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TOPIC

TRANSPORTATION - SENSORS - AND THE SMART CITY

GEOG 491/591 ADVANCED GIS

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY

- **DATA FROM SENSORS**
What part does data play in the ecosystem of the smart city?
- **BIG DATA**
How does Big Data fit into a “Sensored Smart City”
- **BALANCE**
balance the practical, nuts and bolts aspects of data with the conceptual and philosophical

TOPIC

GOALS - OBJECTIVES - OUTCOMES

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#3 - DATA FOR THE SMART CITY

Data

From Wikipedia, the free encyclopedia

For data in computer science, see [Data \(computing\)](#). For other uses, see [Data \(disambiguation\)](#).

Data (/dətə/ *DAY-tə*, /dætə/ *DA-tə*, or /dɑ:tə/ *DAH-tə*)^[1] is a set of values of qualitative or quantitative variables; restated, pieces of data are individual pieces of information. Data is measured, collected and reported, and analyzed, whereupon it can be visualized using graphs or images. Data as a general concept refers to the fact that some existing information or knowledge is represented or coded in some form suitable for better usage or processing.

Raw data, i.e., unprocessed data, refers to a collection of numbers, characters and is a relative term; data processing commonly occurs by stages, and the "processed data" from one stage may be considered the "raw data" of the next. **Field data** refers to raw data that is collected in an uncontrolled *in situ* environment. **Experimental data** refers to data that is generated within the context of a scientific investigation by observation and recording.

The word "data" has been considered the plural of "datum" until recently, but is now generally used in the singular as a mass noun.^[2][dubious – discuss]

Contents [hide]

- 1 Meaning of data, information and knowledge
- 2 In other fields
- 3 See also
- 4 References
- 5 External links

Meaning of data, information and knowledge [edit]

Data, [information](#) and [knowledge](#) are closely related terms, but each has its own role in relation to the other. Data is collected and analyzed to create information suitable for making decisions,^[3] while [knowledge](#) is derived from extensive amounts of experience dealing with information on a subject. For example, the height of [Mt. Everest](#) is generally considered data. This data may be included in a book along with other data on Mt. Everest to describe the mountain in a manner useful for those who wish to make a decision about the best method to climb it. Using an understanding based on experience climbing mountains to advise persons on the way to reach Mt. Everest's peak may be seen as "knowledge".

TOPIC

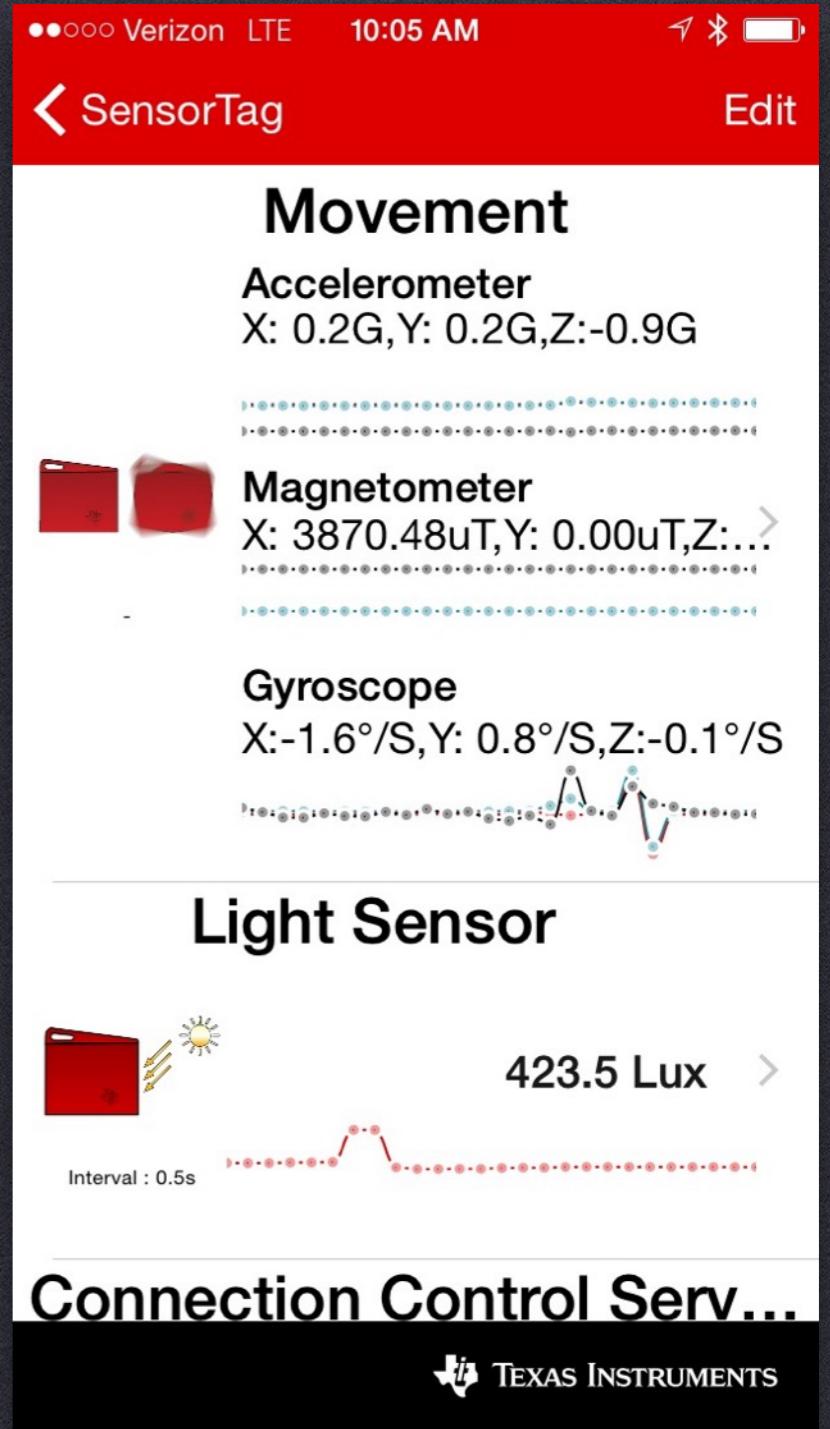
DATA

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#3 - DATA FOR THE SMART CITY



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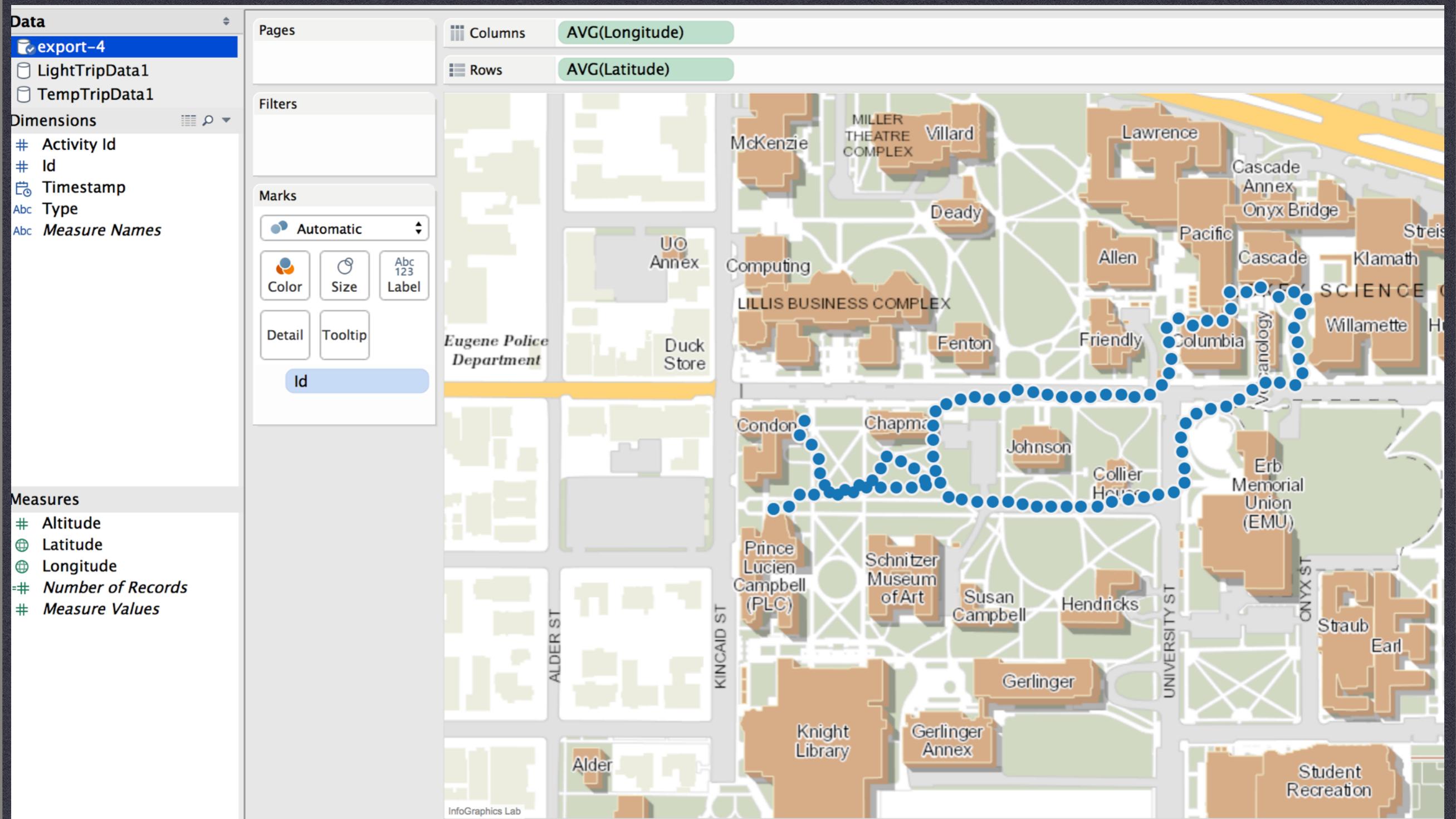
A SENSOR EXPERIMENT

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A SENSOR EXPERIMENT

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Viewing live data

7a:a3:c5:c2:8d:53
Device connected [Change device](#)
at 10:36:14 AM
You cannot register this device.

What's next?

[Sign up](#)

Sign up provides the option of a free trial through IBM Bluemix or the IBM Marketplace

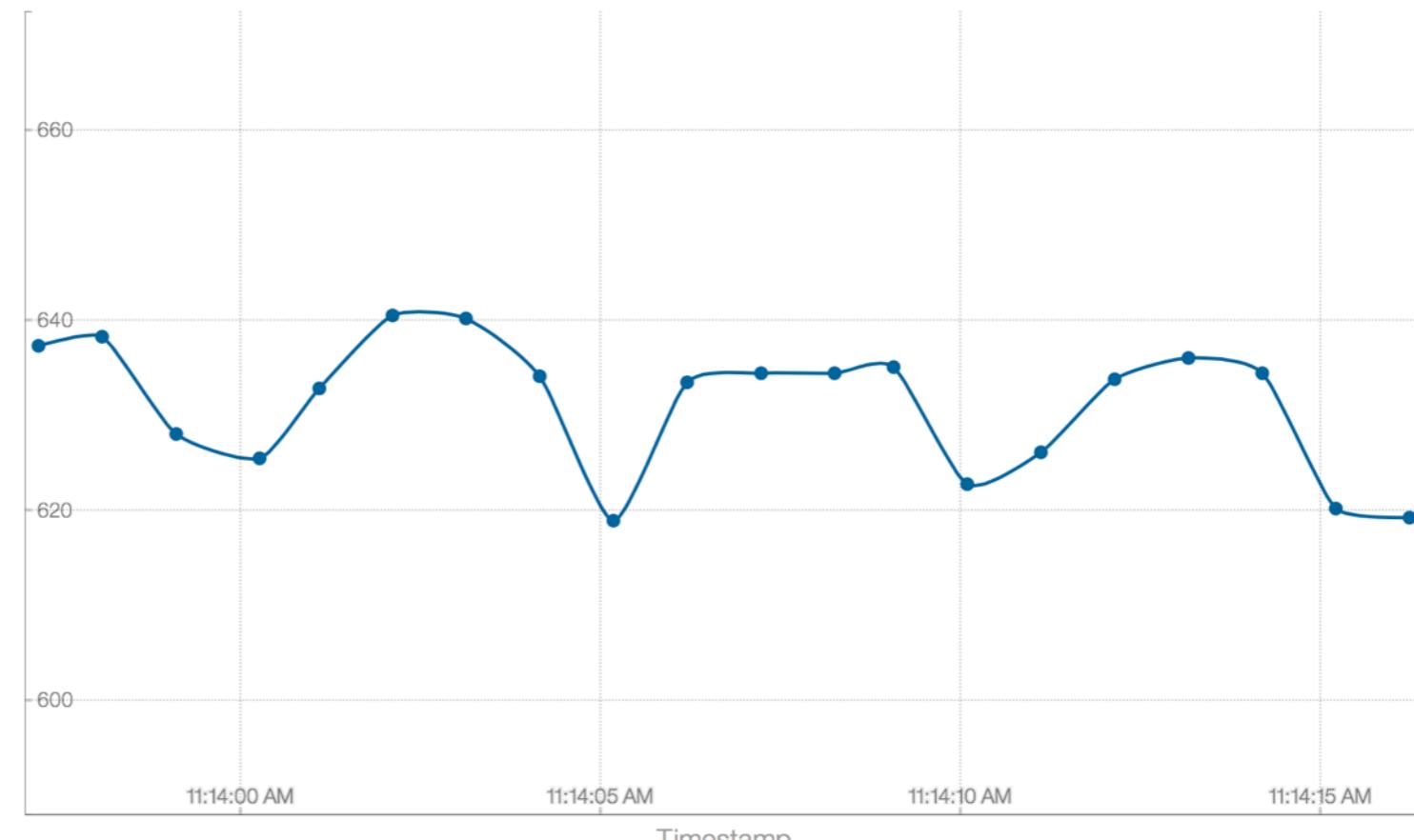
Once signed up you can

- Communicate securely with your devices
- Access historical as well as real-time device data
- Send commands to your devices

Already signed up for IoT Foundation?

[Log In](#)

SensorTag 2.0



Sensor	Reading	Sparkline
AmbTemp	24.4063	
IRTemp	19.3125	

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A SENSOR EXPERIMENT

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#3 - DATA FOR THE SMART CITY

- THE REAL-TIME CITY? BIG DATA AND SMART URBANISM
GEOJOURNAL
- BIG DATA BASIC CONCEPTS AND BENEFITS EXPLAINED
TechRepublic

TOPIC

ASSIGNED READINGS FOR TODAY

DATE

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LECTURE

#3 - DATA FOR THE SMART CITY

- **OPEN DATA VERSUS CLOSED DATA**

Informing internally versus opening to the public

- **RAW VERSUS VISUALIZED DATA**

What exactly are you sharing? How useful is it?

- **HISTORICAL CHALLENGES — QUANTITATIVE DATA AND QUALITATIVE DATA**

relied on samples, are generated on a non-continuous basis, the number of variables are quite small, are aggregated to a relatively coarse spatial scale, and are often limited in access.

questionnaire surveys, case studies, city audits, inter- views and focus groups, and ethnographies—that capture a relatively limited sample of data that are tightly focused, time and space specific, restricted in scope and scale, and relatively expensive to generate and analyze, t

- **DEFINITION**

big data consists of massive, dynamic, varied, detailed, inter-related, low cost datasets that can be connected and utilized in diverse ways, thus offering the possibility of studies shifting from: data-scarce to data-rich; static snapshots to dynamic unfoldings; coarse aggregation to high resolution; relatively simple hypotheses and models to more complex, sophisticated simulations and theories

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SMART CITY DATA - CONCEPTS (KITCHIN ARTICLE)

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#3 - DATA FOR THE SMART CITY

- **POLITICS OF BIG URBAN DATA**

“raw data is an oxymoron” - Data inflected by social privilege and social values

- **TECHNOCRATIC GOVERNANCE AND CITY DEVELOPMENT**

“It’s not me , it’s the data” (Haque).

- **CORPORATIZATION OF CITY GOVERNANCE AND CORPORATE LOCK-IN**

“smart cities are currently more vendor push than city government pull based” (Shaffers)
The comparison to US auto industry and urban development.

- **BUGGY, BRITTLE AND HACKABLE CITIES**

“What if the seeds of smart cities’ own destruction are already built into their DNA? What if the smart cities of the future are buggy and brittle? What are we getting ourselves into?”

- **THE PANOPTIC CITY**

Big Brother?

TOPIC

SMART CITY DATA - CHALLENGES (KITCHIN ARTICLE)

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#3 - DATA FOR THE SMART CITY

- **VOLUME**

90% of all data ever created, was created in the past 2 years.

- **VELOCITY**

Velocity is the speed at which the data is created, stored, analyzed and visualized.

- **VARIETY**

The wide variety of data requires a different approach as well as different techniques to store all raw data.

- **VERACITY**

Is the data reliable - accurate - actionable - correct

- **VARIABILITY**

Meaning is changing (rapidly)

- **VISUALIZATION**

Making all that vast amount of data comprehensible in a manner that is easy to understand and read

- **VALUE**

What's it worth? The value is in the analyses done on that data and how the data is turned into information and eventually

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BIG DATA - SEVEN V'S

Why The 3V's Are Not Sufficient To Describe Big Data, Mark van Rijmenam, 2013

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#3 - DATA FOR THE SMART CITY

● VOLUME

...the amount of data in the world will double every two years. By 2020, we will have 50 times the amount of data as that we had in 2011. The sheer volume of the data is enormous and a very large contributor to the ever expanding digital universe is the Internet of Things with sensors all over the world in all devices creating data every second. The era of a trillion sensors is upon us.

If we look at airplanes they generate approximately 2.5 billion Terabyte of data each year from the sensors installed in the engines. Self-driving cars will generate 2 Petabyte of data every year. Also the agricultural industry generates massive amounts of data with sensors installed in tractors. Shell uses super-sensitive sensors to find additional oil in wells and if they install these sensors at all 10.000 wells they will collect approximately 10 Exabyte of data annually. That again is absolutely nothing if we compare it to the Square Kilometer Array Telescope that will generate 1 Exabyte of data per day.

In the past, the creation of so much data would have caused serious problems. Nowadays, with decreasing storage costs, better storage solutions like Hadoop and the algorithms to create meaning from all that data this is not a problem at all.

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	A	B	C	D	E	F	G	H	I	J	K
1	People	Sensor Count	Sensor Type	Second	Minute	Hour	Day	Week	Month	Year	
2	1	1	Light	0.040666667	2.44	146.4	3513.6	24595.2	98380.8	1180569.6	
3	2	1	Light	0.081333333	4.88	292.8	7027.2	49190.4	196761.6	2361139.2	
4	4	1	Light	0.325333333	19.52	1171.2	28108.8	196761.6	787046.4	9444556.8	
5	16	1	Light	5.205333333	312.32	18739.2	449740.8	3148185.6	12592742.4	151112908.8	
6	256	1	Light	1332.565333	79953.92	4797235.2	115133644.8	805935514	3223742054	38684904653	
7	65536	1	Light	87331001.69	5239.860101	314391.6061	7545398.546	52817789.8	211271159.3	2535253911	GB
8	4294967296	1	Light	3.75084E+17	22505027.77	1350301666	32407239989	226850680	907402719.7	10888832636	TB
9	1.84467E+19	1	Light	6.91907E+36	4.15144E+17	2.49087E+28	5.97808E+29	4.1847E+27	1.67386E+28	2.00864E+29	
10	3.40282E+38	1	Light	2.35444E+75	1.41266E+56	8.47598E+66	2.03424E+68	1.424E+66	5.69586E+66	6.83503E+67	
11	1.15792E+77	1	Light	2.7263E+152	1.6358E+133	9.8145E+143	2.3555E+145	1.649E+143	6.5954E+143	7.9144E+144	
12											
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14											
15											

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BIG DATA - VOLUME

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● **VELOCITY**

The Velocity is the speed at which the data is created, stored, analyzed and visualized. In the past, when batch processing was common practice, it was normal to receive an update from the database every night or even every week. Computers and servers required substantial time to process the data and update the databases. In the big data era, data is created in real-time or near real-time. With the availability of Internet connected devices, wireless or wired, machines and devices can pass-on their data the moment it is created.

The speed at which data is created currently is almost unimaginable: Every minute we upload 100 hours of video on Youtube. In addition, every minute over 200 million emails are sent, around 20 million photos are viewed and 30.000 uploaded on Flickr, almost 300.000 tweets are sent and almost 2,5 million queries on Google are performed.

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- **VARIETY**

In the past, all data that was created was structured data, it neatly fitted in columns and rows but those days are over. Nowadays, 90% of the data that is generated by organization is unstructured data. Data today comes in many different formats: structured data, semi-structured data, unstructured data and even complex structured data. The wide variety of data requires a different approach as well as different techniques to store all raw data.

There are many different types of data and each of those types of data require different types of analyses or different tools to use. Social media like Facebook posts or Tweets can give different insights, such as sentiment analysis on your brand, while sensory data will give you information about how a product is used and what the mistakes are.

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LightTripData1.csv

	A	B	C	D	E	F	G
178	4/3/15 10:37	154.4					
179	4/3/15 10:37	155.1					
180	4/3/15 10:37	160.6					
181	4/3/15 10:37	163					
182	4/3/15 10:37	166.2					
183	4/3/15 10:37	167					
184	4/3/15 10:37	163.5					
185	4/3/15 10:37	160					
186	4/3/15 10:37	160					
187	4/3/15 10:37	160.5					
188	4/3/15 10:37	161					
189	4/3/15 10:37	161.1					
190	4/3/15 10:37	161.1					
191	4/3/15 10:37	161					
192	4/3/15 10:37	161					
193	4/3/15 10:37	161					
194	4/3/15 10:37	160.6					
195	4/3/15 10:37	160.6					
196	4/3/15 10:37	161					
197	4/3/15 10:37	162.9					
198	4/3/15 10:37	168.6					
199	4/3/15 10:37	172.2					
200	4/3/15 10:37	176.8					
201	4/3/15 10:37	181.3					
202	4/3/15 10:37	181.6					
203	4/3/15 10:37	200.8					
204	4/3/15 10:37	219.8					
205	4/3/15 10:37	248.7					
206	4/3/15 10:37	278.7					
207	4/3/15 10:37	279.8					
208	4/3/15 10:37	288.5					
209	4/3/15 10:38	297.1					
210	4/3/15 10:38	148.6					
211	4/3/15 10:38	0.1					
212	4/3/15 10:38	0.1					
213	4/3/15 10:38	80.5					
214	4/3/15 10:38	160.5					
215	4/3/15 10:38	160					
216	4/3/15 10:38	160.6					

LightTripData1.csv

	A	B	C	D	E	F	G	H
1	Id	ActivityId	Timestamp	Latitude	Longitude	Altitude	Type	
2		1	540732196	0	44.045318	-123.07793	135.730966	start
3		2	540732196	12.0269206	44.045232	-123.07797	136.027982	gps
4		3	540732196	20.0879846	44.045169	-123.07786	136.395443	gps
5		4	540732196	25.0294575	44.045083	-123.0778	136.856539	gps
6		5	540732196	33.0836148	44.04499	-123.07779	137.192921	gps
7		6	540732196	42.0208181	44.044902	-123.07774	137.373266	gps
8		7	540732196	49.9173852	44.044847	-123.07763	137.56561	gps
9		8	540732196	57.0878653	44.044864	-123.07749	137.838498	gps
10		9	540732196	64.0414666	44.044888	-123.07737	138.030109	gps
11		10	540732196	73.0534728	44.044895	-123.07724	138.29139	gps
12		11	540732196	82.0156465	44.044892	-123.0771	138.319483	gps
13		12	540732196	91.0227354	44.04489	-123.07696	138.221066	gps
14		13	540732196	99.0729927	44.044913	-123.07683	138.114615	gps
15		14	540732196	107.022555	44.044998	-123.07675	137.88823	gps
16		15	540732196	115.033602	44.045094	-123.07677	137.567979	gps
17		16	540732196	125.088812	44.045195	-123.07676	137.234527	gps
18		17	540732196	132.019504	44.045287	-123.07677	137.073594	gps
19		18	540732196	138.026347	44.045379	-123.07675	137.051832	gps
20		19	540732196	144.027632	44.045438	-123.07665	137.026414	gps
21		20	540732196	149.014103	44.045463	-123.07652	137.189517	gps
22		21	540732196	156.042277	44.045475	-123.0764	137.241807	gps
23		22	540732196	164.013975	44.045468	-123.07627	137.583232	gps
24		23	540732196	174.014337	44.04548	-123.07612	138.028228	gps
25		24	540732196	180.036365	44.045525	-123.07601	138.373477	gps
26		25	540732196	188.017956	44.045511	-123.07587	138.591117	gps
27		26	540732196	193.058234	44.045498	-123.07574	138.726504	gps
28		27	540732196	200.094243	44.045478	-123.07561	138.825392	gps
29		28	540732196	206.053038	44.045473	-123.07549	138.814955	gps
30		29	540732196	214.205534	44.045474	-123.07535	138.788854	gps
31		30	540732196	221.002581	44.045487	-123.07521	138.768746	gps
32		31	540732196	227.029392	44.045488	-123.07508	138.731936	gps
33		32	540732196	232.033267	44.045477	-123.07495	138.651053	gps
34		33	540732196	238.034987	44.0455	-123.07482	138.528678	gps
35		34	540732196	244.039194	44.04556	-123.07471	138.263774	gps
36		35	540732196	254.193152	44.04564	-123.07464	138.074521	gps
37		36	540732196	261.162273	44.045729	-123.07463	138.078699	gps
38		37	540732196	267.087858	44.045829	-123.07465	138.107785	gps
39		38	540732196	273.056652	44.045933	-123.07466	138.17950	gps

export-4.csv

• **VISUALIZATION**

This is the hard part of big data. Making all that vast amount of data comprehensible in a manner that is easy to understand and read. With the right analyses and visualizations, raw data can be put to use otherwise raw data remains essentially useless. Visualizations of course do not mean ordinary graphs or pie charts. They mean complex graphs that can include many variables of data while still remaining understandable and readable.

Visualizing might not be the most technologically difficult part; it sure is the most challenging part. Telling a complex story in a graph is very difficult but also extremely crucial. Luckily there are more and more big data startups appearing that focus on this aspect and in the end, visualizations will make the difference. Visualizations help organizations answer questions they did not know to ask.

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BIG DATA - SEVEN V'S

Why The 3V's Are Not Sufficient To Describe Big Data, Mark van Rijmenam, 2013

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#3 - DATA FOR THE SMART CITY



Home Layout Tables

Edit
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A89

	A	B	C
3	4/3/15 10:36	21	
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5	4/3/15 10:36	21.2	
6	4/3/15 10:36	21.4	
7	4/3/15 10:36	21.2	
8	4/3/15 10:36	21	
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12	4/3/15 10:36	19.2	
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18	4/3/15 10:36	21.2	
19	4/3/15 10:36	23	
20	4/3/15 10:36	23	
21	4/3/15 10:36	23	
22	4/3/15 10:36	92.6	
23	4/3/15 10:36	162.2	
24	4/3/15 10:36	113.4	
25	4/3/15 10:36	64.6	
26	4/3/15 10:36	64.6	
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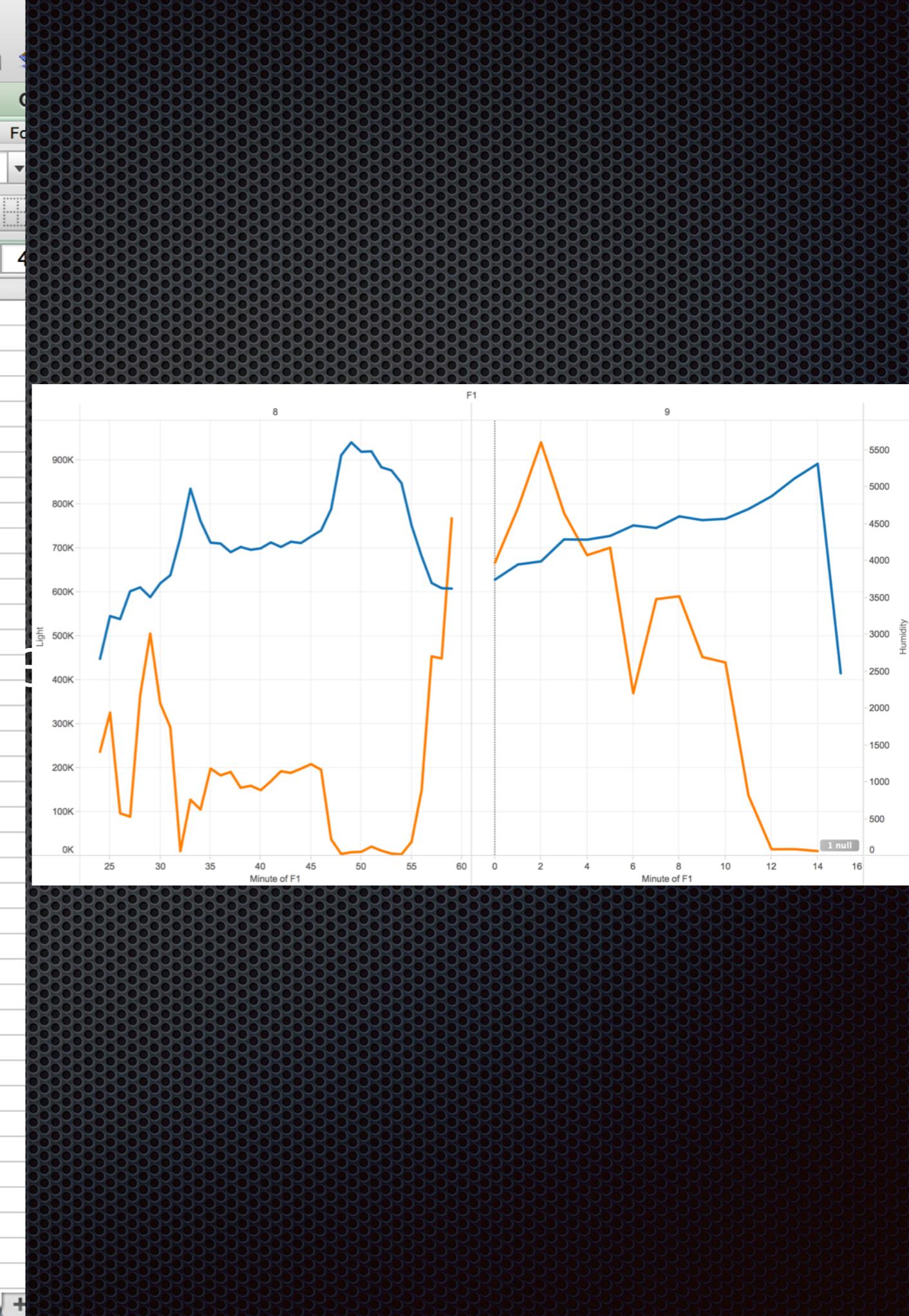


Home Layout Tables

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A1

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5	4/3/15 8:24	1010.5	
6	4/3/15 8:24	1010.5	
7	4/3/15 8:24	1010.5	
8	4/3/15 8:24	1010.1	
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13	4/3/15 8:24	1010.5	
14	4/3/15 8:24	1010.5	
15	4/3/15 8:24	1010.1	
16	4/3/15 8:24	1010.5	
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23	4/3/15 8:24	1010.1	
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36	4/3/15 8:24	1010.5	
37	4/3/15 8:24	1010.5	
38	4/3/15 8:24	1010.1	
39	4/3/15 8:24	1010.5	



● VERACITY

Having a lot of data in different volumes coming in at high speed is worthless if that data is incorrect. Incorrect data can cause a lot of problems for organizations as well as for consumers. Therefore, organizations need to ensure that the data is correct as well as the analyses performed on the data are correct. Especially in automated decision-making, where no human is involved anymore, you need to be sure that both the data and the analyses are correct.

If you want your organization to become information-centric, you should be able to trust that data as well as the analyses. accountability.

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BIG DATA - SEVEN V'S

Why The 3V's Are Not Sufficient To Describe Big Data, Mark van Rijmenam, 2013

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● VARIABILITY

Big data is extremely variable. Brian Hopkins, a Forrester principal analyst, defines variability as the “variance in meaning, in lexicon”. He refers to the supercomputer Watson who won Jeopardy. The supercomputer had to “dissect an answer into its meaning and [...] to figure out what the right question was”. That is extremely difficult because words have different meanings and all depends on the context. For the right answer, Watson had to understand the context.

Variability is thus very relevant in performing sentiment analyses. Variability means that the meaning is changing (rapidly). In (almost) the same tweets a word can have a totally different meaning. In order to perform a proper sentiment analyses, algorithms need to be able to understand the context and be able to decipher the exact meaning of a word in that context. This is still very difficult.

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BIG DATA - SEVEN V'S

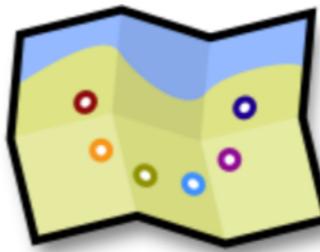
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#3 - DATA FOR THE SMART CITY



We don't want to wake you in the night! If you're outside the UK [please click here for important time zone info.](#)

mappiness maps happiness across space in the UK

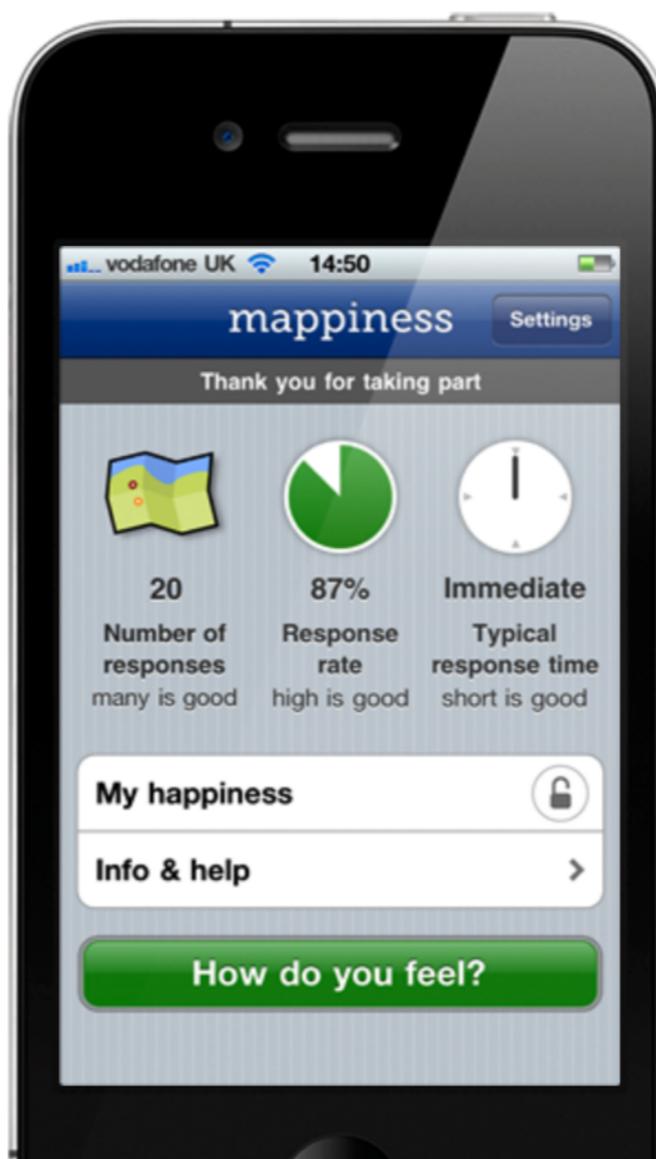
mappiness is a free app for your iPhone

It's part of a research project at the London School of Economics

We'd love to have you on board!

how does it work?

- You [get mappiness from the App Store](#), open it, and sign up
- We beep you once (or more) a day to ask how you're feeling, and a few basic things to control for: who you're with, where you are, what you're doing (if you're outdoors, you can also take a photo)
- The data gets sent back — anonymously and securely — to our data store, along with your approximate location from the phone's GPS

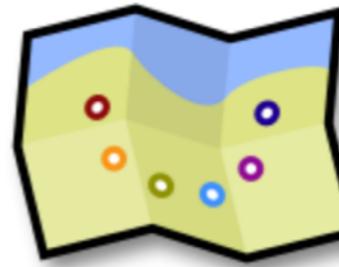


what's in it for you?

- Interesting information about your own happiness, which you can download or see charted inside the app — including when, where and with whom you're happiest
- The warm glow of helping increase the sum of human knowledge

what's in it for us?

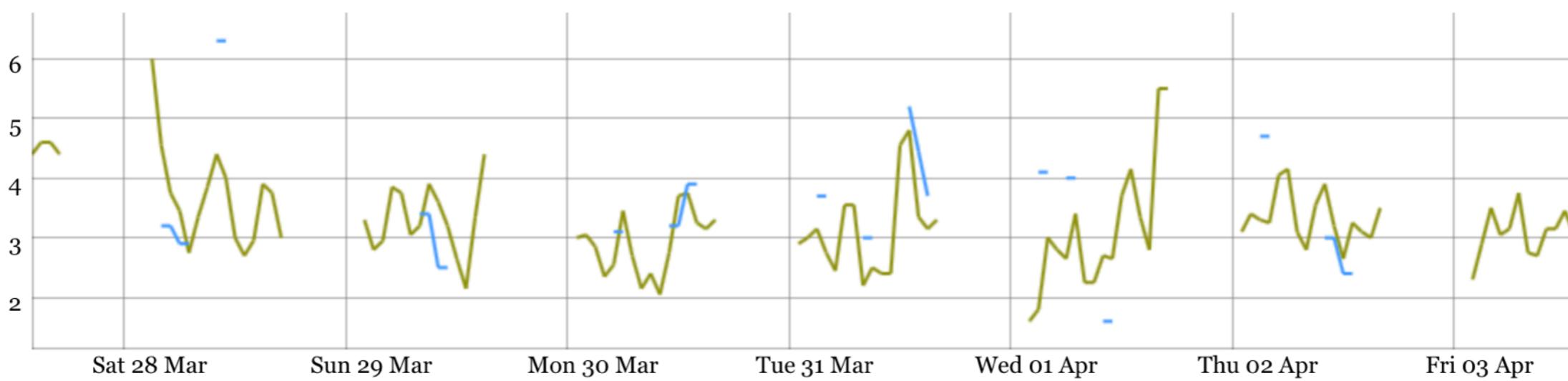
- We're particularly interested in how people's happiness is affected by their local environment — air pollution, noise, green spaces, and so on — which the data from mappiness will be absolutely great for investigating
- We'll be publishing the results in academic journals and elsewhere — starting with this [paper in Global Environmental Change](#).



when are we happy?

The **hedonimeters** on the right display mappiness users' happiness in real-time, compared against the all-time average.

Below, happiness levels are charted hour by hour over the past week.



Click and drag to zoom. Double-click to zoom back out.

- **VALUE**

All that available data will create a lot of value for organisations, societies and consumers. Big data means big business and every industry will reap the benefits from big data. McKinsey states that potential annual value of big data to the US Health Care is \$ 300 billion, more than double the total annual health care spending of Spain. They also mention that big data has a potential annual value of € 250 billion to the Europe's public sector administration. Even more, in their well-regarded report from 2011, they state that the potential annual consumer surplus from using personal location data globally can be up to \$ 600 billion in 2020. That is a lot of value.

Of course, data in itself is not valuable at all. The value is in the analyses done on that data and how the data is turned into information and eventually turning it into knowledge. The value is in how organizations will use that data and turn their organization into an information-centric company that relies on insights derived from data analyses for their decision-making.

TOPIC

BIG DATA - SEVEN V'S

Why The 3V's Are Not Sufficient To Describe Big Data, Mark van Rijmenam, 2013

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY

● LUX

The primary sensor data type for ambient light sensors is illuminance in lux (lumens per square meter). The principles outlined in this topic are based on taking lux values as input and reacting to that data in a program. Lux readings are directly proportional to the energy per square meter that is absorbed per second. Human perception of light levels is not so straightforward. Human perception of light is complicated because our eyes are constantly adjusting and other biological processes are affecting our perception. However, we can think of this perception from a simplified perspective by creating several ranges of interest with known upper and lower thresholds.

The following example data set represents rough thresholds for common lighting conditions, and the corresponding lighting step. Here, each lighting step represents a change in lighting environment.

TOPIC

SENSOR DATA - LIGHT

<https://msdn.microsoft.com/en-us/library/windows/desktop/dd319008%28v=vs.85%29.aspx>

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY

Lighting condition	From (lux)	To (lux)	Mean value (lux)	Lighting step
Pitch Black	0	10	5	1
Very Dark	11	50	30	2
Dark Indoors	51	200	125	3
Dim Indoors	201	400	300	4
Normal Indoors	401	1000	700	5
Bright Indoors	1001	5000	3000	6
Dim Outdoors	5001	10,000	7500	7
Cloudy Outdoors	10,001	30,000	20,000	8
Direct Sunlight	30,001	100,000	65,000	9

TOPIC

SENSOR DATA - LIGHT

<https://msdn.microsoft.com/en-us/library/windows/desktop/dd319008%28v=vs.85%29.aspx>

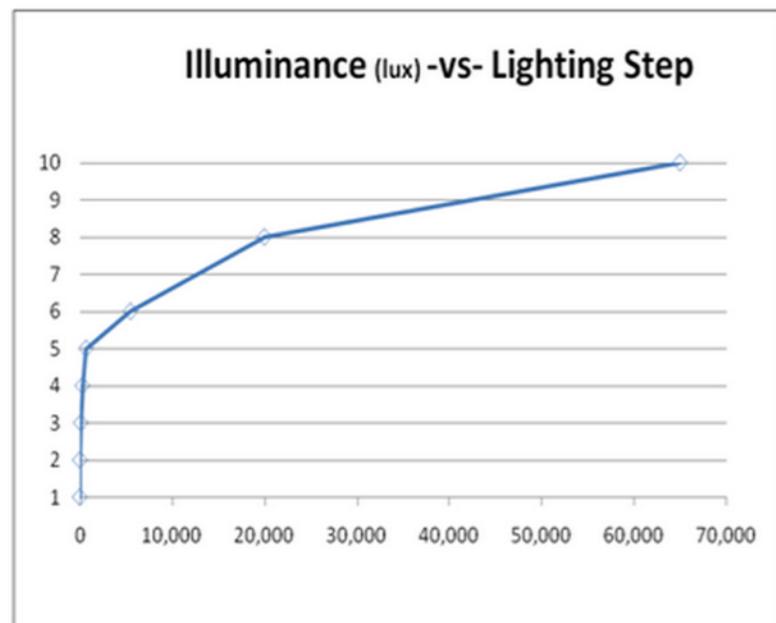
DATE

4-6-15

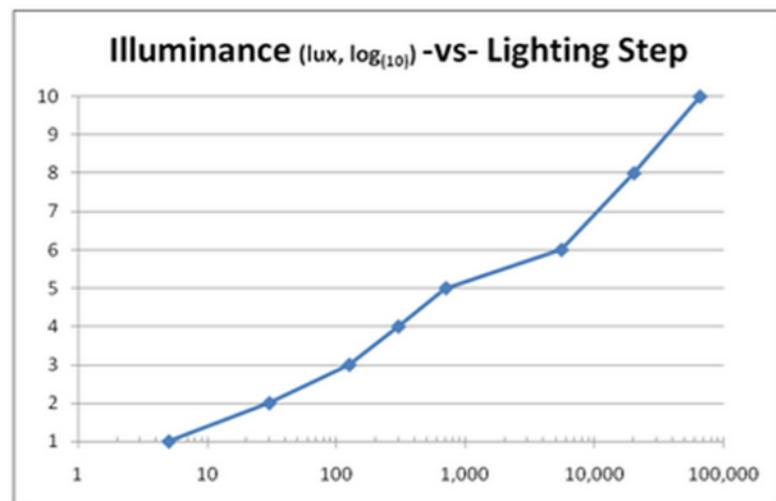
LECTURE

#3 - DATA FOR THE SMART CITY

If we visualize this data by using the mean values from this table, we see that the lux-to-lighting-step relationship is not linear, as shown in the following graph.



However, if we view this data by using a logarithmic scale on the x-axis, we can see that a roughly linear relationship emerges.



An Example Transform

Based on the sample data set for ambient light sensors previously provided, you could arrive at the following equation to map lux values to human perception. In this example, the expected values range from 0 lux to 1,000,000 lux.

$$Light_{Normalized} = \frac{\log_{10}(x)}{5.0}$$

This equation results in values that vary in a roughly linear fashion between 0.0 and 1.0. This result indicates how human-perceived lighting changed based on the example data set that was shown previously.

TOPIC

SENSOR DATA - LIGHT

<https://msdn.microsoft.com/en-us/library/windows/desktop/dd319008%28v=vs.85%29.aspx>

DATE

4-6-15

LECTURE

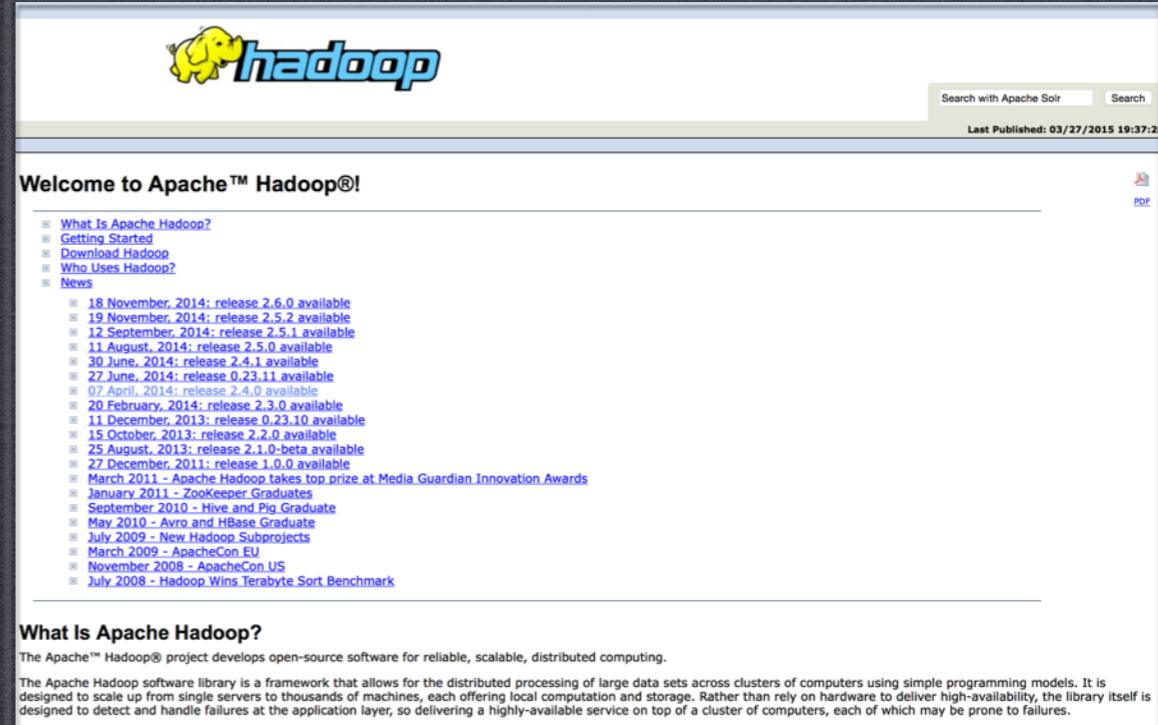
#3 - DATA FOR THE SMART CITY

● APACHE HADOOP

Wikipedia definition

Apache Hadoop is an open-source software framework written in Java for distributed storage and distributed processing of very large data sets on computer clusters built from commodity hardware. All the modules in Hadoop are designed with a fundamental assumption that hardware failures (of individual machines, or racks of machines) are commonplace and thus should be automatically handled in software by the framework.

- **STORAGE**
Hadoop Distributed File System (HDFS)
- **PROCESSING**
Map Reduce



The screenshot shows the official Apache Hadoop website. At the top, there's a navigation bar with links for "Search with Apache Solr" and "Search". Below the search bar, it says "Last Published: 03/27/2015 19:37:28". The main header features the Hadoop logo (a yellow elephant icon next to the word "hadoop"). The page title is "Welcome to Apache™ Hadoop®!". Below the title is a sidebar with links to "What Is Apache Hadoop?", "Getting Started", "Download Hadoop", "Who Uses Hadoop?", and "News". The "News" section lists various release dates and versions, such as "18 November, 2014: release 2.6.0 available" and "19 November, 2014: release 2.5.2 available". The main content area contains a section titled "What Is Apache Hadoop?" with a brief description of the project's purpose and its design principles.

TOPIC

HOW DO WE STORE AND PROCESS BIG DATA?

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY

MapReduce explained in 41 words

Fri 05 August 2011 [hadoop / mapreduce](#)

 Follow @stucchio

 Tweet 0

 Like  Share Be the first of your friends to like this.

 +12 Recommend this on Google

Goal: count the number of books in the library.

Map: You count up shelf #1, I count up shelf #2.

(The more people we get, the faster this part goes.)

Reduce: We all get together and add up our individual counts.

TOPIC

MAP REDUCE

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY

▲ [yummyfajitas](#) 1339 days ago

The explanation I gave to my mom:

We want to count all the books in the library. You count up shelf #1, I count up shelf #2. That's map. The more people we get, the faster it goes.

Now we get together and add our individual counts. That's reduce.

[edit: created a blog post containing this explanation. Maybe it will be useful after this HN conversation vanishes. <http://crazybear.posterous.com/mapreduce-explained-in-41-wor...>]

▲ [wisty](#) 1339 days ago

That's the first iteration of the explanation.

Next iteration: But what if you want to count up non-fiction and fiction? There's just a slightly different "reduce" step, because people have to add up both of them. You could also break it down by subject, or author, or pretty much anything else.

Next iteration: Of course, you don't have to just add. You can do anything that has mathematical properties similar to addition (my mom teaches math at high-school). You could add up the set of all the offense words found, since addition and set addition have similar mathematical properties. You couldn't take the average number of pages, because some people will find more books than others, and you wouldn't get a fair average. But you might be able to keep track of a couple of attributes that can be summed up, and use that to calculate the average after the job is done (for example, total number of pages, and total number of books).

▲ [dcosson](#) 1339 days ago

Now if only everyone's mom understood the concept of a linear set operation... :)

A good example that would also help someone appreciate the power of map/reduce might be something like "how many times does each letter of the alphabet appear in the library?" For the map step I take shelf 1 and you take shelf 2, but now for each book we'll have to write down something like "a:300,b:50,c:90,..." and so on. But now there are more things to "reduce" than there are books (since we now have a value for the number of occurrences of each letter in the alphabet for every book). Well, there's nothing stopping us from splitting up the reduce step as well - we could split it up by shelf again (where you add the totals for each letter on your shelf and I'll do mine, then we'll do one final reduce of at the end), or you can add up letters a-m and I'll do n-z.

▲ [th0ma5](#) 1339 days ago

I remember a Stanford lecture video that said "well, if we called it Map, Sort, Group-reduce then no-one would use it :P" but that's a minor naming issue to describe just the subtly you are exactly describing, which is to just not be entirely random in the specifics of how the MapReduce is executed.

▲ [Igarron](#) 1339 days ago

Why wouldn't you be able to get the average? Return page count + total pages for each shelf and then take the ratio of "reduced" sums.

▲ [michael_dorfman](#) 1339 days ago

Re-read the comment. It discusses exactly what you are proposing.

TOPIC

MAP REDUCE

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY



Learn MapReduce with Playing Cards



Jesse Anderson

 Subscribe 269

33,306

 262  8

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Autoplay



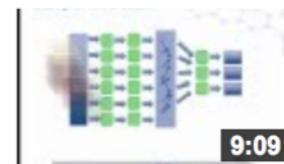
Introduction to NoSQL by Martin Fowler
by GOTO Conferences
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by MapRAcademy
53,473 views



Introduction To Apache Cassandra
by CernerEng
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Intro To MapReduce
by MapRAcademy
109,053 views



MapReduce Flow Chart
by Durga Software Solutions
36,484 views



Hadoop Tutorial 1 - What is Hadoop?
by handsonerp
Recommended for you

TOPIC

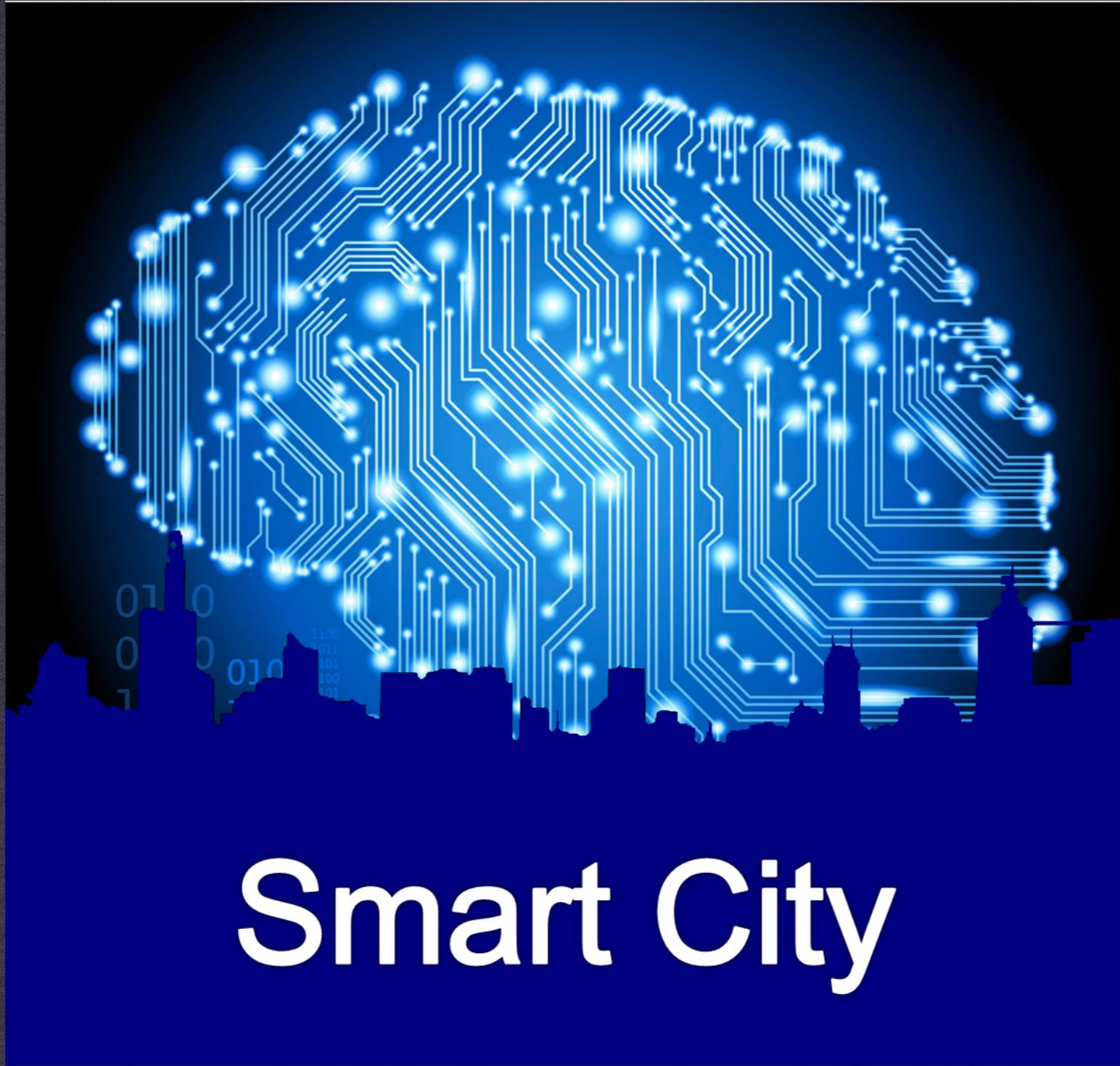
MAP REDUCE

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY



Smart City

TOPIC

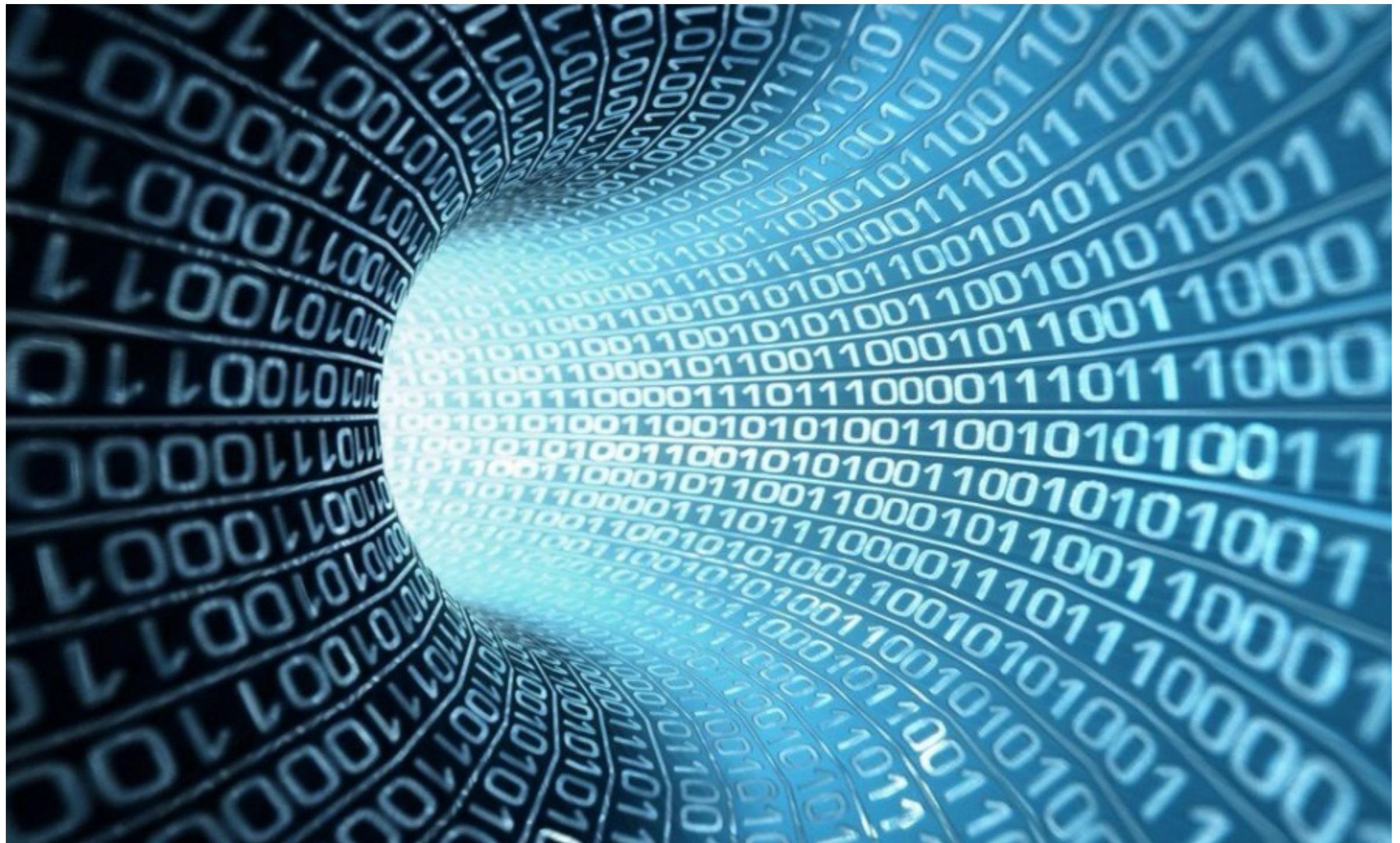
BIG DATA - IMAGES AND INFOGRAPHICS

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY



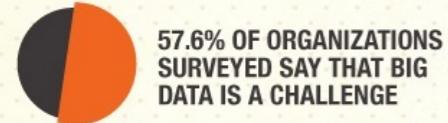




BIG DATA

Big Data is data that is too large, complex and dynamic for any conventional data tools to capture, store, manage and analyze.

The right use of *Big Data* allows analysts to spot trends and gives niche insights that help create value and innovation much faster than conventional methods.



The “three V’s”, i.e the Volume, Variety and Velocity of the data coming in is what creates the challenge.

VOLUME



VARIETY

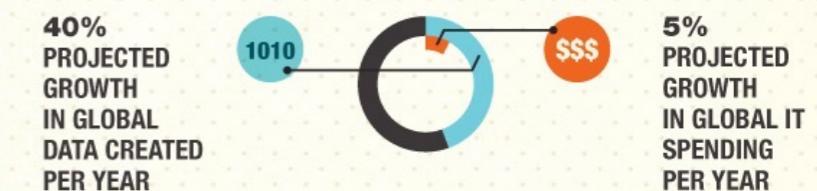
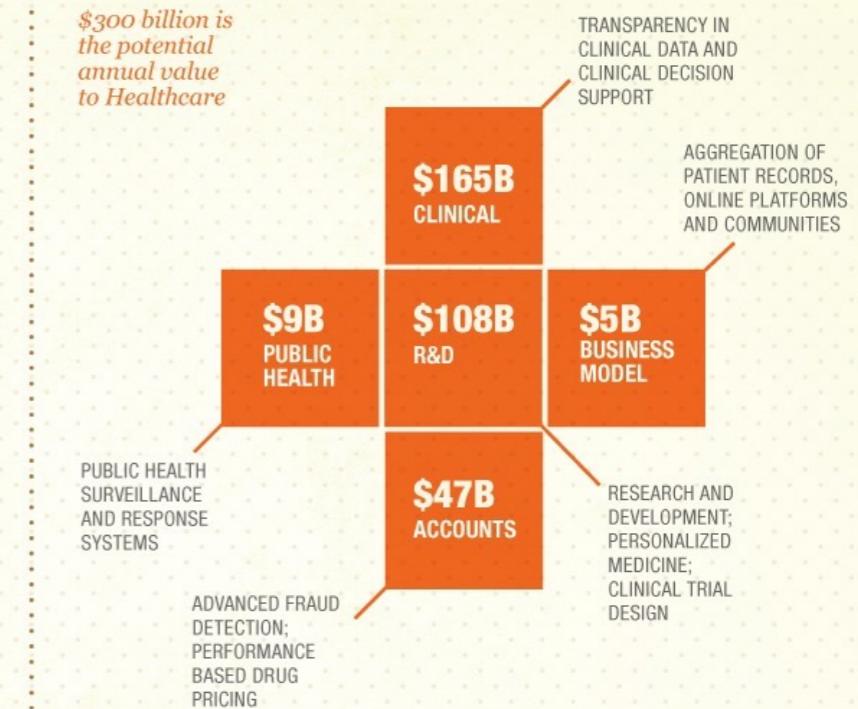


VALUE



CASE STUDY - Healthcare

\$300 billion is the potential annual value to Healthcare



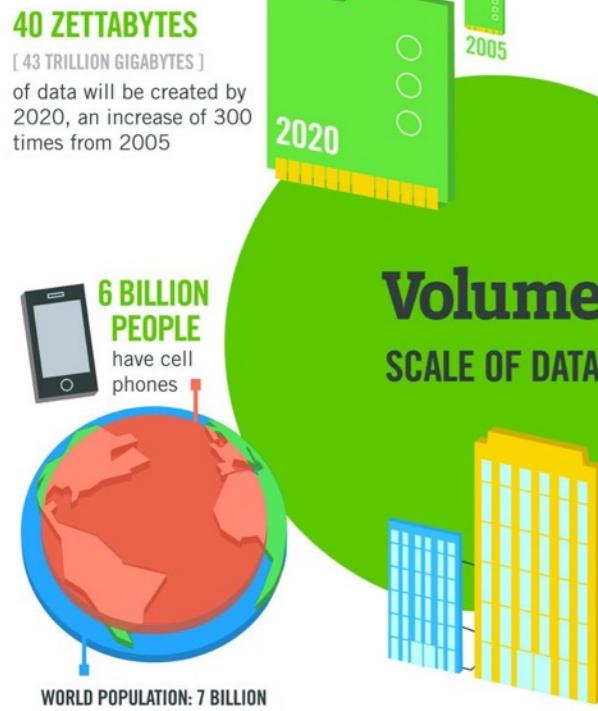
The estimated size of the digital universe in 2011 was 1.8 zettabytes. It is predicted that between 2009 and 2020, this will grow 44 fold to 35 zettabytes per year. A well defined data management strategy is essential to successfully utilize Big Data.

Sources - ① Reaping the Rewards of Big Data - Wipro Report ② Big Data: The Next Frontier for Innovation, Competition and Productivity - McKinsey Global Institute Report ③ comScore, Radicati Group ④ Measuring the Business Impacts of Effective Data - study by University of Texas, Austin ⑤ US Department of Labour.

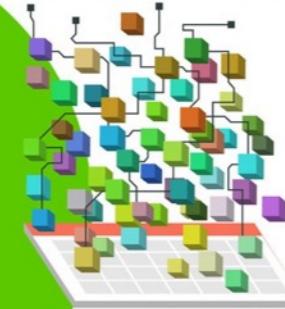
DO BUSINESS BETTER

NYSE:WIT | OVER 130,000 EMPLOYEES | 54 COUNTRIES | CONSULTING | SYSTEM INTEGRATION | OUTSOURCING





It's estimated that
2.5 QUINTILLION BYTES
[2.3 TRILLION GIGABYTES]
of data are created each day



The FOUR V's of Big Data

From traffic patterns and music downloads to web history and medical records, data is recorded, stored, and analyzed to enable the technology and services that the world relies on every day. But what exactly is big data, and how can these massive amounts of data be used?

As a leader in the sector, IBM data scientists break big data into four dimensions: **Volume**, **Velocity**, **Variety** and **Veracity**

Depending on the industry and organization, big data encompasses information from multiple internal and external sources such as transactions, social media, enterprise content, sensors and mobile devices. Companies can leverage data to adapt their products and services to better meet customer needs, optimize operations and infrastructure, and find new sources of revenue.

By 2015
4.4 MILLION IT JOBS
will be created globally to support big data, with 1.9 million in the United States



Velocity ANALYSIS OF STREAMING DATA

The New York Stock Exchange captures
1 TB OF TRADE INFORMATION
during each trading session



By 2016, it is projected there will be
18.9 BILLION NETWORK CONNECTIONS
– almost 2.5 connections per person on earth



Modern cars have close to
100 SENSORS
that monitor items such as fuel level and tire pressure



As of 2011, the global size of data in healthcare was estimated to be

150 EXABYTES
[161 BILLION GIGABYTES]



Variety DIFFERENT FORMS OF DATA

30 BILLION PIECES OF CONTENT

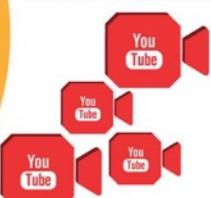
are shared on Facebook every month



By 2014, it's anticipated there will be
420 MILLION WEARABLE, WIRELESS HEALTH MONITORS

4 BILLION+ HOURS OF VIDEO

are watched on YouTube each month



400 MILLION TWEETS
are sent per day by about 200 million monthly active users

1 IN 3 BUSINESS LEADERS

don't trust the information they use to make decisions



27% OF RESPONDENTS

in one survey were unsure of how much of their data was inaccurate

Veracity UNCERTAINTY OF DATA

Poor data quality costs the US economy around
\$3.1 TRILLION A YEAR



BIG DATA & ANALYTICS

Core to Nex-Gen Manufacturing

With more Asia Pacific manufacturers looking at making better and faster business decisions based on evidence, Big Data technologies can help them analyze larger volumes of data from a variety of sources and deliver the analysis at greater velocity. IDC observes that regional manufacturers are taking a cautious approach and the use of Analytics tools today is basic. Here is the 2013 landscape in Asia Pacific.

Asia Pacific manufacturers

3 key highlights

4 of the top 10 automotive producing countries are from Asia Pacific



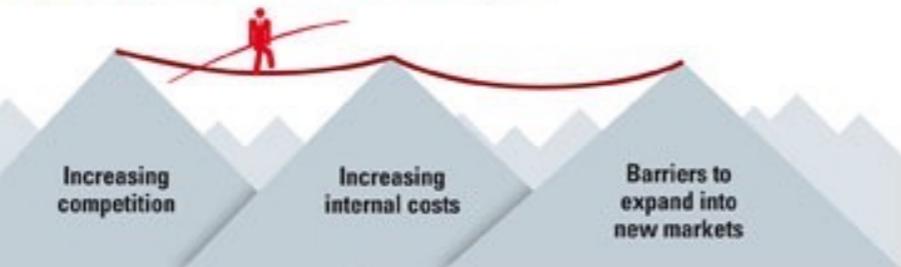
+370%

China's share of exports has nearly quadrupled over the past 15 years, rising from 3% in 1995 to 11% in 2012

\$ x2

The average labor cost in China has nearly doubled in the past five years

Top 3 regional business challenges



What Big Data means

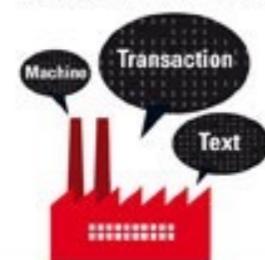
Awareness

2/5

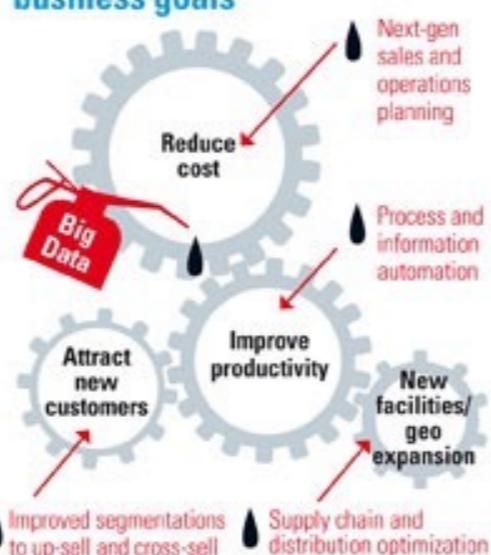
are still not aware of Big Data



Data that manufacturers are analyzing: Transactions (47%), text (45%) and machine (32%)



Big Data helps meet business goals



Common use cases

Today

37%
29%

The % of Asia Pacific manufacturers using Big Data and Analytics technologies to improve production quality management

Enabling technologies

- Production reporting tools (46%)
- Supply Chain analytics (35%)
- Hadoop (8%)

Industry leader use cases



Condition-based monitoring

Reduce downtime from unscheduled maintenance



Supply chain resilience

Decrease production delays due to global disruptions



Shop floor automation

Increase production velocity, productivity and consistency

Moving forward

In the next 2 years



Asset Oriented Value Chain (AOVC) and Engineering Oriented Value Chain (EOVC)
manufacturers will use Big Data and Analytics technologies to gain Real time Asset Management



Brand Oriented Value Chain (BOVC)
manufacturers will use Big Data and Analytics technologies to improve real time sales and operations planning



Technology Oriented Value Chain (TOVC)
manufacturers will use Big Data and Analytics technologies to improve Supply Chain Management

Enabling technologies



M2M, Telematics



Predictive analytics



Data warehouse, Analytics appliances

Source:

- IDC Big Data Pulse Survey
- IDC Manufacturing Insights Industry Survey
- IDC Big Data and Analytics CIS
- IDC Manufacturing Insights
- IMF
- China Ministry of Commerce

Research team:

- Chris Holmes
- Daniel-Zoe Jimenez
- William Lee
- Craig Stiles

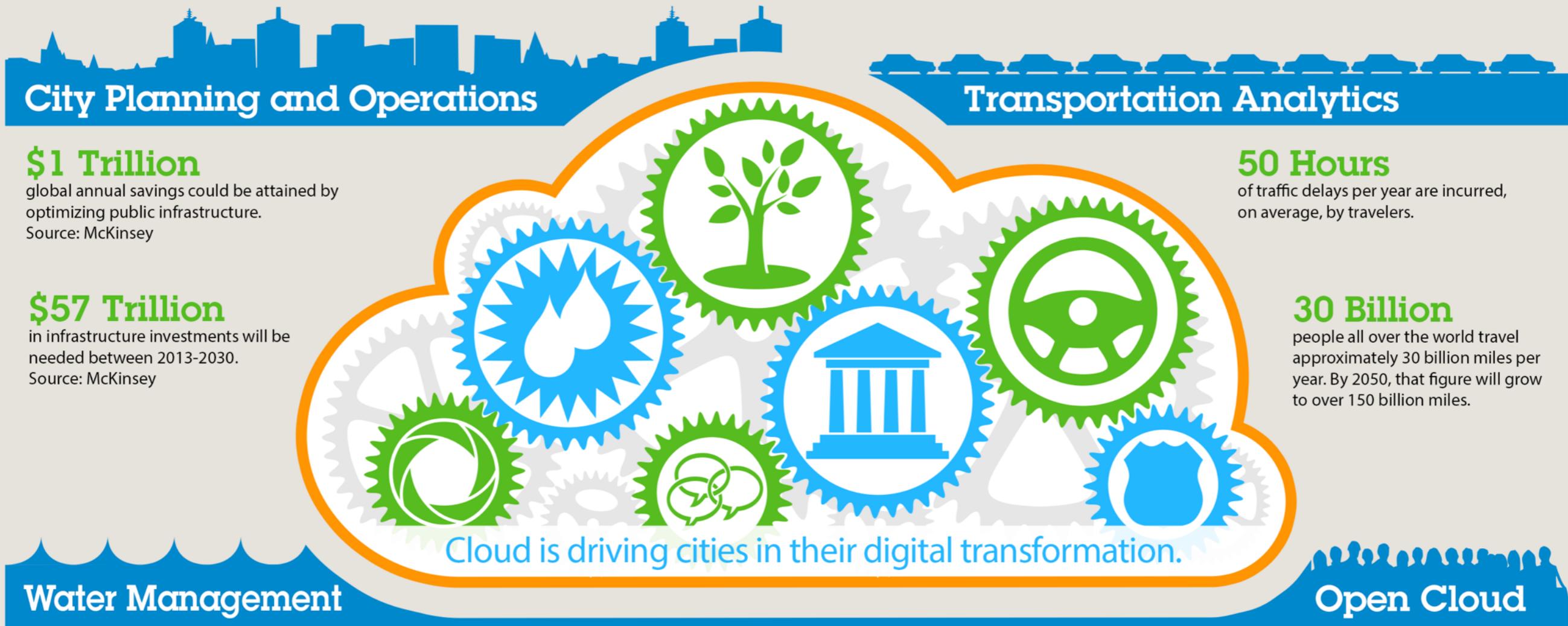
 Art direction:

- ST Lang

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Smarter Cities: Turning Big Data Into Insight

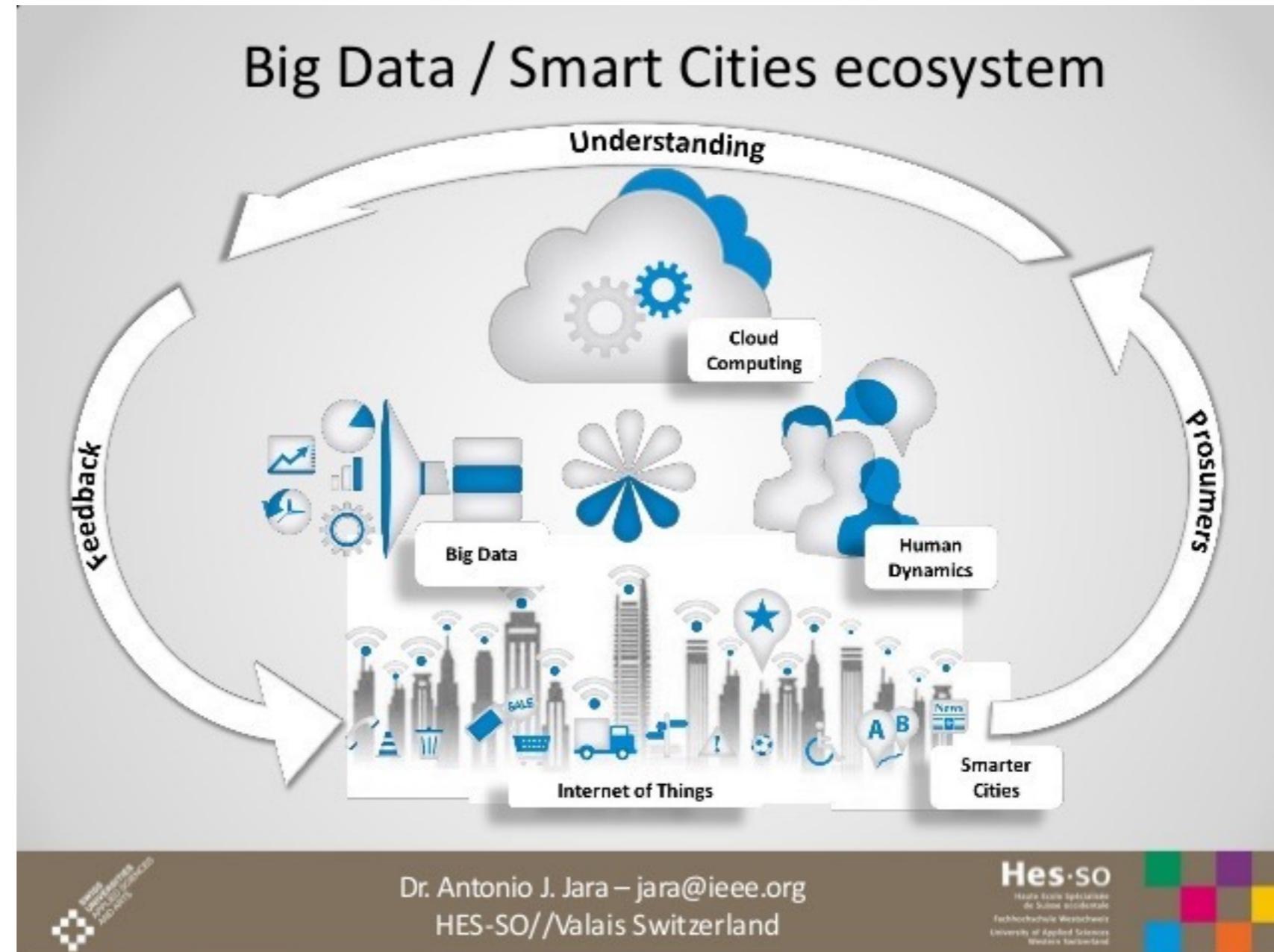
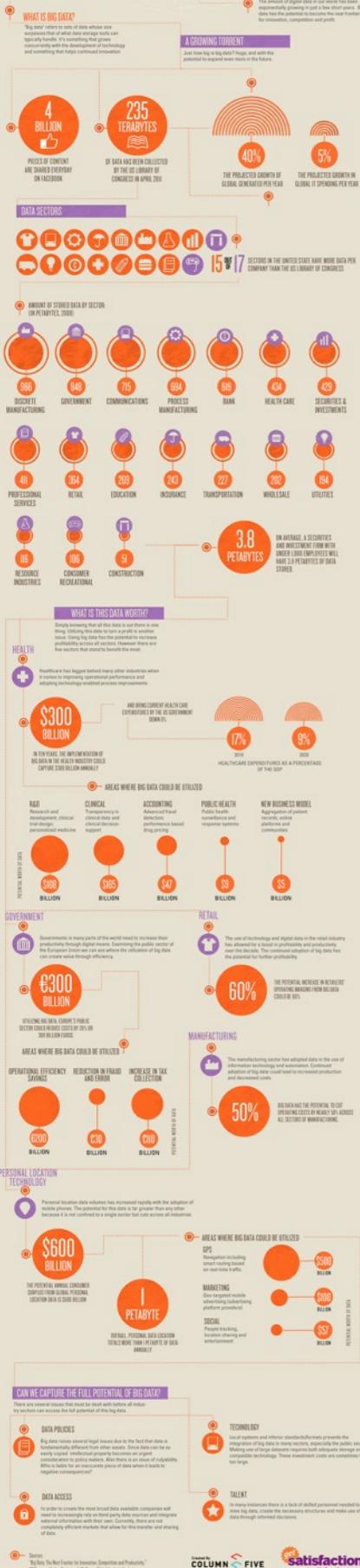


IBM Intelligent Operations software is designed with cities, for cities, to provide the tools to monitor, visualize and analyze vital city services such as water and wastewater systems, transportation, infrastructure planning, permit management and emergency response.



EXPLODING DATA

THE POTENTIAL OF BIG DATA



- WIRELESS SENSOR NETWORKS – AN INTRODUCTION.GEOJOURNAL INTECH
- SMARTSANTANDER – THE CITY OF SANTANDER'S SENSOR NETWORK FACILITY
Smart Santander

TOPIC

ASSIGNED READINGS FOR NEXT LECTURE

DATE

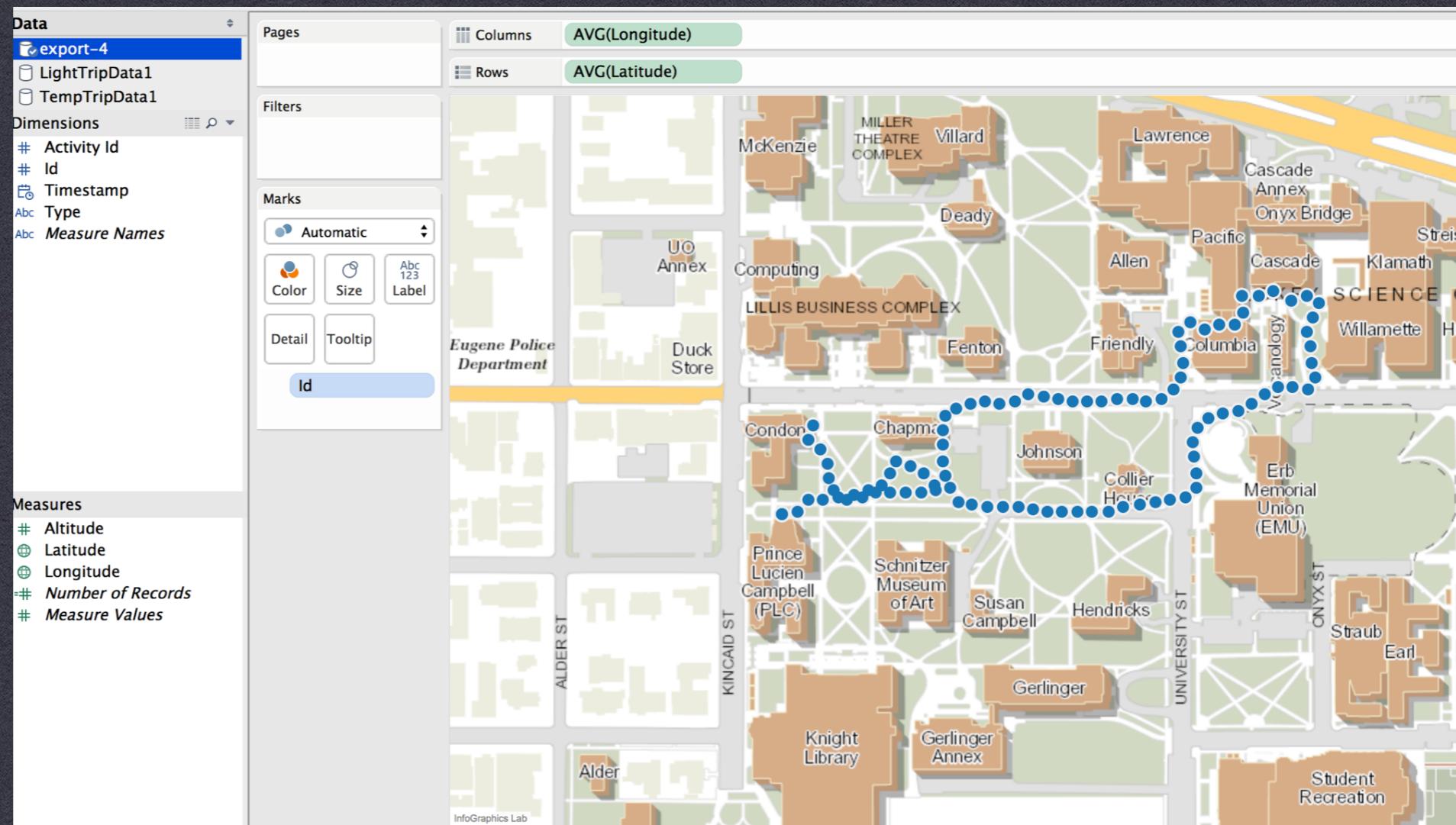
4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY

MAPPING YOUR LOCATION SENSOR DATA

Run Keeper



TOPIC

THIS WEEK'S LAB

DATE

4-6-15

LECTURE

#3 - DATA FOR THE SMART CITY