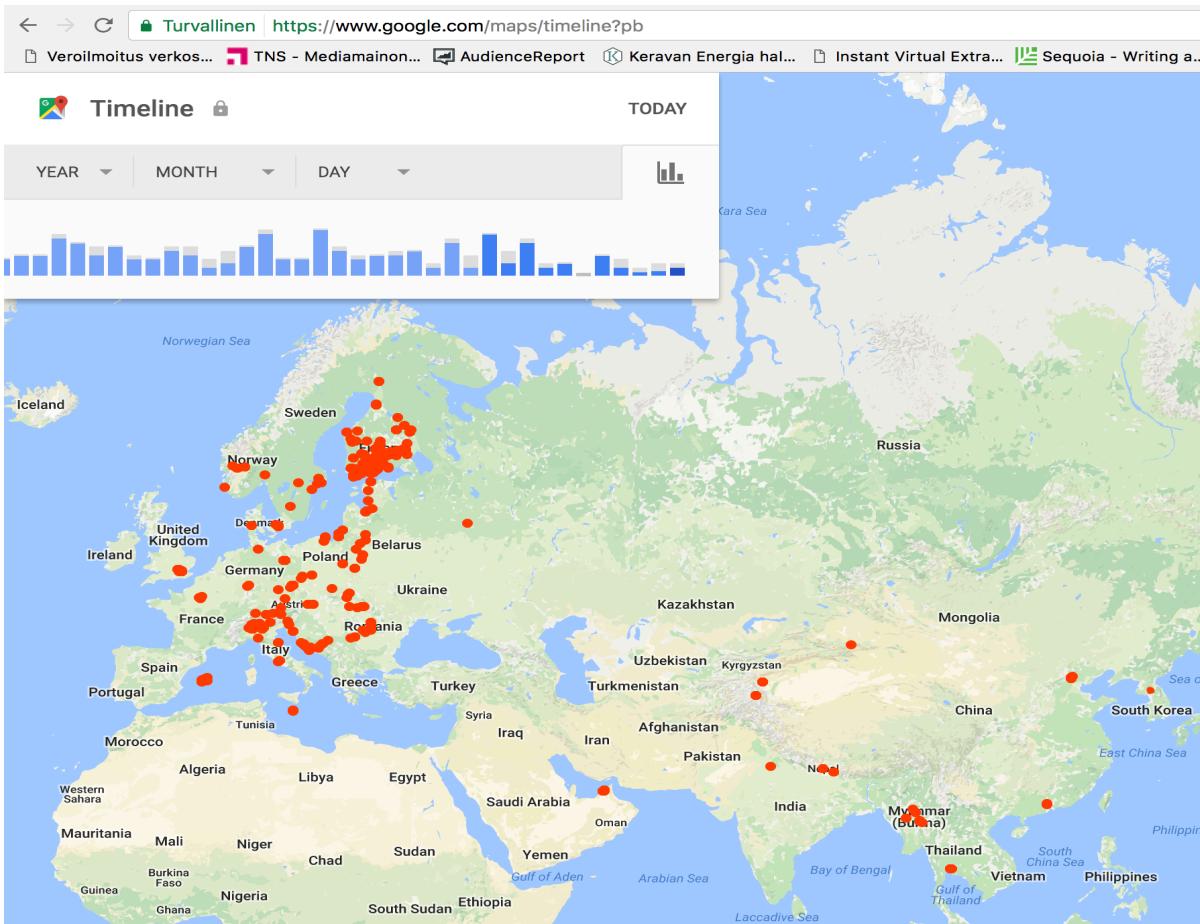
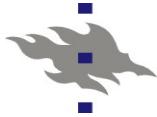
How we leave traces (=data)



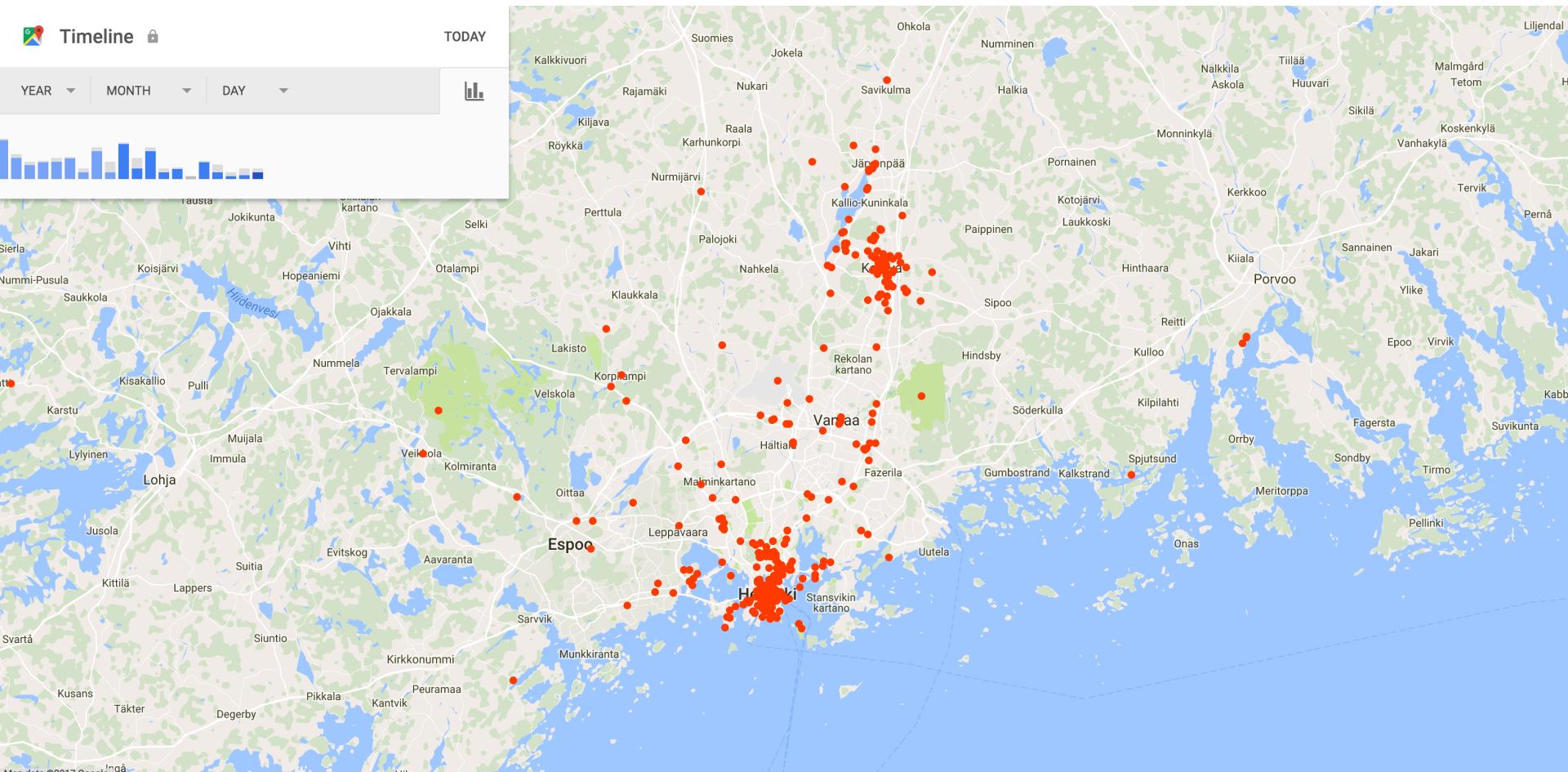


Amount of data sources and real time data is increasing

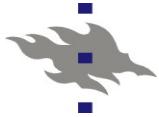
- With the advancements of sensor and communication technologies, new sources of geospatial big data are emerging
 - sensors (or sensor networks) become more prevalent in these days.
 - loop detectors for detecting traffic in roads,
 - electrical grids,
 - environmental sensors for measuring air quality
 - Mobile



Ilkka Lavas, Helsingin Yliopisto, Tietojenkäsittelytieteen laitos

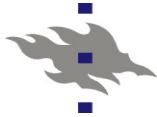


Ikkka Lavas, Helsingin Yliopisto, Tietojenkäsittelytieteen laitos



Collecting data is not key issue anymore

The most important issue
is how we use & exploit
these geospatial big data



Applications

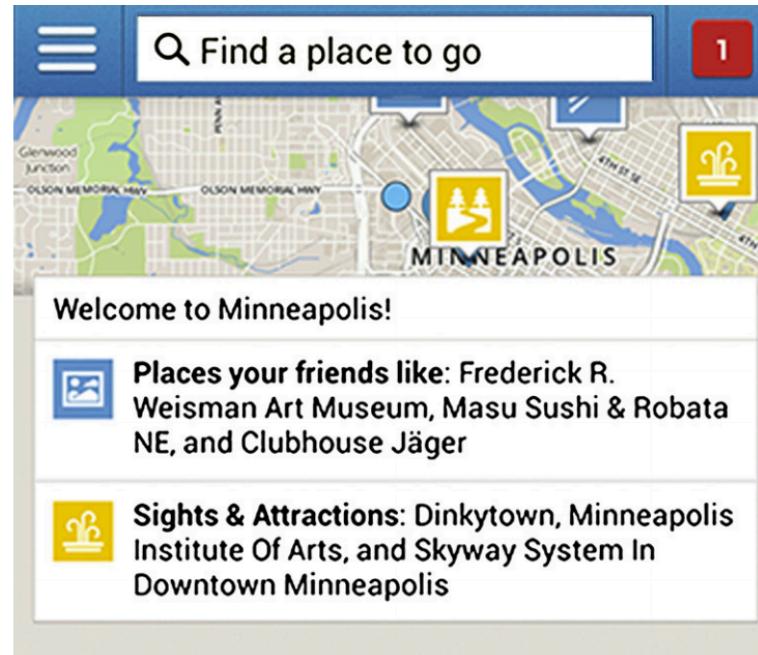


Fig. 6. Foursquare Android app with recommendations.

Ilkka Lavas, Helsingin Yliopisto, Tietojenkäsittelytieteen laitos



TILOA HINTA JÄRJESTÄ

Kaikki Ei väliä Yleisarvosanan mukaan

Tyhjennä Näytä suodattimet

Ravintola	Luokka	Arvo
Tortilla House Kamppi	fresh mex	4.4/5
New Bamboo Center	kiinalais-malesialainen	4.2/5
Brasserie Le Havre	ranskalaisia	4.2/5
STADIN EKA CHILIBAARI Papiito	chilibaari	4.1/5
Chilli Asematunneli	pikakebab	3.9/5
India House	intialaisia, suomalaista	3.9/5
Ravintola Lasipalatsi	ruokaravintola	3.8/5
Rulla	riisipiperirolla ja viini	3.8/5
Suburritos		3.7/5
Kippo Kukontori		3.7/5

Kasvisruoka Lapsistävällinen Kasvisruoka

Kasvisruoka Lapsistävällinen Kasvisruoka

Kasvisruoka

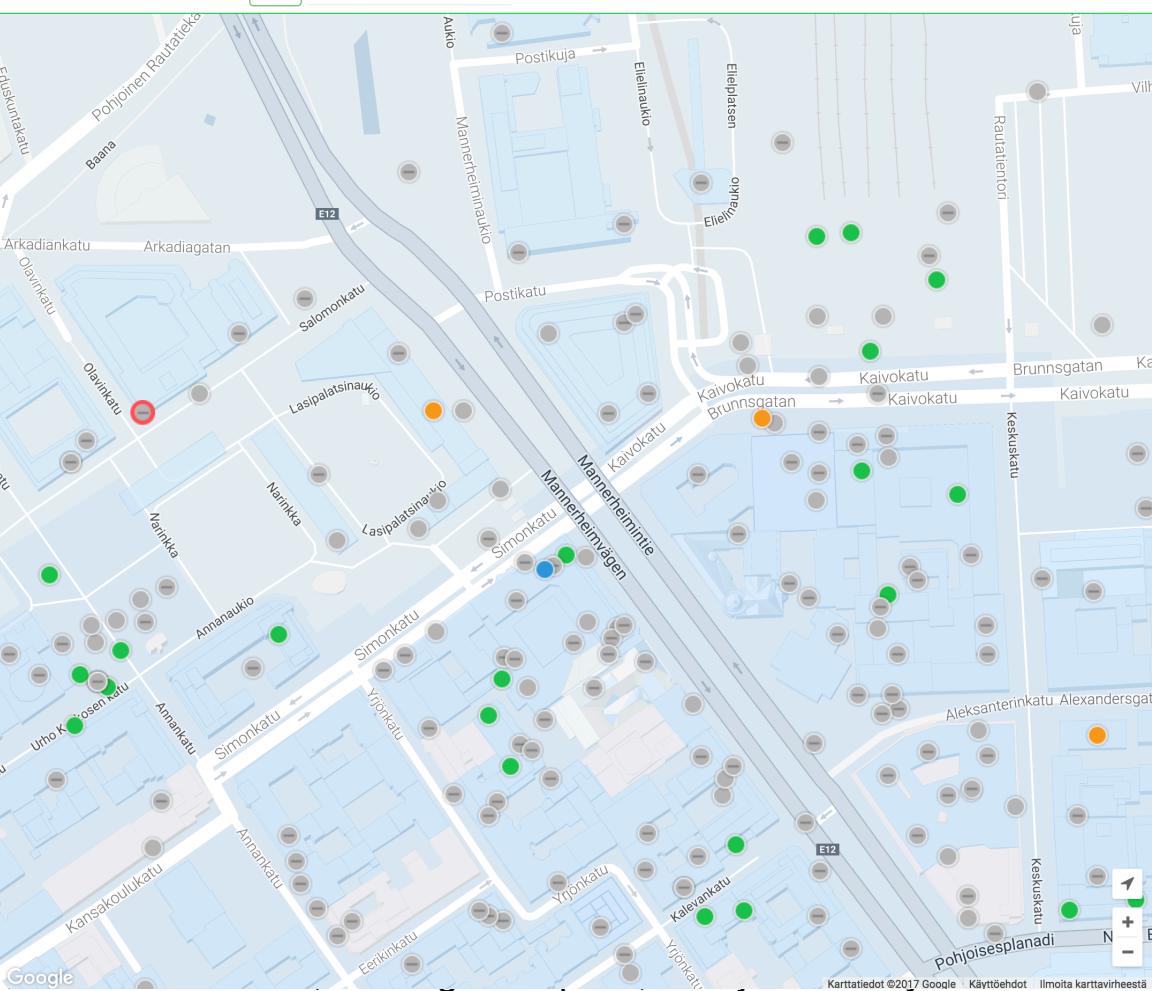
Kasvisruoka

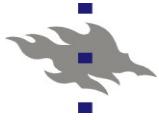
Kasvisruoka Lapsistävällinen Kasvisruoka

Kasvisruoka

Lapsistävällinen Kasvisruoka Unicef

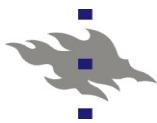
Kasvisruoka





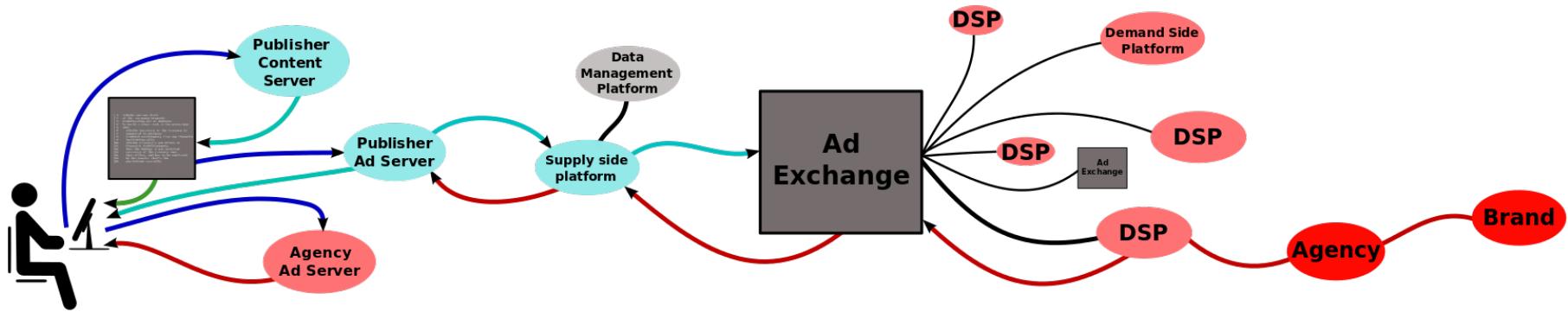
Advertisement Automation with Big Data

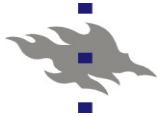
- Data can be used to provide better advertising



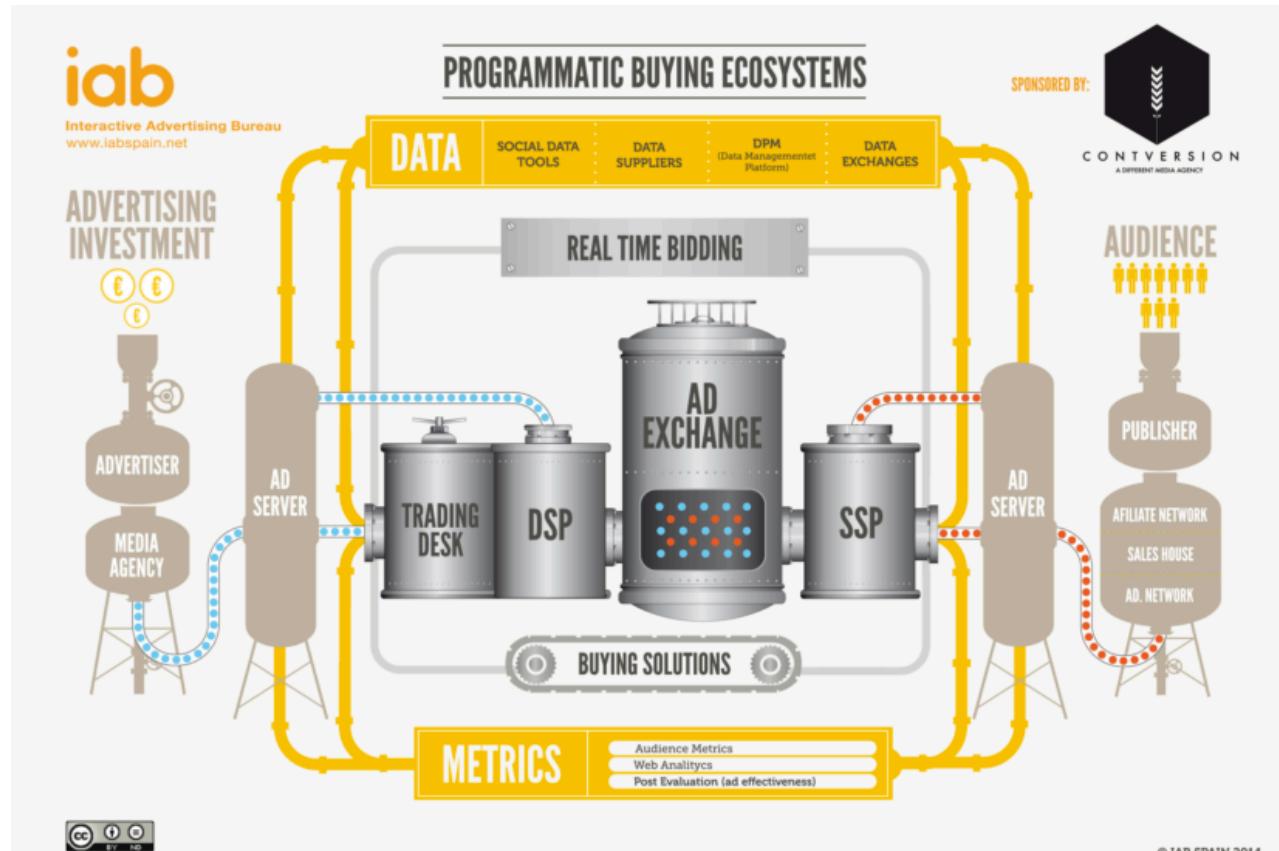
Advertising

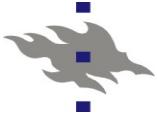
Programmatic marketing ecosystem





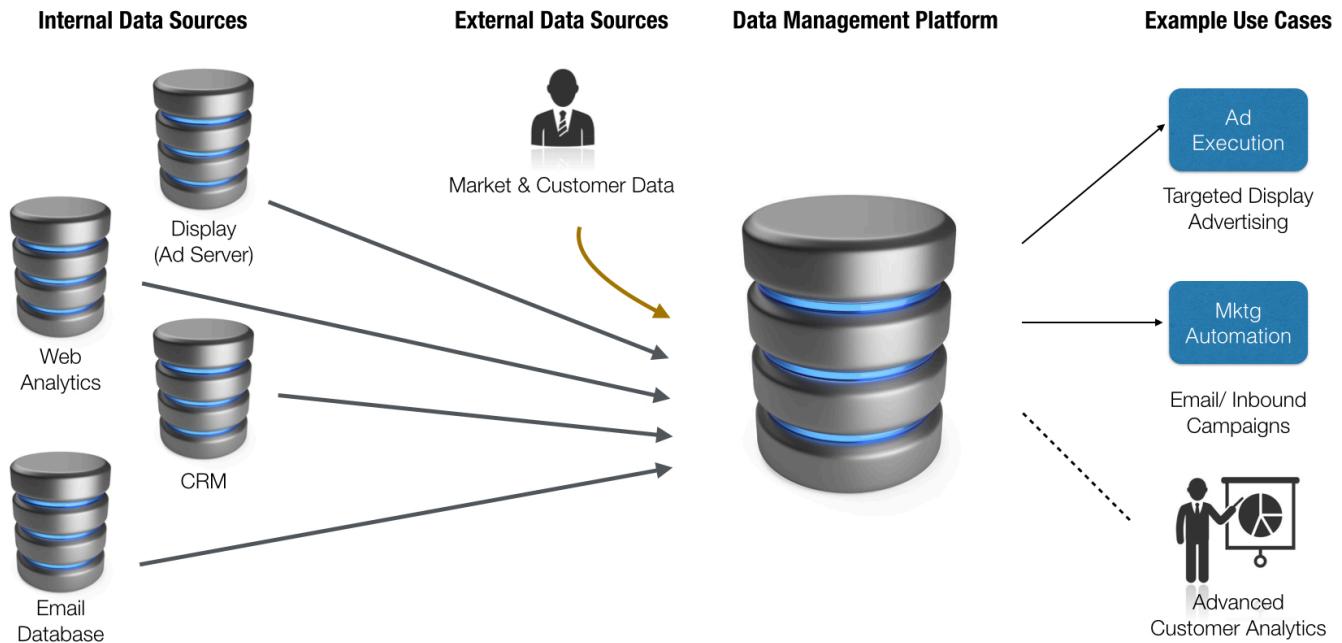
Advertising Programmatic marketing ecosystem

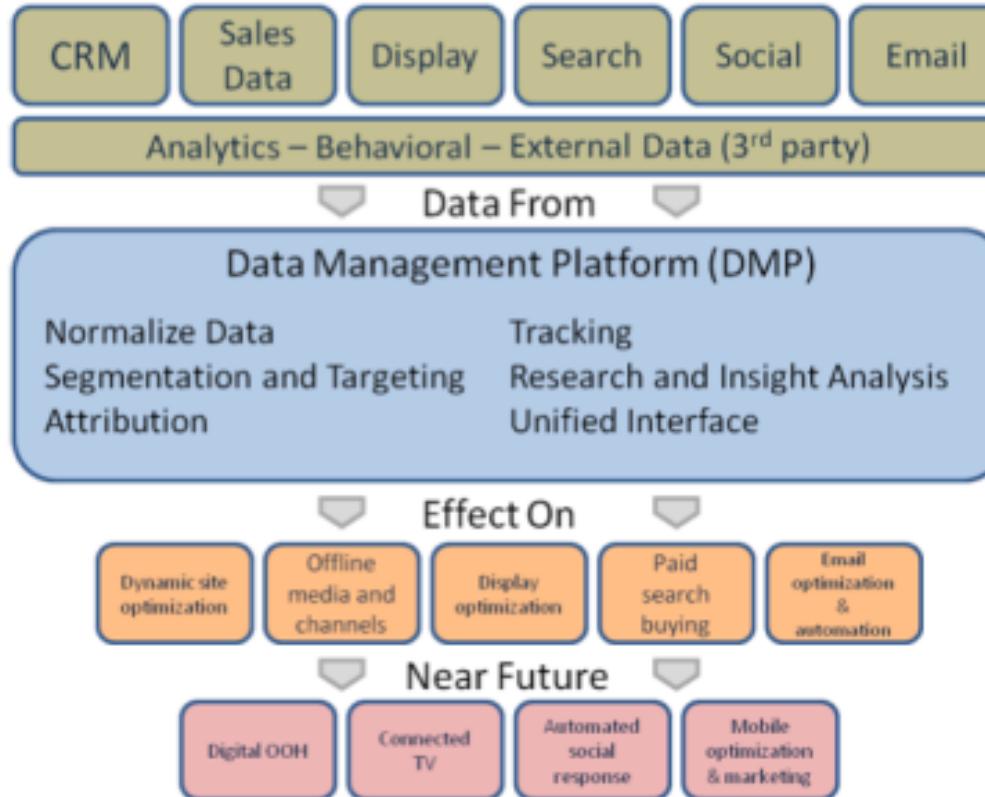




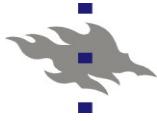
Data Management Platforms

Approach to collecting, organizing and activating customer data

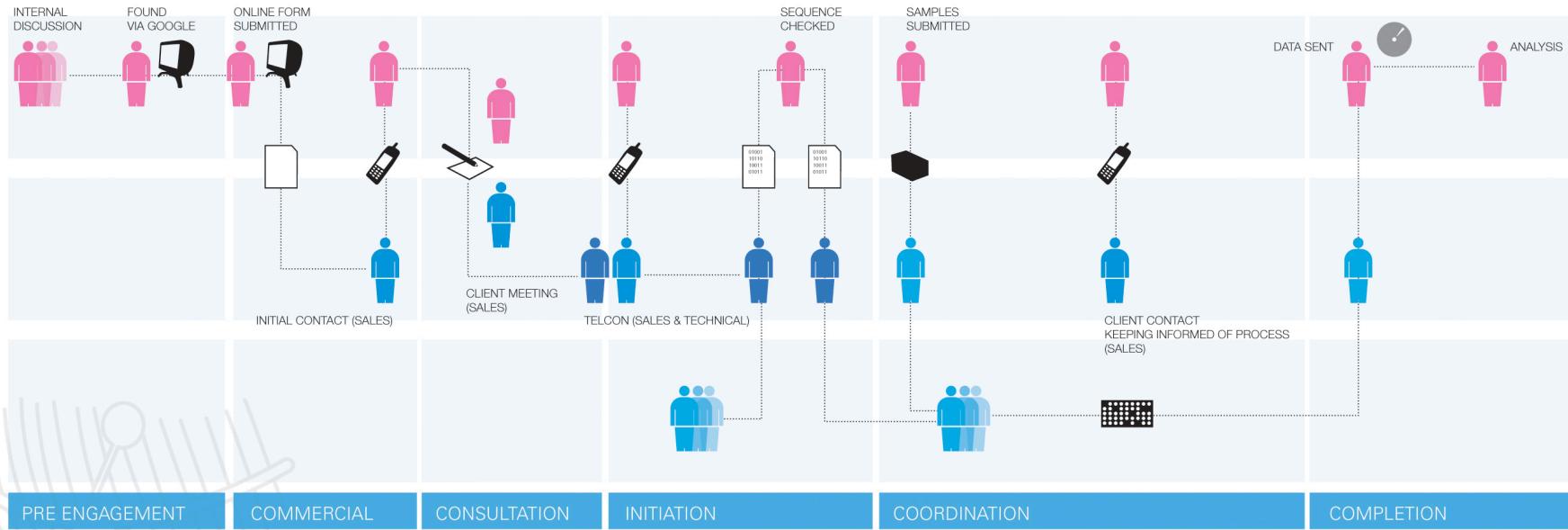


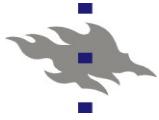


Source: Econsultancy

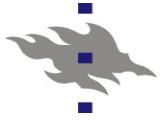


Customer path

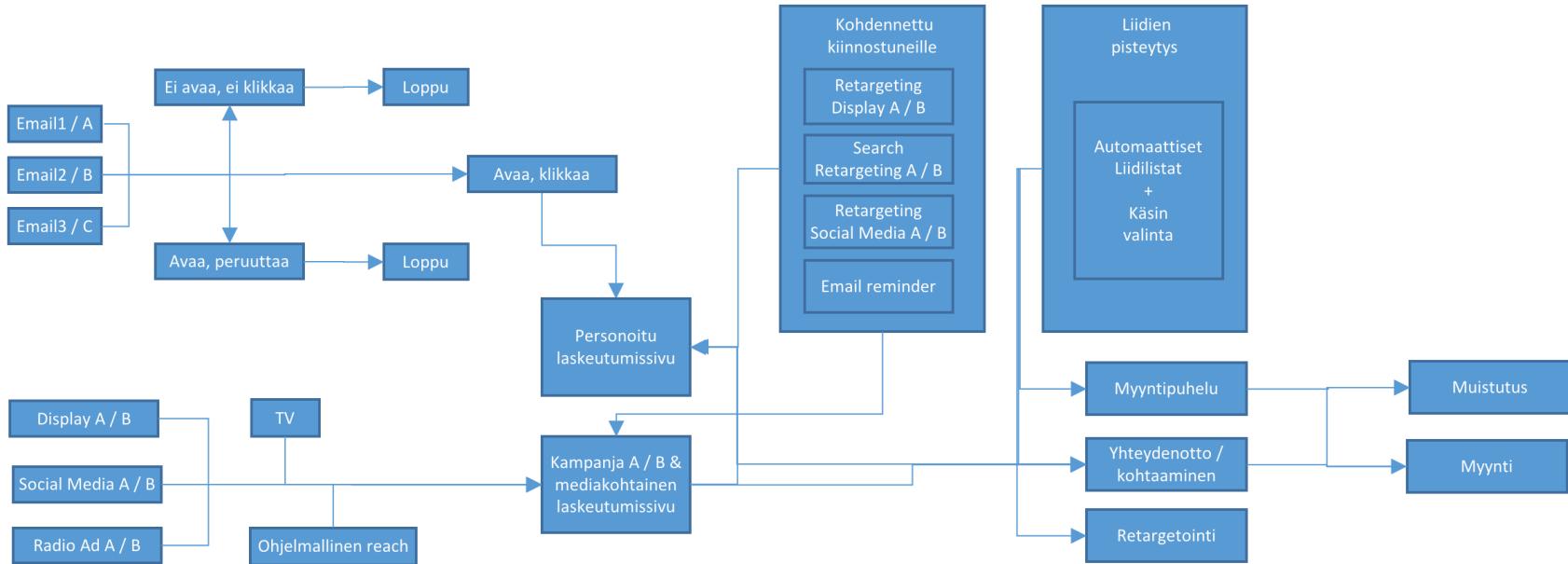


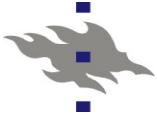


The future of advertising will be data driven



Continuous marketing w/ Big Data





Open problems

- General Data Protection Regulation:
REGULATION (EU) 2016/679 OF THE EUROPEAN
PARLIAMENT AND OF THE COUNCIL of 27 April 2016
on the protection of natural persons with regard to the
processing of personal data and on the free movement of
such data, and repealing Directive 95/46/EC

8960.74

Outotec
Atria A

+4.40% Oriola-KD A
+4.24% Neo Industrial

-3.49%
-2.50%

ma 13.2. 5:00
Kauppalehden toimitus



LUE JA KOMMENTOI ▶

Kauppalehti
Blogit

KAUPALLINEN YHTEISTYÖ

"Mikä hidastaa B2B-verkkokaupan yleistymistä?"

Jari Valtanen, Toimitusjohtaja, Rocla Solutions Oy

Kauppalehti

Uutiset

Pörssi

Yritykset

Keskustelu

Blogit

Kirjaudu



KL NYT

Tärkeimmät uutiset ja terävimmät näkemykset nyt.



13.2. 13:38

KOROT

Euriborit laskussa, 12 kk euriborin muutos -0,001 prosenttiyksikköä

13.2. 13:16 (pv)

SJOITAMINEN

Tokmannin suuromistaja osti osakkeita 1,8 miljoonalla eurolla

13.2. 13:14

KOROT

Euriborit laskussa, 12 kk euriborin muutos -0,001 prosenttiyksikköä

13.2. 13:14

AUTO



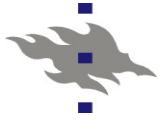
Henri Elo: Yhtiökokoukset alkavat ja niin myös osakkeiden mitätöinti

Pörssiolumnisti ja sijoittaja Henri Elo jakaa vinkkejä tuleviin yhtiökokouksiin.

TULOKSET / PÖRSI / SJOITAMINEN

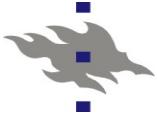
Käytämme Alma Median sivustoilla evästeitä. Jatkamalla hyväksyt evästeiden käytön. [Lisätietoja](#).

SELVÄ



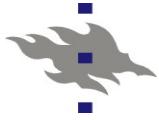
Summary & Conclusion

- Big Data consist of collection, analysis and usage
- Collection is not anymore an issue
- Analysis needs a lot of work
- Right use gives more value to brand, customer, user and ecosystem
 - Better advertising
 - Less intrusive ads
 - More relevancy in ads
 - What before was spooky, will be new standard and we expect to be served according to data traces we leave:
"How nice that you think of me"

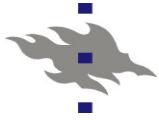


Sources

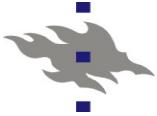
- Paul Suganthan G. C., Chong Sun, Krishna Gayatri K.,
Haojun Zhang, Frank Yang, Narasimhan Rampalli, Shishir
Prasad, Esteban Arcaute, Ganesh Krishnan, Rohit Deep,
Vijay Raghavendra, AnHai Doan: Why Big Data Industrial
Systems Need Rules and What We Can Do About It.
SIGMOD Conference 2015: 265-276 [pdf paper]
- <https://www.cs.helsinki.fi/u/jilu/paper/bigdataapplication05.pdf>



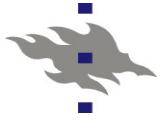
- (2) Javier Andréu Pérez, Carmen C. Y. Poon, Robert D. Merrifield, Stephen T. C. Wong, Guang-Zhong Yang: Big Data for Health. IEEE J. Biomedical and Health Informatics 19(4): 1193-1208 (2015) [pdf paper]
- <https://www.cs.helsinki.fi/u/jilu/paper/bigdataapplication01.pdf>



- (3) Jae-Gil Lee, Minseo Kang: Geospatial Big Data: Challenges and Opportunities. *Big Data Research* 2(2): 74-81 (2015) [pdf paper]
- <https://www.cs.helsinki.fi/u/jilu/paper/bigdataapplication02.pdf>



- (4) Taruna Seth, Vipin Chaudhary: Big Data in Finance.
Big Data - Algorithms, Analytics, and Applications 2015:
329-356 [pdf paper]



- (5) Kesheng Wu, E. Wes Bethel, Ming Gu, David Leinweber, Oliver Rübel: A big data approach to analyzing market volatility. Algorithmic Finance 2(3-4): 241-267 (2013) [pdf paper]