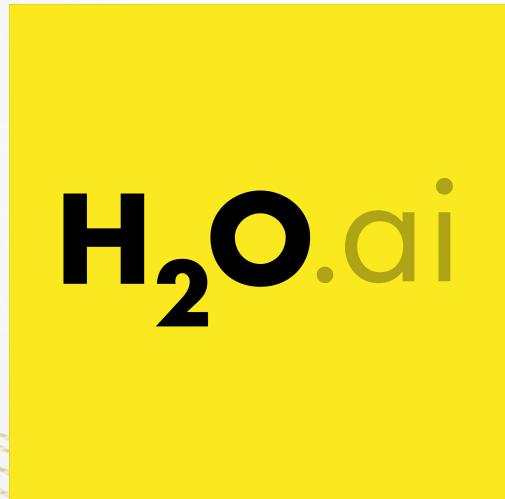


Applied Machine Learning for IoT



Hank Roark

13-April-2016

Data Science Pop-up Austin

WHO AM I

Lead, Customer Data Science @ H2O.ai

John Deere: Research, Software Product Development, High Tech Ventures
Lots of time dealing with data off of machines, equipment, satellites, weather,
radar, hand sampled, and on.

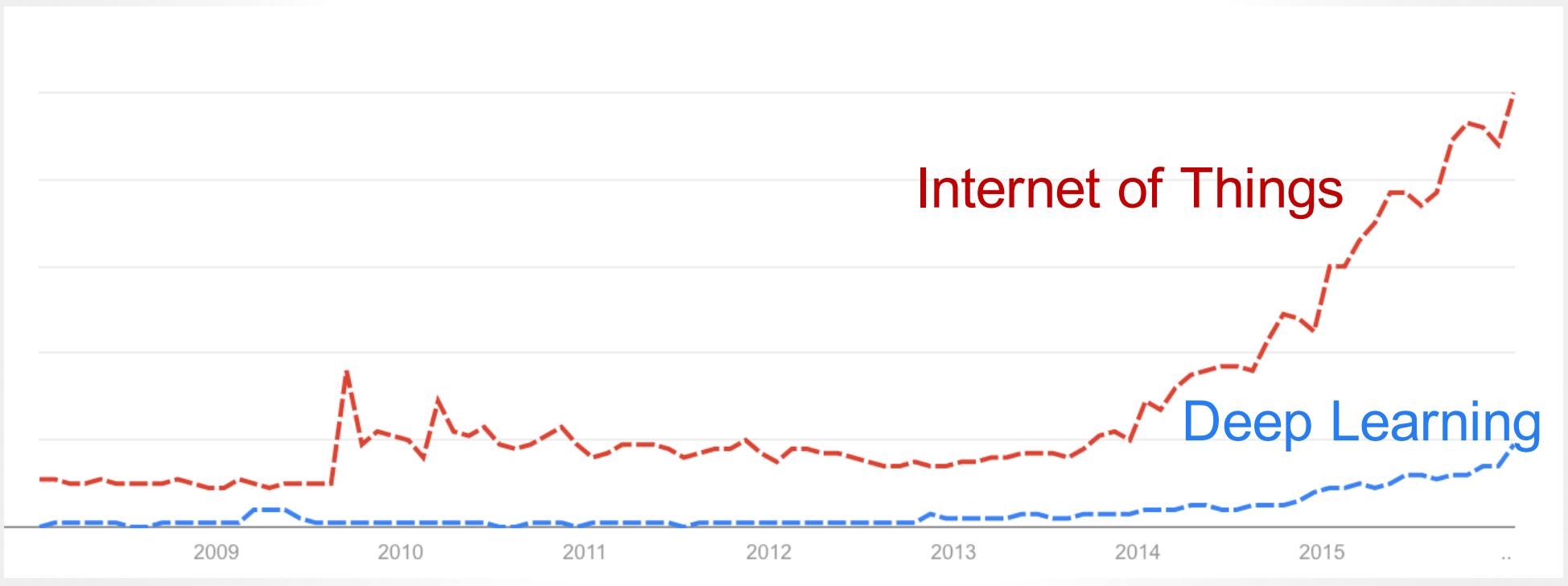
Geospatial, temporal / time series data almost all from sensors.

Systems Design and Management: MIT
Physics: Georgia Tech

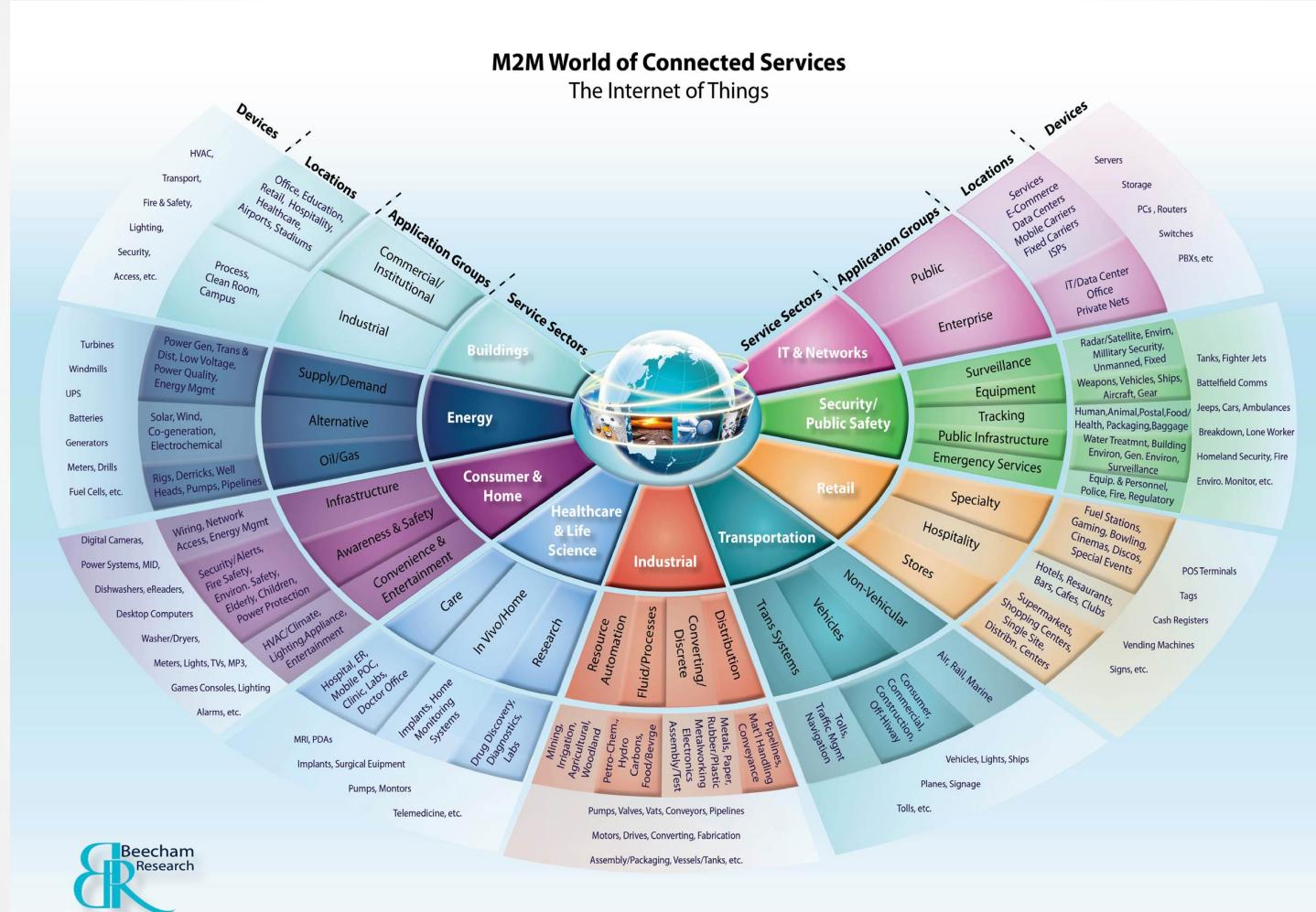
hank@h2o.ai
@hankroark
<https://www.linkedin.com/in/hankroark>

WHY IOT?

Because it's trending (so it must be promising)!

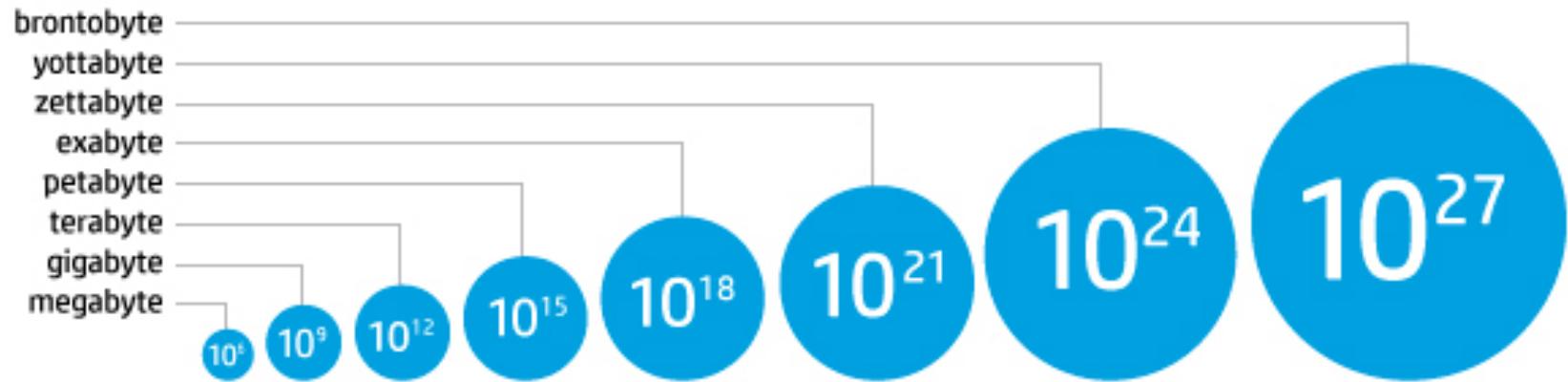


IF YOU ARE INTO DATA, THE IOT HAS IT



LOTS OF DATA

Information & the Internet of Things

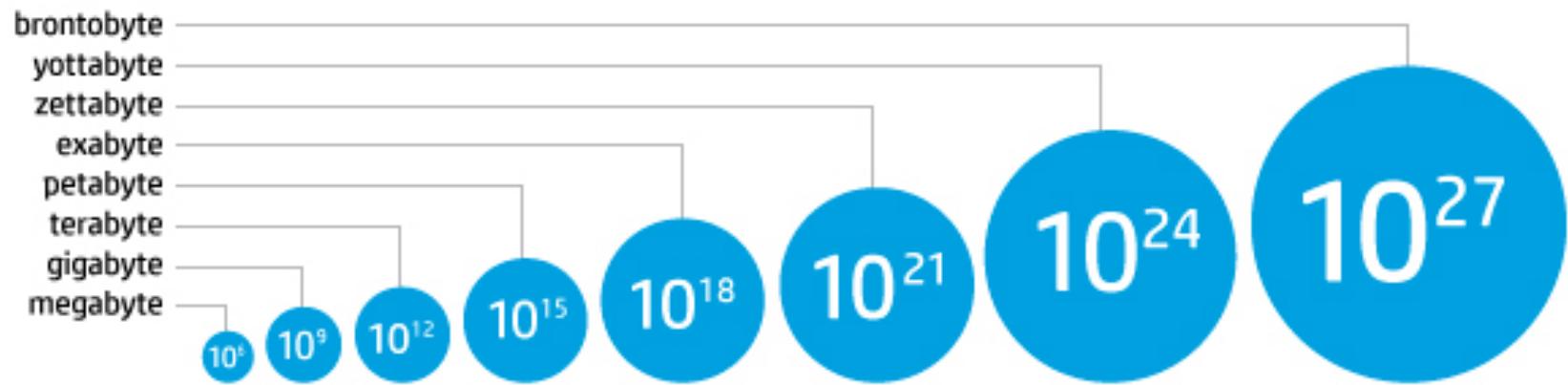


Today, data scientists max out at yottabytes, but soon, brontobytes will measure the volume of sensor data generated by the Internet of Things.

Source: HP

LOTS OF DATA

Information & the Internet of Things



Today, data scientists max out at yottabytes, but soon, brontobytes will measure the volume of sensor data generated by the Internet of Things.

Source: HP

GET READY FOR BRONTOBYTES!!

WOW, HOW BIG IS A BRONTOBYTE?

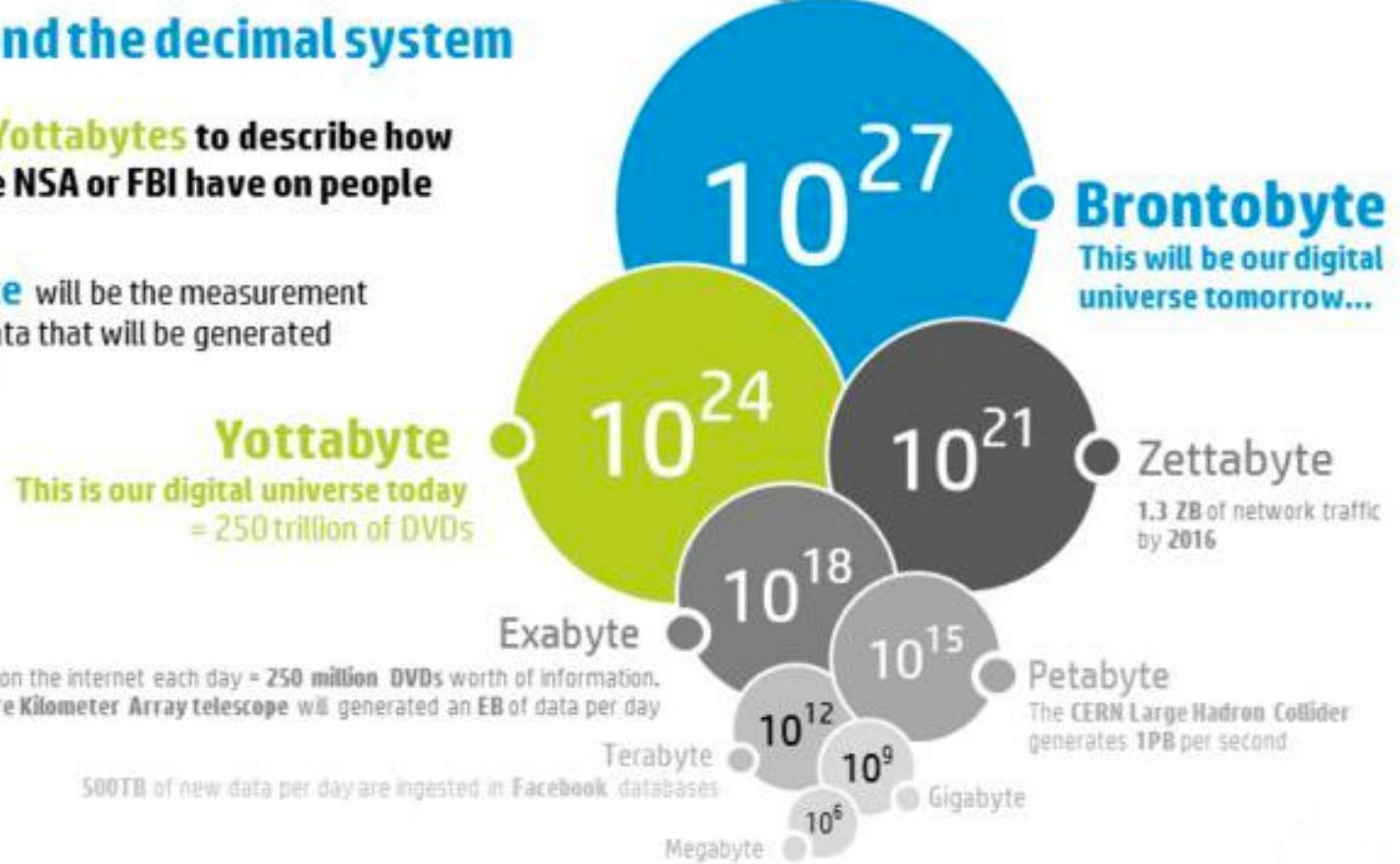
Information from the Internet of Things: We have gone beyond the decimal system

Today data scientist uses **Yottabytes** to describe how much government data the NSA or FBI have on people altogether.

In the near future, **Brontobyte** will be the measurement to describe the type of sensor data that will be generated from the IoT (Internet of Things)

1 EB of data is created on the internet each day = 250 million DVDs worth of information.
The proposed Square Kilometer Array telescope will generate an EB of data per day

500TB of new data per day are ingested in Facebook databases



IOT ARCHITECTURE IS SETUP FOR REINFORCING, EXPONENTIAL GROWTH

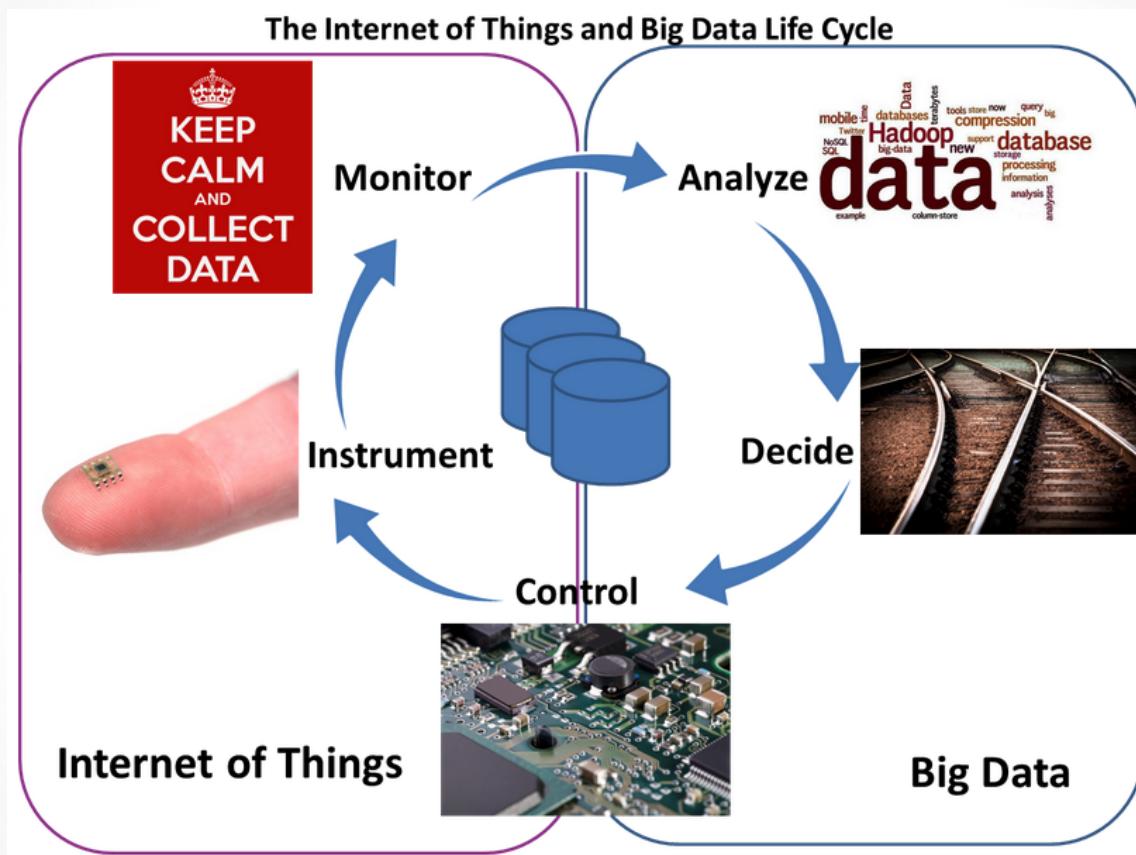
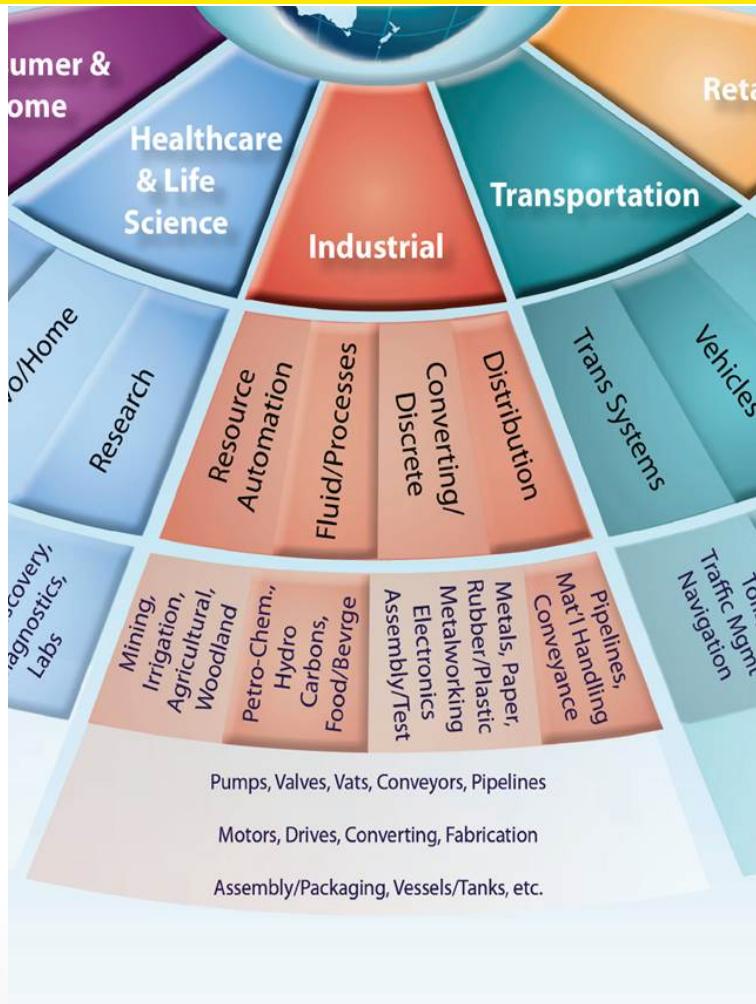


Image courtesy <http://www.telecom-cloud.net/wp-content/uploads/2015/05/Screen-Shot-2015-05-27-at-3.51.47-PM.png>

MY PARTICULAR CURIOSITY



EXAMPLE FROM THE IOT

Domain: Prognostics and Health Management

Machine: Turbofan Jet Engines

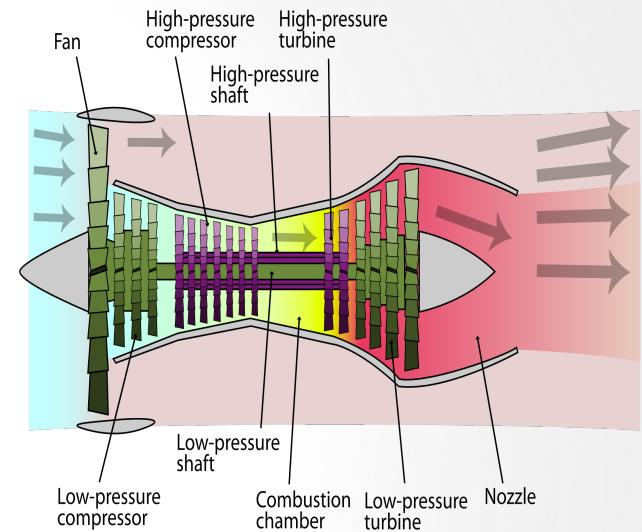
Data Set: A. Saxena and K. Goebel (2008). "Turbofan Engine Degradation Simulation Data Set", NASA Ames Prognostics Data Repository

Predict **Remaining Useful Life** from Partial Life Runs

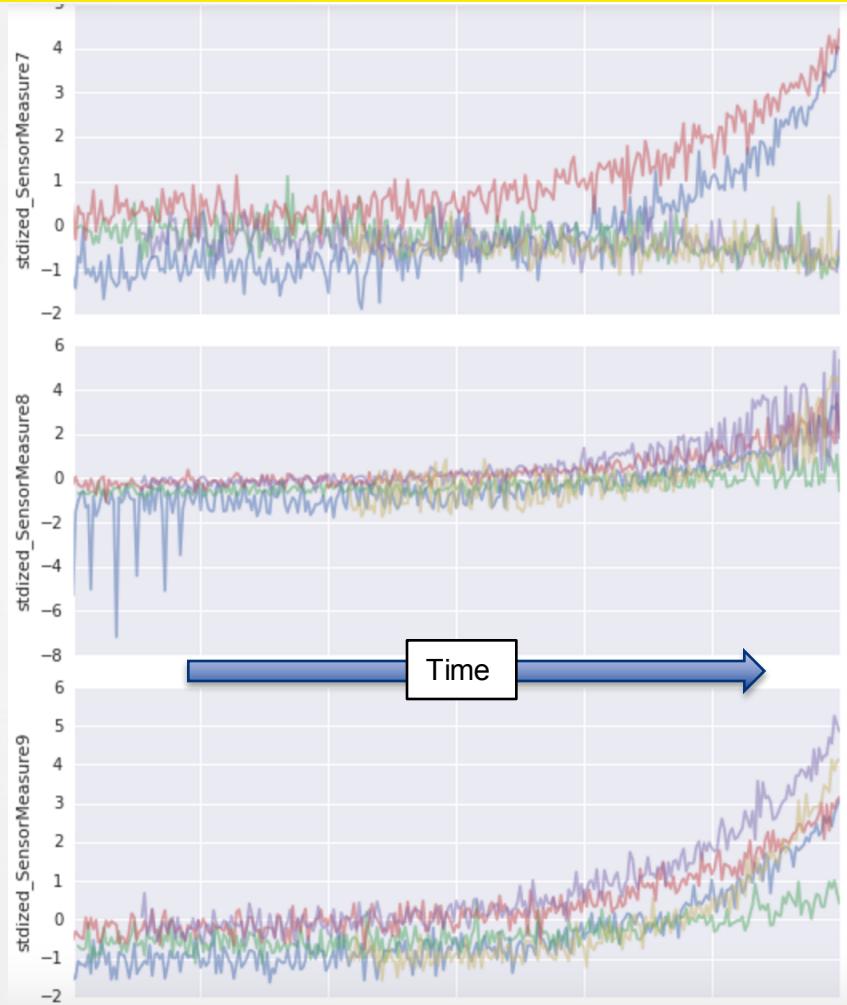
Six operating modes, two failure modes,
manufacturing variability

Training: 249 jet engines run to failure

Test: 248 jet engines



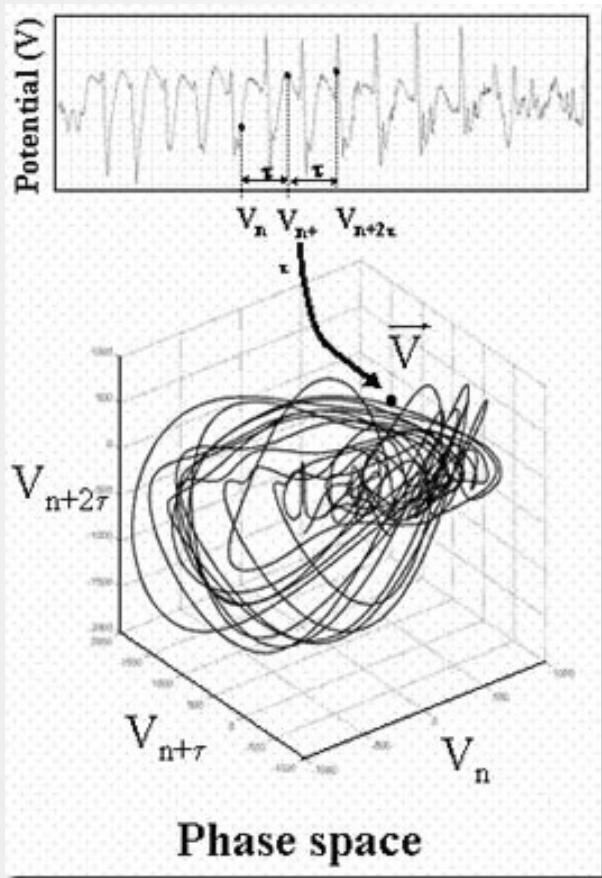
SENSORS SHOW TRENDS OVERTIME



We also know remaining useful life decreases by at least one (1) cycle each operation.

How can one take advantage of this knowledge?

INCORPORATING PRIOR STATE



One option:
Phase Space Embedding

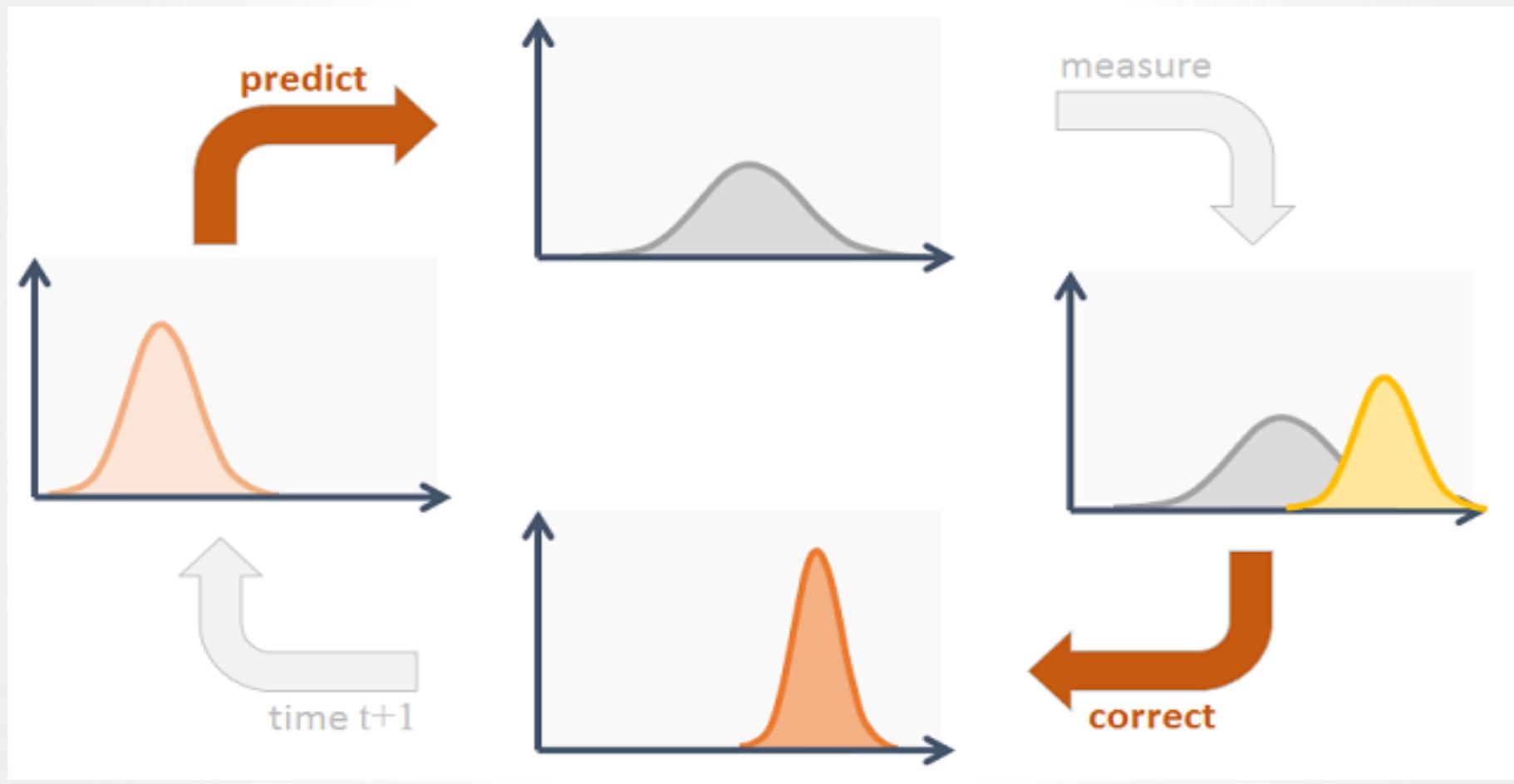
Drawbacks:
Incorporates knowledge from only
small number of prior states

Curse of dimensionality

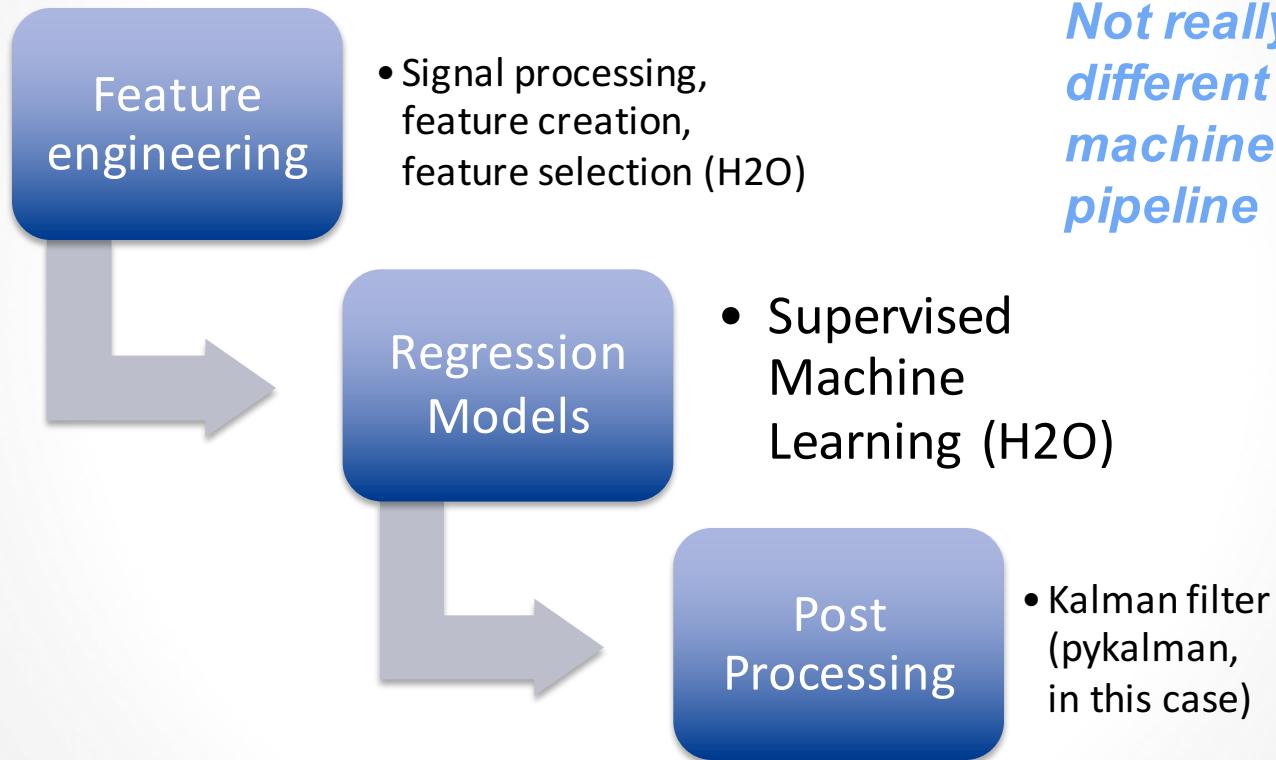
Another parameter (tau)

Disentangling the dynamic core: a research program for
a neurodynamics at the large-scale
MICHEL LE VAN QUYEN
Biol. Res. v.36 n.1 Santiago 2003
<http://dx.doi.org/10.4067/S0716-97602003000100006>

KALMAN FILTER

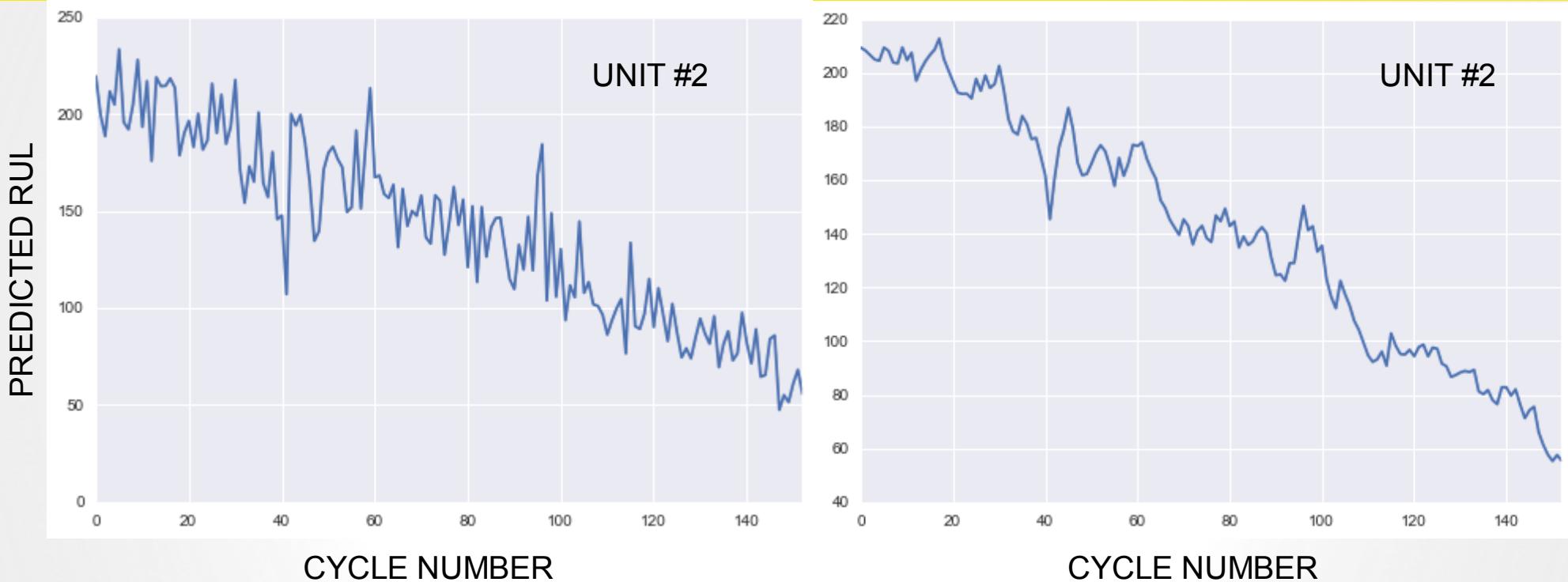


FINAL PIPELINE



Not really that much different than typical machine learning pipeline

BEFORE AND AFTER POST-PROCESSING



This presentation, data, and associated Jupyter notebooks
(look in 2016_04_13_AppliedMachineLearningForIoT):
<https://github.com/h2oai/h2o-meetups/>

FLEET TELEMATICS: *PREDICTIVE MAINTENANCE*

“Annual Savings are **\$7M**”

– Anonymized, Member Technical Staff

Leading Mobile Telecom Operator

PROBLEM

- Fleet telematics—analyze maintenance records and vehicle performance to make predictions on when to do preventive maintenance
- Couldn’t scale by sampling data
- Took days to create models

WHY H2O

- H2O support for customer’s Kerberos authentication mechanism for Hadoop
- Support for MapReduce, YARN, R, Python and Spark in Hadoop
- In-memory, distributed architecture
- Rapid deployment to production with POJO
- Quick prototyping with H2O Flow

IMPACT

- When you look at the cost of towing a stranded vehicle, technician loss of productivity, and the customer lifetime value, the annual savings is **\$7M.**”
- Anonymized, Lead Member Technical Staff



Telecommunications

TAKE-AWAYS

- Know your “physics”
 - “Physics” is will be like a fish finder in this sea of Brontobytes
 - Linear dynamic systems, Queues, Signals, etc.
- Big data *and* Big models *and* Small models
- Keeping and eye on promising new methods (e.g., Convolution Neural Nets and LSTM-RNN)

H2O RESOURCES

- Download and go: <http://www.h2o.ai/download>
- Documentation: <http://docs.h2o.ai/>
- Booklets, Datasheet: <http://www.h2o.ai/resources/>
- Github: <http://github.com/h2oai/>
- Training: <http://learn.h2o.ai/>

THANK YOU