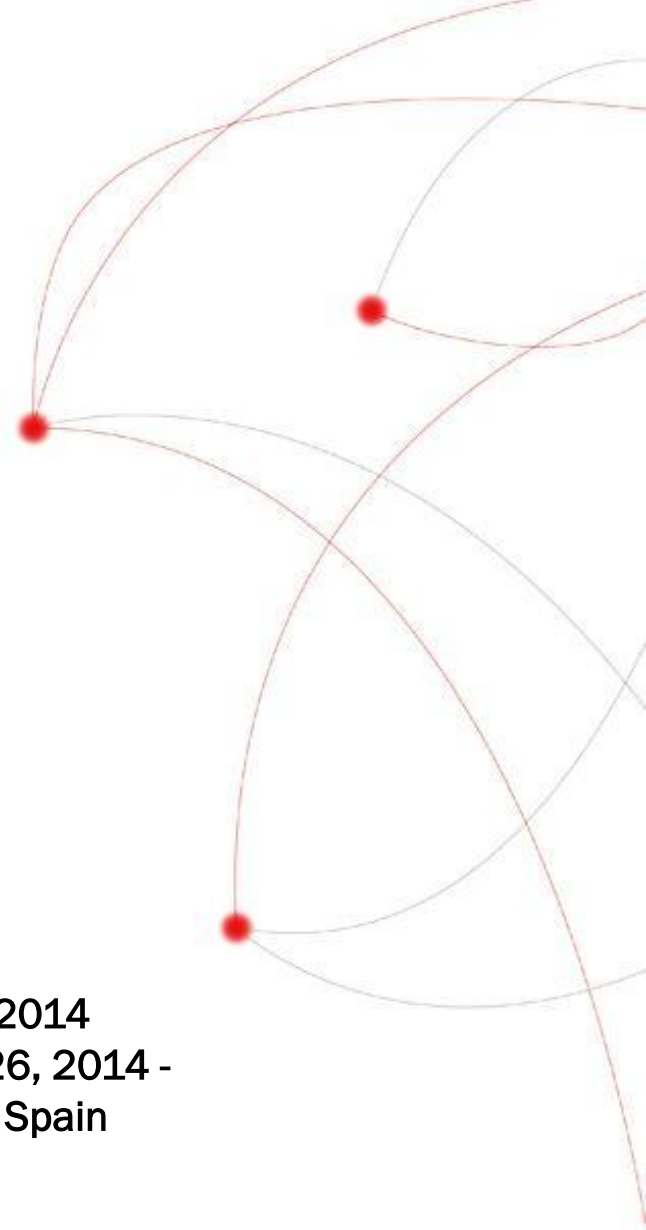


**PANEL: Information,
Information, Information...
How Much can we Handle?**



ICCGI 2014
June 22 - 26, 2014 -
Seville, Spain



PANEL: Information, Information, Information... How Much can we Handle?

PANELISTS

- ❑ **Arno LEIST, Massey University, New Zealand**
- ❑ **Félix BUENDÍA, Universidad Politécnica de Valencia, Spain**
- ❑ **Jefferson CAPOVILLA, CPqD, Brazil**
- ❑ **Antonio J. TALLÓN-BALLESTEROS, University of Seville, Spain**

**Structure of the Panel:
Introduction on the topic
5 minutes speech per panelist**

Discussion with the Audience is the key!

Who am I?

Claudio BOREAN

- ❑ **Leading SWARM – Beyond the Internet of Things Joint Open Lab of Telecom Italia**
- ❑ **Working on new innovative technical solutions leveraging on collaborative approaches among smart devices (Internet of Everything)**
- ❑ **Application fields: distributed energy systems, “swarm homes”, collective intelligence for mobile applications**

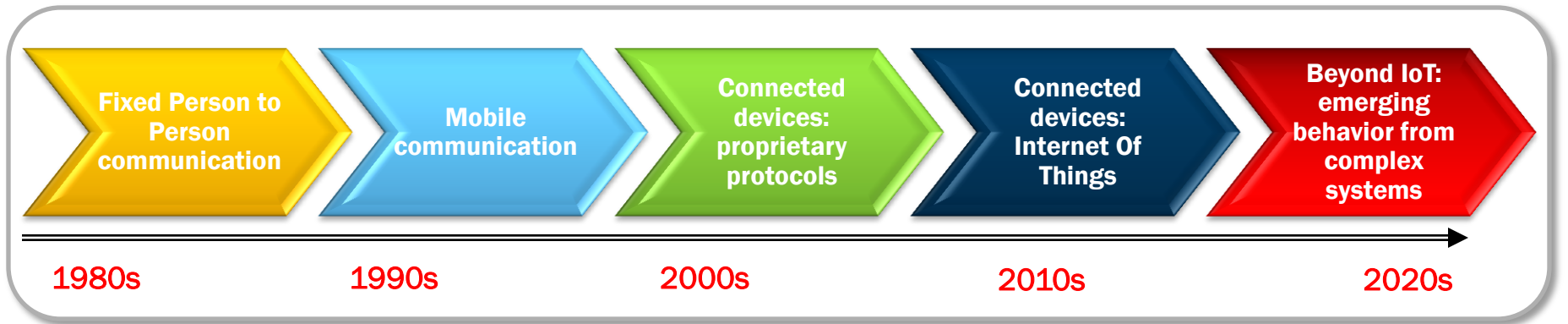
Areas: Swarm intelligence, Internet of Things, cooperative distributed applications

SWARM

Beyond the Internet of Things

<http://jol.telecomitalia.com/jolswarm/>

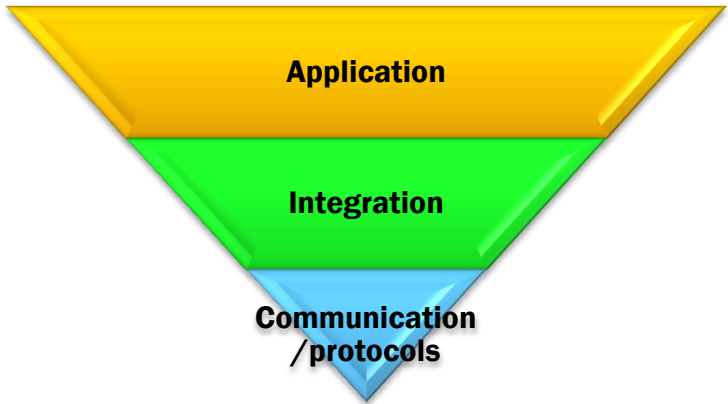
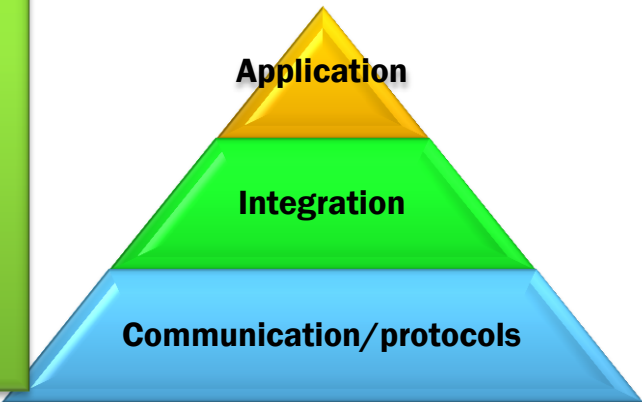
Topic summary: From “connected devices” to “beyond IoT” and “IoE”



FROM:

TO:

VALUE



PRESENTATIONS FROM PANELIST

DISCUSSION



MASSEY UNIVERSITY
TE KUNENGA KI PŪREHUROA

UNIVERSITY OF NEW ZEALAND

InfoWare 2014: ICCGI Panel

Information, Information, Information...
How Much Can We Handle?

Arno Leist, PhD
a.leist@massey.ac.nz

Who Am I?

- Lecturer in Computer Science
- Massey University, Auckland
- Research interests
 - Parallel computing
 - Data-parallel accelerators
 - Computational simulations
 - Visualisation



Processing Data

- A computational challenge
 - Processors are not getting much faster any more
 - But they are getting more parallel
- All components of the system must scale with the problem size
 - Processors
 - Memory bandwidth
 - Network bandwidth
 - Algorithms (limited by Amdahl's law)
 - Parallel frameworks

Processing Data

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 - Algorithms (limited by Amdahl's law)
 - Parallel frameworks
- What we call *big data* today is just data tomorrow

What To Use?

- Which architecture and environment should your software target?
 - x86, ARM, Power, ...
 - Embedded, mobile, PC, servers, compute clusters, supercomputers
 - Accelerators
 - Cloud computing
- Which languages / frameworks should you use?
 - Programming languages: C++, Java, Go, Python, CUDA, OpenCL, ...
 - Multi-threading libraries: PThreads, OpenMP, ...
 - Multi-tasking libraries: TBB, Cilk Plus, ...
 - Data-parallelism: intrinsic instructions, ...
 - Distributed computing: MPI, Hadoop, ...
- Where and how should you store your data?

Discussion

- Where do you see parallel computing going in the next 5 to 10 years?
- What is keeping *you* from utilising all the parallelism available in your computer?
- Do we really have to process *all* that data?

Mobile Technologies

New chances for education

Félix Buendía

**Ninth International Multi-Conference on
Computing in the Global Information
Technology
ICCGI 2014
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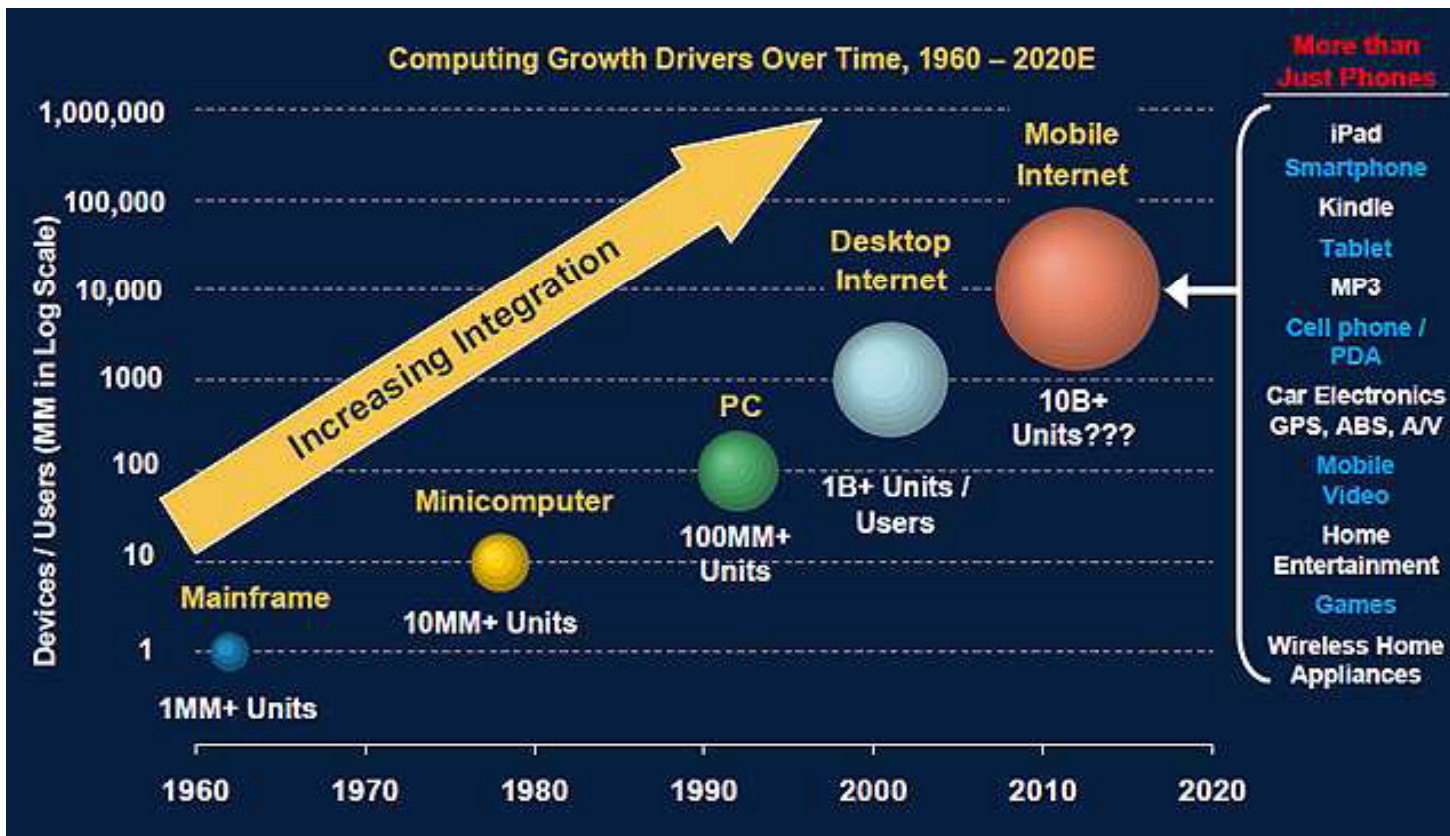
UNIVERSIDAD
POLITECNICA
DE VALENCIA

PAID-UPV/2791



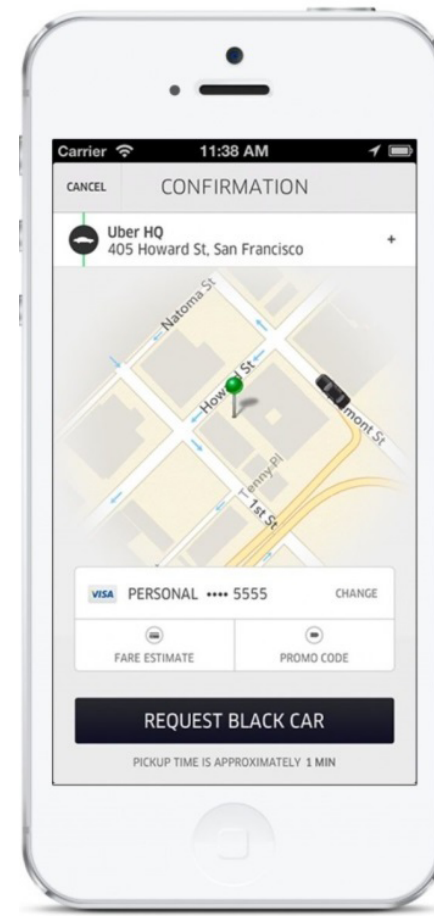
Mobile technologies

- Evolution towards Mobile Internet



Mobile applications

- Moving to mobile apps



<http://blog.uber.com/2012/04/18/chicago-taxi-uber/>

Mobile education

- Platforms & environments



Mobile education

- Current status
 - Mobile Training Implementation Framework (MoTIF)



<http://www.adlnet.gov/mobile-learning/motif/>

Mobile education

- Challenges & opportunities
 - Need to address
 - supporting alternative learning/instructional methods
 - leveraging the capabilities of the mobile platform
 - Some questions
 - Is enough converting existing eLearning courses /MOOC by only “resizing” them to smaller screens?
 - What services and functionalities are required in the mobile learning context?



Information, Information, Information... How Much can we Handle?

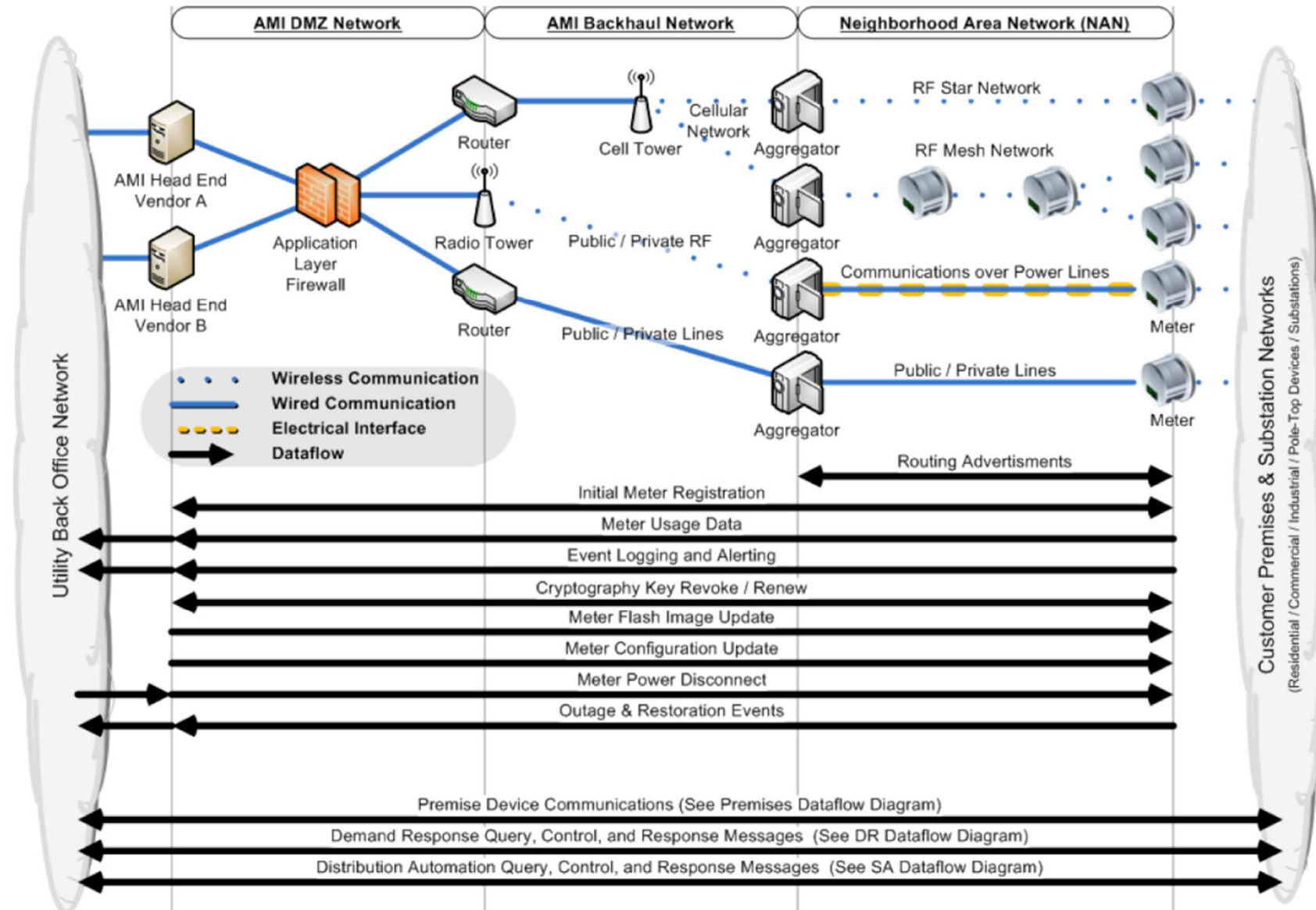
Jefferson Capovilla – CPqD
E-mail: jrodrigo@cpqd.com.br

TURNING
INTO REALITY

Introduction

- CPqD: nonprofit R&D organization*
- Project: security analysis of smart meters
- Big data scenarios:
 - Smart grid dataflow
 - Binary firmware analysis

Smart grid Dataflow



Smart grid dataflow

- Amount of data proportional to:
 - Number of smart meters installed in the grid
 - Control and monitoring messages
- Each communication channel presents attack points
 - Needs of security features for system reliability
- Data flow volume for 27 million smart meters*:

| Meter Type | Single Meter p.a. | Total Meter Population p.a. |
|--------------------|--------------------------|------------------------------------|
| Electricity | Less than 1.5 MB | 30 - 40 TB ³⁶ |

*Data extracted from:
http://www.energynetworks.org/modx/assets/files/electricity/futures/smart_meters/ENA-CR008-001-1%204%20_Data%20Traffic%20Analysis.pdf

Firmware binary analysis

- Firmware: embedded software running in smart meter
- As part of the methodology of smart meter security analysis:
 - Firmware binary extraction from non-volatile memory
 - Firmware binary disassembly(IDA-pro)
 - Firmware binary code analysis
 - Firmware binary exploitation
- Firmware assembly code: > 30000 lines
- Tends to increase with new functionalities in smart-meters
- Analysis performed manually with some tools help

Firmware Binary Analysis

- Example of firmware analysis using IDA-pro

The screenshot displays the IDA Pro interface for a firmware binary analysis. The main window shows assembly code for the function `loc_373768E3`. The code includes instructions for pushing registers, calling `deleteFile`, and creating files. A control flow graph (CFG) is visible in the center, showing a loop structure with nodes `sub_373768E3`, `sub_373768E4`, `sub_373768E5`, and `sub_373768E6`. The `CloseHandle` function is highlighted in the graph. The right side of the interface shows the `WinGraph32` graph, the `Structures` window, and the `Program Segment Action` table. The `Program Segment Action` table lists various segments and their actions:

| Name | Start | End |
|-------|----------|----------|
| text | 37371000 | 37378000 |
| idata | 37378000 | 37378184 |
| idata | 37378184 | 37379000 |
| data | 37379000 | 37370788 |

At the bottom, a command window shows the results of various search operations, including `Command "JumpEnter" failed` and `Search completed. Found at: 37376910.`



Thank You!

Contact:

Jefferson Capovilla– CPqD

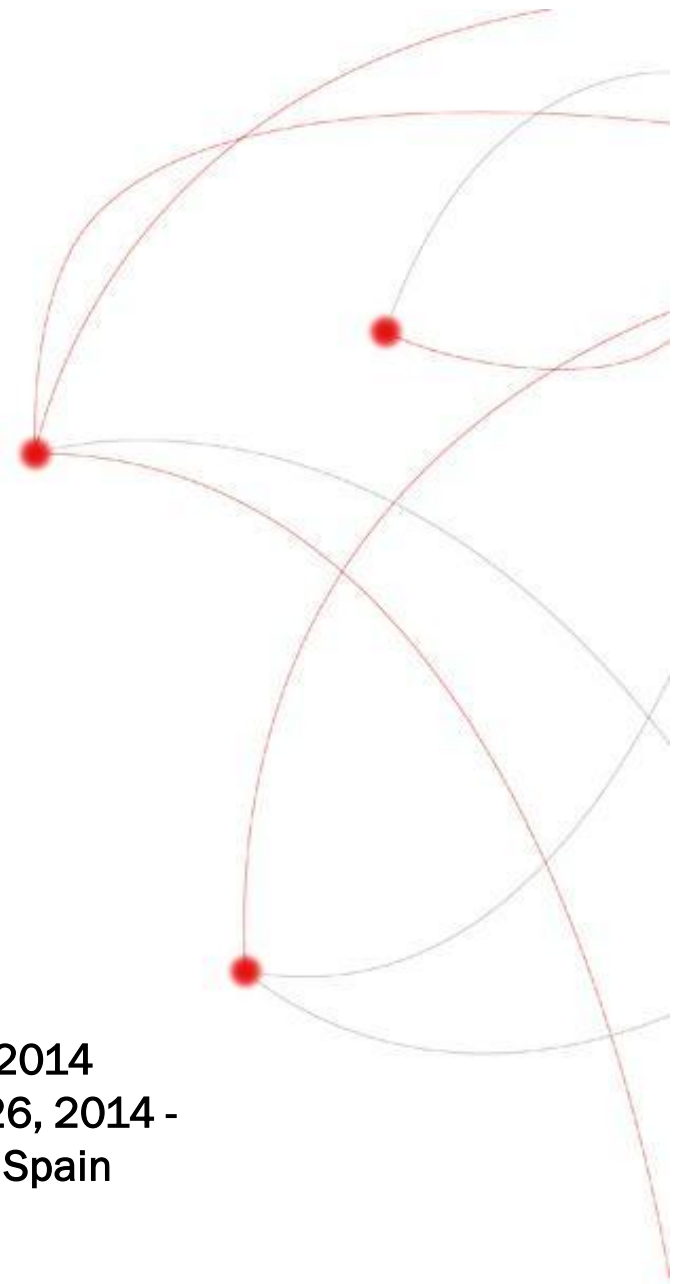
E-mail: jrodrigo@cpqd.com.br

www.cpqd.com.br

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Structure of the Panel:
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Information Processing in Machine Learning (IPML)

Antonio J. Tallón-Ballesteros, Ph. D.

Lecturer

Department of Languages and Computer Systems.

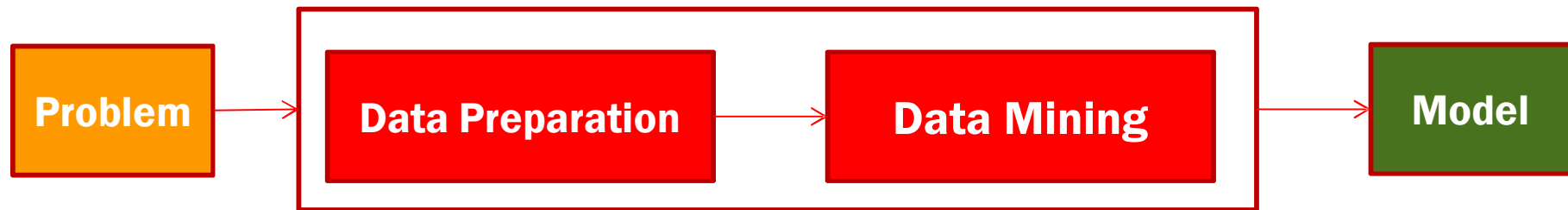
University of Seville (Spain)

atallon@us.es

Research Areas:

Data Mining, Supervised Learning, Neural Networks, Evolutionary Algorithms

IPML (1)



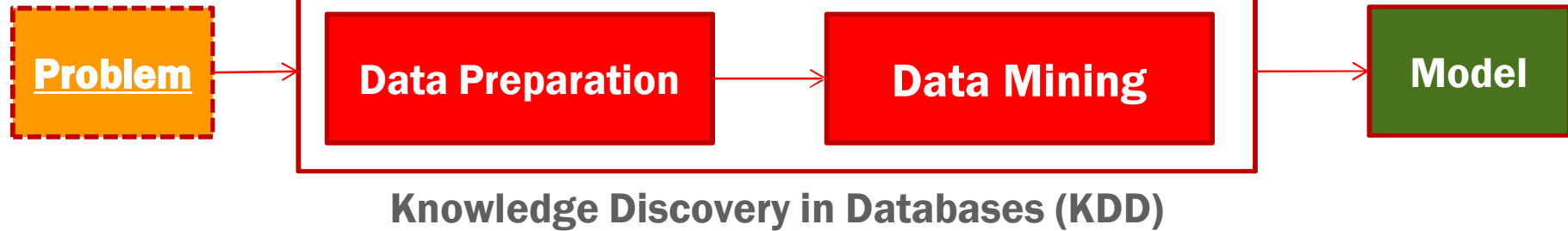
Knowledge Discovery in Databases (KDD)

- ❑ **Complexity**
- ❑ ~~Storage units~~
(gigas, ...)
- ❑ **Instances, features and/or classes**
- ❑ **Data cleansing**
(missing, attribute noise*)
- ❑ **Feature Selection**
- ❑ **Instance Selection**
- ❑ **Specific tools: WEKA, Orange,...**
- ❑ **Classification approaches**
 - ❑ Decision trees (C4.5)
 - ❑ Based on Naive Bayes
 - ❑ Rule-based classifiers
 - ❑ k-Nearest Neighbour (k-NN)

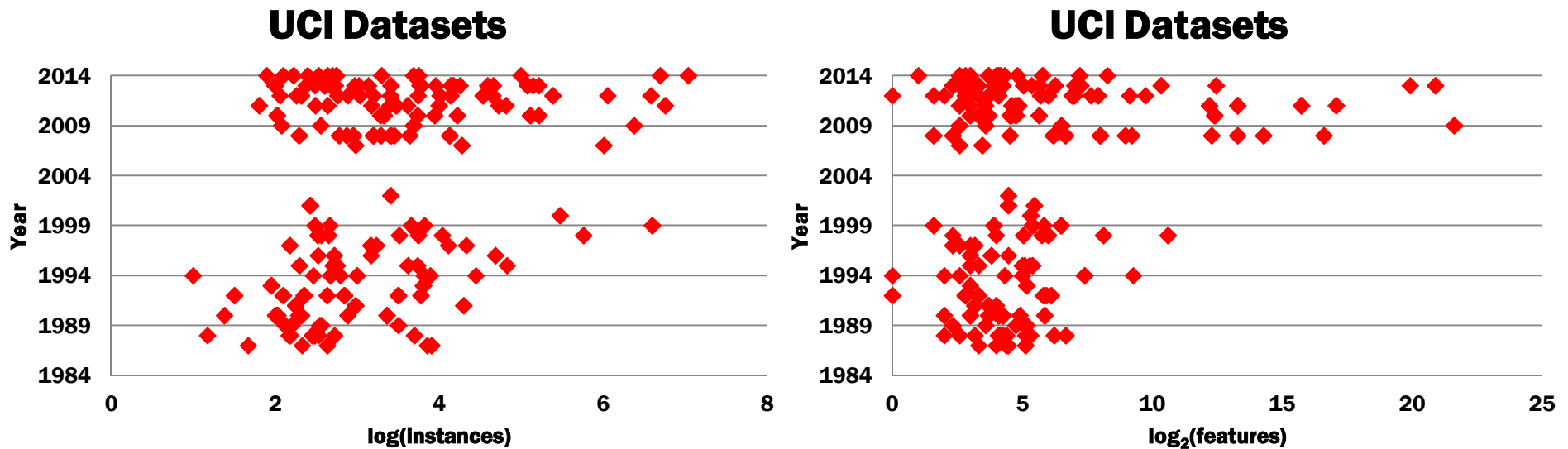
* **Deleting or Keeping Outliers for Classifier Training?** Tallón-Ballesteros, A.J. and Riquelme, J.C. (to appear in **IEEE NaBIC 2014**)

IPML (2)

ICCGI 14
PANEL

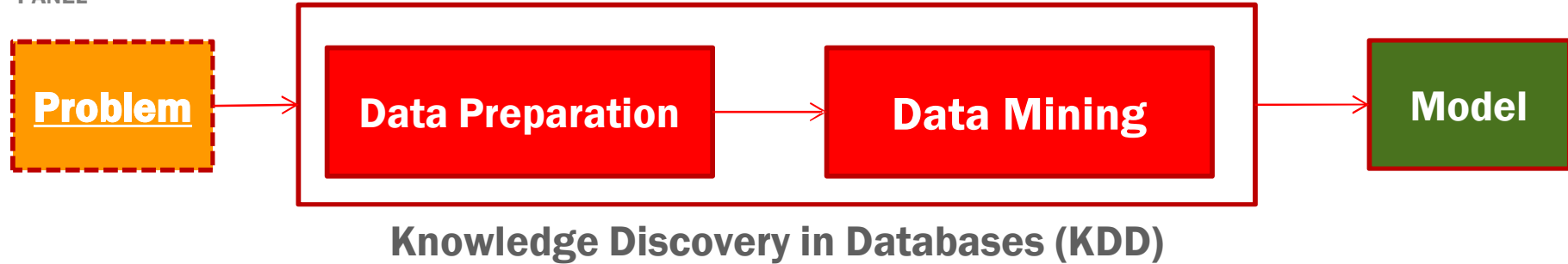


- How much can we handle?
- UCI repository (Univ. of California at Irvine) → ~ 200 problems

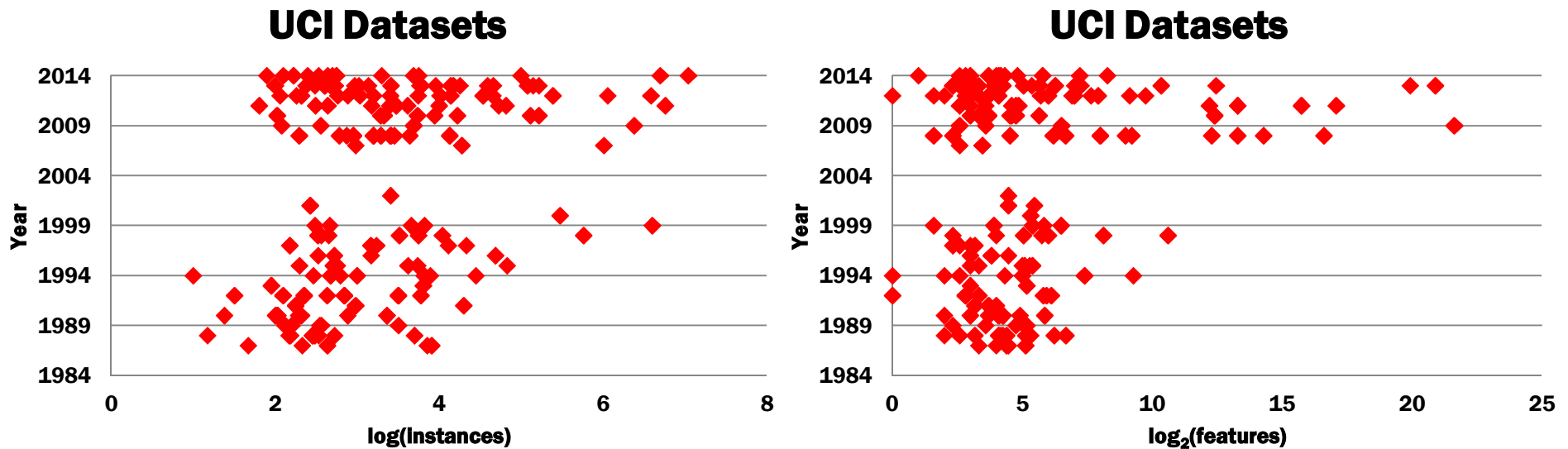


IPML (3)

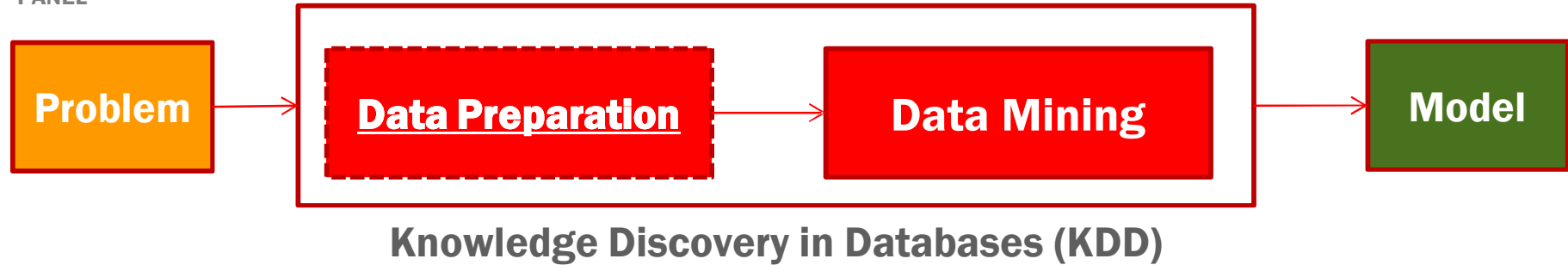
ICCGI 14
PANEL



- Number of instances is increasing (Top 1: 2014)
- Number of features stopped the increase 5 years ago (Top 1: 2009)

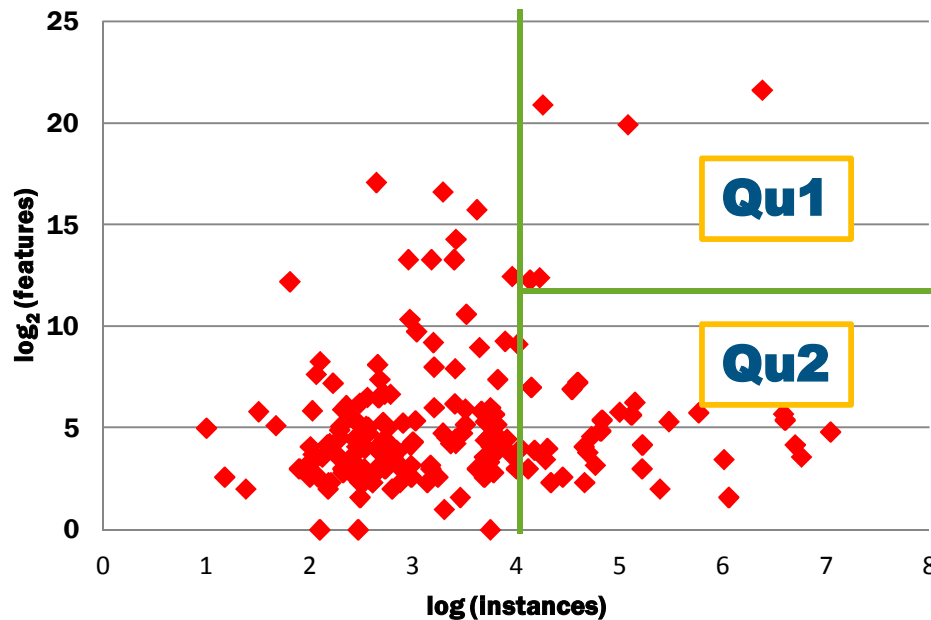


IPML (4)



□ Which preparation techniques must be applied?

UCI Datasets



FS + IS

X=4 → ~ 10 Thousand INST.
X=6 → ~ 1 Million INST.

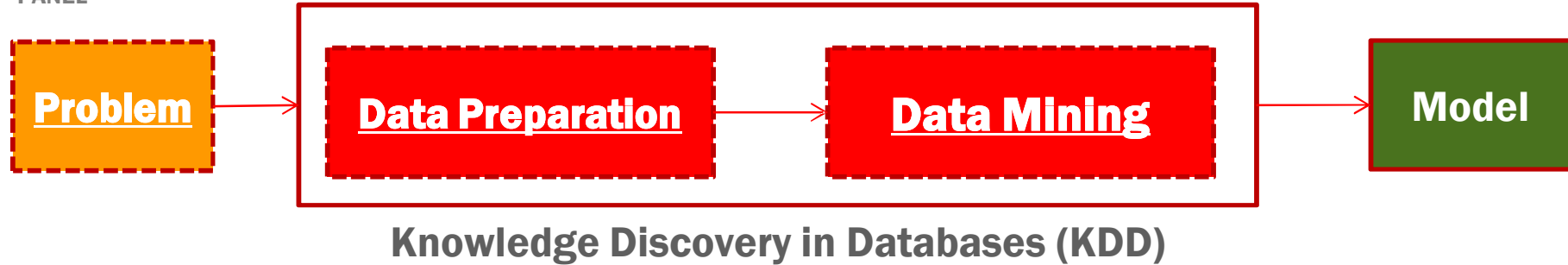
FS + IS

Y=5 → ~ 32 FEAT.
Y=10 → ~ 1 Thousand FEAT.

IS

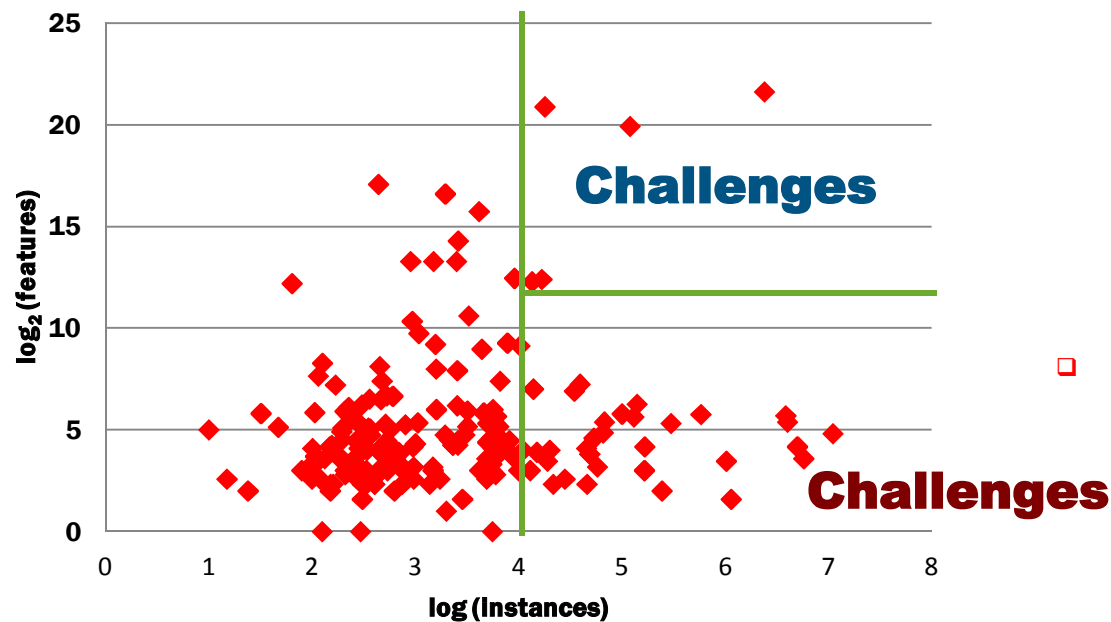
IPML (5)

ICCGI 14
PANEL



Top 3 problems

UCI Datasets



W.r.t. features:

- URL Reputation (2009): Computer
- Gas sensor (2013): Chemistry
- YouTube (2013): Computer

W.r.t. instances:

- HIGGS (bossons) (2014): Physics
- Record Linkage (2011): Computer
- SUSY (particles) (2014): Physics