



Big Data, Behavioral Change and IOT Architecture

Euro-CASE Annual Conference 2016 BIG DATA - SMARTER PRODUCTS, BETTER SOCIETIES

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> > Version 0.5

Outline

- Big Data: A Multiple Paradigm Shift Our NATF 2015 report on the disruptive nature of Big Data
- 2. Facilitate Learning New Behaviors Through Games Exponential rate of change yield adaptation stress: the value of games for learning
- Internet Of Things Architecture

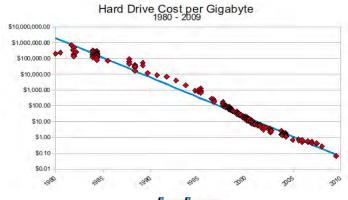
 From Event-Driven Architecture towards A Society of

 Minds

The NATF 2014 Mandate on Big Data

- Big Data is a natural evolution of software technology in an exponentially changing world
 - Wealth of data sources thanks to sensors and digital transformation
 - Spectacular decline of storage cost
 1To: 1 M\$ in 1995, less than 40\$ in 2013
 - New challenges (3V : Volume, Variety, Velocity)
 have spawn new technologies
 - Massively distributed data avoid data transfer with petabytes @
 - Massively parallel processing Flow computing, Event processing
 - Unstructured data









Should we see Big Data as an evolution or a revolution?

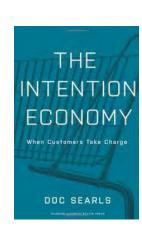
Big Data – A Revolution That Will Transform How We Live

A Revolution That Will Fransform How We Live, Work and Think

Viktor Mayer-Schönberger and Keuneth Cukier

 Intension economy and synchronized marketing Retargeting (Criteo) and Recommendation (Amazon / Netflix)

 Predictive Learning has shown its relevance Target example / de-anonymization



Big Data and connected objects
 to enrich service through memory and analysis



Continuous Feedback Loop as a Data Mining Paradigm



- Break away from causal analysis
 « The data deluge makes the scientific method obsolete »
- A systemic approach that is closer from computer sciences than from statistics
 A real-time learning loop that is evaluated on its operational results
- Simple methods that are tuned on extremely large data sets over perform sophisticated algorithms tuned on regular data sets

Data Ecosystems and Data Privacy



 Data taxonomy: anonymized / aggregates, profiles, usage and scores (computed)
 Big Data undercuts anonymization



- Debate around data processing finality?
 - External audit (CNIL), non-transferable
 - Need for OPT-IN & shared domain finality constraint (basis for consent)



- Who owns the data?
 - Profile : customer; usage : co-ownership,
 - Computed scores belong to companies (within a domain finality)

A New Way to Write Code

Learn to Code
Teach yourself essential coding skills

Perfect for Raspherry Pi

« Data is the new code »
 Algorithms may be reconstructed from data

- Massively parallel, distributed, open-source Thomas Hoffman:
 - « Big Data is getting at the core of computer science »



 Algorithms are synthesized through Machine Learning + sub-linear techniques



 Tools and techniques from Big Data enable the reinvention of traditional IT business systems



Part II

- 1. Big Data : A Multiple Paradigm Shift
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 BigData
- Exponential rate of change yield adaptation stress: the value of games for learning
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The Promises of a World of Sensors and Data Science

- An era of abundant scientific information about the impact of behavior on health and well being
 - Example : Brain oxygenation on IQ
 - Big data techniques allows for more detailed & in-depth studies
 - Impressive rate of expansion of studies & scientific literature
- IOT & e-Health: exponential growth of sensors adds to this trend
 - Example : Withings Body Cardio scale with Pulse Wave Velocity
 - Smartphones will become part of our healthcare journeys
 - Connected wearables generates longitudinal traces (individualized)

STRENGTHEN YOUR HEART
STRENGTHEN YOUR HEART

brain

rules

- Empowering users to better control their health and wellbeing
 - More information (personalized, accessible at any time)
 - Better feedback (real time, insights, ...) through wearables



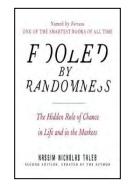
Tracking Without Sense: User-Centric Design

- Users do not want dashboards
 - They become bored quickly (cf. wearable quick and sad fate)
 - Tracking without sense (not everyone is a data scientist)
 - True in home automation (remote control and monitoring is not enough)

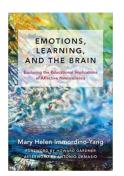


- We are moved (and we remember) stories, not data sets (N. Taleb)
- To react on an sustained time horizon, we need positive feedback loops
- Behavioral change is hard
 - Cf. temporal cycles of engagement
 - Need to leverage emotions, hence the popularity of games
 - Biorhythms are important : find the right moment and the right state of mind





Health Gateway



Adaptive Stress Due to Technology Change Rate



ancis Jauréguiberry

Usages et enjeux

Summary

- Adoption of ICT is fast & widespread but worries are being expressed
- Adaptation stress: Technology > usage > understanding
 « a world of accelerated permanent change »
- A fear of not being in charge (mastery), especially temporally
- Physical experiences are still preferred to virtual ones, additive logic versus substitution

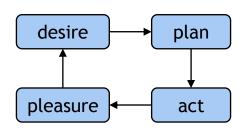
Recommendations

- 1. Reduce stress though training / learning / explanations
- 2. Invite « real back into virtual » : leverage mirror neurons ©
- 3. Life hygiene and detox : promote temporal and spatial rest zones
- 4. Leverage emotions

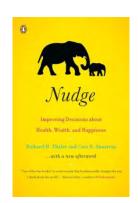


To Build a Successful Digital Experience Means "To Include a Fun Learning Journey"

- Learning without a User Manual
 - Use Analytics: measure discovery and appropriation
 - Continuous feedback with IOT
- Leveraging Pleasure through Games
 - Play with simple emotions (badges/rewards) but also surprise/ ...
 - Cf Mikio Kaku (emotion: evolutionary mechanism of learning)
- Leveraging Communities of Learning
 - Social digital experience « Virality is the product »
 - Learn by imitation (e.g. video tutorial with other users)
- Learn by Doing (Montessori heritage ⁽³⁾)
 - Kinetics memory
- Leverage Cognitive Biases
 - Dan Ariely
 - Richard Thaler
 - Daniel Kahneman

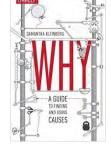


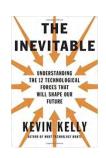




Internet of Things and Life Coaching

- The best tool for behavior change is a "game to learn about yourself"
 - Sensors to monitor yourself and your environment
 - Digital tools to tell a story
 - Data science to generate insights (without the burden of causality)
- Learn by doing: "Yourself as a system" => insights discovery
 (i.e. discover what makes you gain weight versus loosing weight)
 - This is natural for regular "quantified self" user
 - Bringing this experience to everyone is a challenge
 - Leverage behavioral science into digital sciences (e.g., notification cycles)
- This is as much about "small data" as "big data"
 - Personalized healthcare (we are all different)
 cf. Health 4P: Personalized, Preventive, Participative, Predictive
 - Data privacy is a true and rising concerns (privacy by design)







Part III

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Internet of Things Architecture is Grown, not Designed

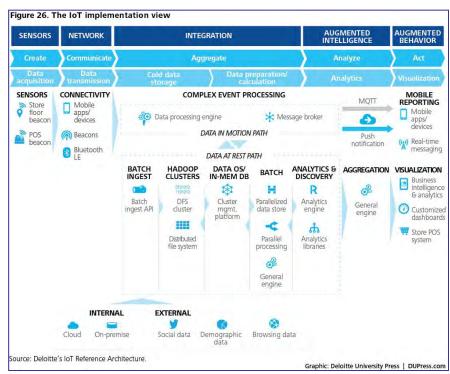
 IoT is foremost a B2B topic, where traditional top-down system engineering approaches make sense ... However the B2C world is governed by customer usage, hence "chaotic" (in a complex systems sense)

Most IOT reference architectures are too hierarchical,

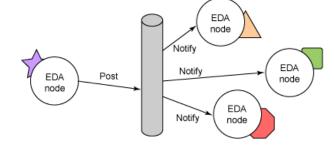
too "Telco-oriented" ©

 One of the best reference is, from my point of view, Deloitte's "implementation view"

- Still this is too:
 - structured,
 - holistic
- ... I advocate for emergent patterns:
 - Systems of Systems
 - Grown from usage (biomimicry)



Event-Driven Architecture



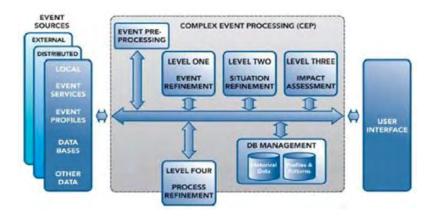
- Event-Driven Architecture is well suited for smart objects systems
 - Scalable, open, distributed
 - Cf. IFTT: open API-based, event-driven architecture



- To adapt continuously to its environment, the preferred architecture is to distribute control and analytics
 - Avoid SPOF, better fault-tolerance
 - Biomimicry for smart home : multiple layers of controls corresponding layers of abstraction since complexity yields fragility
- Redundant and distributed storage
 - Not every piece of data should be moved ...
- Smart systems operate on a multiplicity of time scales
 - cf. lambda-architecture : hot/cold analytics
 - Event-driven learning at multiple time horizons

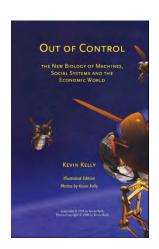
Emergent EDA Systems: « Engineering as Parenting »

- Smart Systems are hierarchical → hierarchies of events
- Complex Event processing → move between abstraction levels



- "Human in the loop" => emergent design
 - Design "conversations" with users
- Common goals must be reified (and distributed)
 - complex systems (teleonomy)
 - Cf SlapOS, OAI, ...
- Biomimetics: "grow what works"
 - E.g., muscle growth
 - Smart Home example : Adhoco

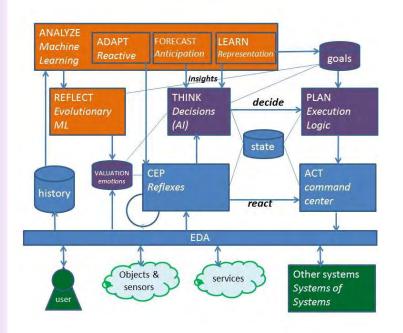




Cognitive EDA Systems

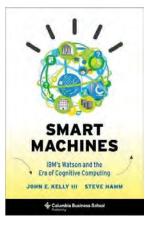
- "The era of cognitive computing"
 - Systems "grown by machine learning, not programmatic design"
 - "Society of minds" Cf. Todai Robot University of Tokyo

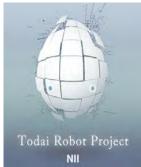




A short overview:

- Reflexes
- Inductive Thinking (ML)
- Deductive Thinking (Rules & Planning)
- "Emotions"
 (evolutionary /
 meta-heuristics control)



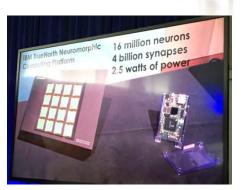


Systems of Systems : Fractal IOT Architecture



- Smart / learning systems are coming
 - Cf. feedback from CES 2016
 - Cf. feedback from Singularity University
 - Ability to integrate cognitive systems into small silicon footprint is growing exponentially ©
- They are embedded into "Systems of Systems"
 - Learning/ data / analytics is distributed & multi-scale
 - Learn from sensors, as close to the data as possible
- It is easier to build really smart smaller objects than large systems, thus they will appear first and "intelligence" will come locally before it comes globally
- Role of meta-learning loops and meta-narrative (coaching)
 - Critical role for SoS (global perspective)





Conclusion

- Big Data represents a set of paradigm shifts
 - Data as code, designing emergent systems though ML
 - The opportunity for disruption is larger than new services and new usages
- Smart data empowers a new generation of smart systems, but they must be designed for human needs
 - IoT does not have to power to change human behavior by itself
 - Educate the user though games because she is the most important part of the smart system
- Event-driven, adaptive and cognitive Internet of Things architecture
 - IoT architecture is derived and built from consumer usage
 - Cognitive large-scale systems will emerge bottom-up since it is easier to derive closed-loop learning systems on a smaller scope