# On Research Challenges in IoT Systems Engineering

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### Smart Systems and the Internet of Things are driven by a combination of:





We are giving our world a digital nervous system. Location data using GPS sensors. Eyes and ears using cameras and microphones, along with sensory organs that can measure everything from temperature to pressure changes.







These networked inputs can then be combined into bi-directional systems that integrate data, people, processes and systems for better decision making.



### Network Neutrality (NN)

All traffic on the Internet must be treated equally.



# Network Neutrality is gone...

T-Mobile Germany Blocks iPhone Skype Over 3G and WiFi

#### James Kendrick Apr 6, 2009 - 5:15 PM CDT





Skype has been one of the top downloaded apps for the iPhone sky since its release last week, even though Apple (s aapl) bowed to AT&T (s T) in the U.S. to prevent the VoIP program from working on 3G. U Skype users are re TECH

on the iPhone, som whatever they can

#### TECHNOLOGY LAB -Netflix performance on Verizon and Comcast has been dropping for months

Latest Netflix data shows some ISPs struggling, while Google Fiber soars.

ION BRODKIN - 2/10/2014, 10:30 PM



### doubt concerned ti Netflix Throttles Its Videos on AT&T, Verizon Networks

Streaming service says it limits video quality to protect users from exceeding data caps

#### By RYAN KNUTSON and SHALINI RAMACHANDRAN

Updated March 24, 2016 10:55 p.m. ET

AT&T Inc. and Verizon Communications Inc. were on the defensive

last week after accusations swirled they were throttling the quality

of Netflix Inc. video on their wireless networks.

It turns out it was Netflix that was doing the throttling.

### Understanding the context



### What about NN and IoT?



Dustdar S. and Duarte, E. (2018) Network Neutrality and its impact on Innovation, IEEE Internet Computing, Nov/Dec 2018

### Smart Evolution – People, Services, and Things







Autonomic Nervous System

# Design Aspects

- Divide & Conquer
  - Complexity, Coherence & Entropy

### Service Engineering Design Strategy

# **Definition of** DIVIDE AND CONQUER

to make a group of people disagree and fight with one another so that they will not join together against one. His military

strategy is to divide and conquer.

# Smart City Example



# Ecosystems: People, Systems, and Things



**Complex system** with networked dependencies and intrinsic adaptive behavior – has:

### 1. Robustness & Resilience

**mechanisms**: achieving stability in the presence of disruption

2. Measures of health: diversity, population trends, other key indicators

### **3. Built-in coherence**

4. Entropy-resistence

Marine Ecosystem: http://www.xbordercurrents.co.uk/wildlife/marine-ecosystem-2

### Ecosystems for IoT Systems



### Perspectives on the IoT: Edge, Cloud, Internet



# Cloud-centric perspective

### Assumptions

 Cloud provides core services; Edge provides local proxies for the Cloud (offloading parts of the cloud's workload)

### Edge Computers

- play supportive role for the IoT services and applications
- Cloud computing-based IoT solutions use cloud servers for various purposes including massive computation, data storage, communication between IoT systems, and security/privacy

### Missing

- In the network architecture, the cloud is also located at the network edge, not surrounded by the edge
- Computers at the edge do not always have to depend on the cloud; they can operate autonomously and collaborate with one another directly without the help of the cloud

### Internet-centric perspective

### Assumptions

- Internet is center of IoT architecture; Edge devices are gateways to the Internet (not the Cloud)
- Each LAN can be organized around edge devices autonomously
- Local devices do not depend on Cloud

### Therefore

- Things belong to partitioned subsystems and LANs rather than to a centralized system directly
- The Cloud is connected to the Internet via the edge of the network
- Remote IoT systems can be connected directly via the Internet. Communications does not have to go via the Cloud
- The Edge can connect things to the Internet and disconnect traffic outside the LAN to protect things ->
   IoT system must be able to act autonomously







Fig. 1. Scenario with a wearable ECG sensor and a smartphone (1:1 mapping of an IoT to edge device).

Fig. 2. A use case scenario of a wearable ECG sensor and a multiple number of smartphones (1:N mapping of an IoT to edge devices).



Fig. 3. Scenario with multiple wearable ECG sensors and a tablet (N:1 mapping of IoT devices to an edge device).

# Cloud-loT vs. Edge/Cloud hybrid IoT



Fig. 4. Cloud-based IoT solution.

Fig. 5. Edge computing architecture for a streaming IoT solution.

### Vertical vs. Horizontal Edge Architecture



Fig. 6. Vertical edge computing architecture for a streaming IoT solution.



Fig. 7. Horizontal edge computing (serverless) architecture of a streaming IoT solution.

### Ecosystem "Society 5.0" (Japan)



Toward realization of the new economy and society, Keidanren (Japan Business Federation), April 2016

### Linear History? Ancient "Computers"





Adam's Calender, Michael Tellinger



**Stonehenge: A Neolithic Computer** *Nature* **202**, 1258 - 1261 (27 June 1964); doi:10.1038/2021258a0



### Collective mess....



September 28



"The scientists of today think <u>deeply</u> instead of <u>clearly</u>.

One must be sane to think clearly, but one can think deeply and be quite insane."





"You cannot hope to build a better world without improving the individuals.

To that end each of us must work for his own improvement, and at the same time share a general responsibility for all humanity,

our particular duty being to aid those to whom we think we can be most useful."

### **Marie Curie**

# If you want to change the world, you have to change the metaphor.

Joseph Campbell

### Assumptions, Models, and Abstractions

- Co-evolution of Science & Technologies
- Smart Cities as models of ecosystems: -> People, Things, and Systems
- Models as abstractions are useful (Platonic Forms)
- We lack a model for such an ecosystem
- From automation to creativity support





IMMANUEL KANT

 Consciousness and creativity support -> lead to new (meta) models and understanding of technologies and science -> Architecture of Values

### Layers of Paradigms

- Not reductionist
- We have to create the abstractions and models we want based on our understanding of human and societal needs
- Ecosystems = Architecture, Structure + Dynamics
- New Paradigms: (1) Elastic Computing,
   (2) Social Compute Units, (3) Osmotic Computing
- Emergent properties on higher levels with own properties



# Paradigm 1: Elasticity (Resilience)

(Physics) The property of returning to an initial form or state following deformation

# stretch when a force stresses them

e.g., *acquire* new resources, *reduce* quality

### shrink when the stress is removed

e.g., release resources, increase quality

# Elastic Computing > Scalability



### **Resource elasticity**

Software / human-based computing elements, multiple clouds



### Quality elasticity

Non-functional parameters e.g., performance, quality of data, service availability, human trust

Dustdar S., Guo Y., Satzger B., Truong H. (2012) <u>Principles of Elastic</u> <u>Processes</u>, IEEE Internet Computing, Volume: 16, <u>Issue: 6</u>, Nov.-Dec. 2012

Elasticity



### elasticity

rewards, incentives

# Specifying and controling elasticity

#### Computing 16(6): 72-77 (2012) Elasticity directive primitives Basic primitives Monitoring Strategy Constraint Resource/ Scale Stop/wait Resource Quality Cost Configure Access quality/cost in/out /notify Data Performance People Storage Network Compute quality

SYBL (Simple Yet Beautiful Language) for specifying elasticity requirements

#### **SYBL-supported requirement levels**

**Cloud Service Level** 

Service Topology Level

Service Unit Level

**Relationship Level** 

Programming/Code Level

#### **Current SYBL implementation**

#### in Java using Java annotations

@SYBLAnnotation(monitoring=,,",constraints=,,",strategies=,,")

#### in XML

<ProgrammingDirective><Constraints><Constraint name=c1>...</Constraint></Constraints>...</ProgrammingDirective >

Dustdar, S. et al.: Programming

**Directives for Elastic** 

Computing. IEEE Internet

#### as TOSCA Policies

<tosca:ServiceTemplate name="PilotCloudService"> <tosca:Policy name="St1" policyType="SYBLStrategy"> St1:STRATEGY minimize(Cost) WHEN high(overallQuality) </tosca:Policy>...

# Specifying and controling elasticity of humanbased services



# High level elasticity control

#### #SYBL.CloudServiceLevel

Cons1: CONSTRAINT responseTime < 5 ms Cons2: CONSTRAINT responseTime < 10 ms WHEN nbOfUsers > 10000 Str1: STRATEGY CASE fulfilled(Cons1) OR fulfilled(Cons2): minimize(cost)

#### **#SYBL.ServiceUnitLevel** Str2: STRATEGY CASE ioCost < 3 Euro : maximize( dataFreshness )

### **#SYBL.CodeRegionLevel** Cons4: CONSTRAINT dataAccuracy>90% AND cost<4 Euro



Georgiana Copil, Daniel Moldovan, Hong-Linh Truong, Schahram Dustdar, "SYBL: an Extensible Language for Controlling Elasticity in Cloud Applications", 13th IEEE/ACM International Symposium on Cluster, Cloud and Grid Computing (CCGrid), May 14-16, 2013, Delft, Netherlands

Copil G., Moldovan D., Truong H.-L., Dustdar S. (2016). **rSYBL: a Framework for Specifying and Controlling Cloud Services Elasticity**. ACM Transactions on Internet Technology

### Elasticity Model for Cloud Services

Moldovan D., G. Copil, Truong H.-L., Dustdar S. (2013). MELA: Monitoring and Analyzing Elasticity of Cloud Service. CloudCom 2013



# Paradigm 2: Social Compute Units (SCUs)



Dustdar S., Bhattacharya K. (2011). <u>The Social Compute Unit</u>, *IEEE Internet Computing*, Volume 15, Issue 3; pp. 64 - 69.

Fernández P., Truong H.-L., Dustdar S., Ruiz-Cortés A. (2015). <u>Programming</u> <u>Elasticity and Commitment in</u> <u>Dynamic Processes</u>. IEEE Internet Computing, Volume 19, Number 2, pp. 68 - 74

### Elastic SCU provisioning (Paradigms 1 and 2 together)



Mirela Riveni, Hong-Linh Truong, and Schahram Dustdar, On the Elasticity of Social Compute Units, CAISE 2014 Muhammad Z.C. Candra, Hong-Linh Truong, and Schahram Dustdar, **Provisioning Quality-aware Social Compute Units in the Cloud, ICSOC 2013.** 

# Paradigm 3: Osmotic Computing

- In chemistry, "osmosis" represents the seamless diffusion of molecules from a higher to a lower concentration solution.
- Dynamic management of (micro)services across cloud and edge datacenters
  - deployment, networking, and security, ...
  - providing reliable IoT support with specified levels of QoS.

Villari M., Fazio M., Dustdar S., Rana O., Ranjan R. (2016). <u>Osmotic</u> <u>Computing: A New Paradigm for Edge/Cloud Integration</u>. *IEEE Cloud Computing*, Volume 3, Issue 6, pp. 76-83





### IoT & Data Science – Research Challenges

Ranjan R., Rana O., Nepal S., Yousif M., James P., Wen Z., Barr S., Watson P., Jayaraman P. P., Georgakopoulos D., Villari M., Fazio M., Garg S., Buyya R., Wang L., Zomaya A. Y., Dustdar S. (2018). <u>The Next Grand Challenges: Integrating</u> <u>the Internet of Things and Data Science</u>, *IEEE Cloud Computing*, Volume 5, Issue 3, pp. 12-26



#### The Next Grand Challenges Integrating the Internet of Things and Data Science

#### Rajiv Ranjan Newcastle University

Omer Rana Cardiff University

Surya Nepal Data61, CSIRO This article discusses research challenges related to devising a new IoT programming paradigm for orchestrating IoT applications' composition and data processing across heterogeneous computing infrastructure (Cloud, Edge, and Things).

#### Mazin Yousif T-Systems, International

Dhilin James 7hann Wen

In the last decade, we have been transitioning from a date

### IoT-driven ecosystems



# IoT/Data/Application Orchestration



# Osmotic movement of MELs in Clouds, Edge, Things



Legend: MEL...Micro Element

### IoT Data Sources

- **1. Representation**: Structure and represent the data to facilitate multiple modalities, exploiting the complementarity and redundancy of different data sources.
- **2. Translation**: Interpret data from one modality to another, i.e., provide a translator that allows the modalities to interact with each other for enabling data exchange.
- **3. Alignment**: Identify the relation among modalities. This requires identifying links between different types of data.
- **4.** Fusion: Fuse information from different modalities (e.g., to predict).
- **5. Co-learning**: Transfer knowledge among modalities. This explores the field of how the knowledge of a modality can help or enhance a computational model trained on a different modality.

# IoT Mircoelements (MELs)

- MicroServices (MS), which <u>implement specific functionalities</u> and can be <u>deployed</u> and migrated across different virtualized and/or <u>containerized infrastructures</u> (e.g., Docker) available across Cloud, Edge, and Things layers
- 2. MicroData (MD), encodes the contextual information about (a) the sensors, actuators, edge devices, and cloud resources it needs to <u>collect data from or send data to</u>, (b) the <u>specific type</u> of data (e.g., temperature, vibration, pollution, pH, humidity) it needs to <u>process</u>, and (c) other <u>data manipulation operations</u> such as where to <u>store</u> data, where to forward data, and where to store results
- **3.** MicroComputing (MC), <u>executing specific types of computational tasks</u> (machine learning, aggregation, statistical analysis, error checking, and format translation) based on a mix of historic and real-time MD data in heterogeneous formats. These MCs could be realized using a variety of data storage and analytics programming models (SQL, NoSQL, stream processing, batch processing, etc.)
- **4. MicroActuator** (MA), <u>implementing programming interfaces</u> (e.g., for sending commands) with actuator devices for changing or controlling object states in the IoT environment

### IoT Programming Patterns needed

- **1. Decomposing IoT data analysis activities into fine-grained activities** (e.g., statistics, clustering, classification, anomaly detection, accumulation, filtering), each of which may impose different planning and run-time orchestration requirements;
- 2. Identifying and integrating real-time data from IoT devices and historical IoT data distributed across Cloud and Edge resources;
- 3. Identifying data and control flow dependencies between data analysis activities focusing on coordination and data flow variables, as well as the handling of dynamic system updates and re-configuration;
- 4. Defining and tagging each **data analysis activity with runtime deployment constraints** (QoS, security and privacy).

### Motivating Case Studies

#### **Building Management System**

• Manages building facilities, e.g., HVAC systems, elevators and emergency alarms



#### **Fleet Management System**

 Manages fleets of electric vehicles worldwide (e.g., on golf courses)





# Motivation

- Lack of systematic support and tools for developing, deploying, and operating IoT systems (Cloud, Fog, and Edge)
- Today IoT systems are vertically closed and tightly coupled
  - Hard to develop and maintain applications
  - Difficult to operate and reuse existing infrastructure



### **Programming Model for IoT Systems**

### Motivation



**Requirements:** 

- *Application:* Custom configuration and behavior of Sens./Act.
- *Runtime:* Dealing with constrained resources
- *Developer:* Domain expert knowledge

Requirements:

- *Application:* Should be generic (independent of underlying devices)
- *Runtime:* Dealing with scalability and elasticity concerns
- *Developer:* Software engineering expertise

### Approach



Programming Model for IoT Applications on Cloud Platforms. SOCA 2013. Hawaii, USA.



# Intent-based Programming Model



 Enable developing loosely coupled applications Intent Structure

Trade expressiveness for more flexible and easier application development

# Some final reflections

# **Beyond Turing**

- Can a machine-only system really be considered "intelligent"?
  - Going <u>beyond Turing Test</u>... (Alexa, Siri, Cortana)
  - Why not utilize <u>societal intelligence</u>? ... and not try to match the intelligence of a *single* human individual?
- Integrate AI, IoT, and human collectives into processes!



Schahram Dustdar Stefan Nastić Ognjen Šćekić

### Smart Cities

The Internet of Things, People and Systems



🖄 Springer

### Thanks for your attention

Prof. Schahram Dustdar

IEEE TCSVC Outstanding Leadership Award in Services Computing Member of Academia Europaea IBM Faculty award ACM Distinguished Scientist IEEE Fellow

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Examples of topics relevant to the journal are:

- Real-world applications, application designs, industrial case studies and user experiences
  of IoT technologies, including standardization and social acceptance
- Communication networks, protocols and interoperability for IoT
- IoT data analytics, machine learning, and associated Web technologies
- Wearable and personal devices, including sensor technologies
- Human-machine and machine-machine interactions
- Edge, fog, and cloud computing architectures
   Neural IoT software architectures consistent middlew
- Novel IoT software architectures, services, middleware as well as future Internet designs
   Fusion of social and physical signals in IoT services
- Fusion of social and physical signals in IoT services
   Non-functional properties of IoT systems, e.g., dependability, timeliness, security
- and privacy, robustness
- Testbeds for IoT

All submissions are expected to provide experimental evidence of their effectiveness in realistic scenarios (e.g., based on field deployments or user studies) and the related datasets. The submission of purely theoretical or speculative papers is discouraged, and so is the use of simulation as the sole form of experimental validation.

Experience reports about the use or adaptation of known systems and techniques in real-world applications are equally welcome, as these studies elicit precious insights for researchers and practitioners alike. For this type of submissions, the depth, rigor, and realism of the experimental component is key, along with the analysis and expected impact of the lessons learned.

For further information, please contact tiot-editors@acm.org.