

WESTERN SYDNEY
UNIVERSITY



Big Data Analytics in Fog Computing

Dr Bahman Javadi

School of Computing, Engineering and Mathematics,
Western Sydney University, Australia



2nd International Conference on
Big Data and Cloud Computing (ICBDCC)



Articles published by
Springer

ICBDCC18 Karunya Institute of Technology & Sciences, Coimbatore, Tamilnadu, India - 641 114

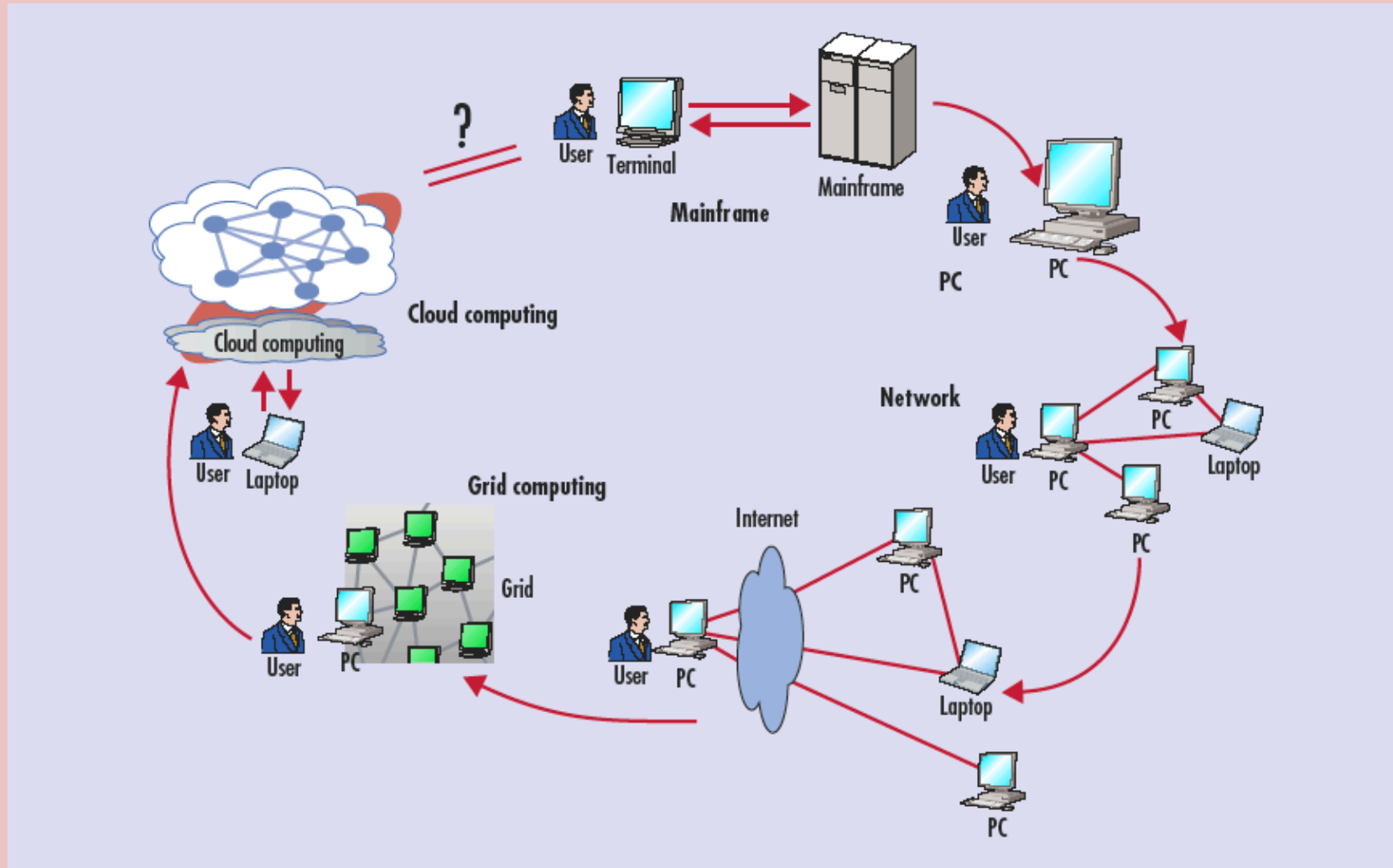


23 March 2018

Outline

- Introduction
- Integration of IoTs in Cloud Computing
- Big Data Analytics in Fog Computing
- FOG-Engine: Fog Computing Realization
- Case Study: Smart Nutrition Monitoring System using Fog Computing
- Open Issues and Challenges

Computing Paradigm Shift

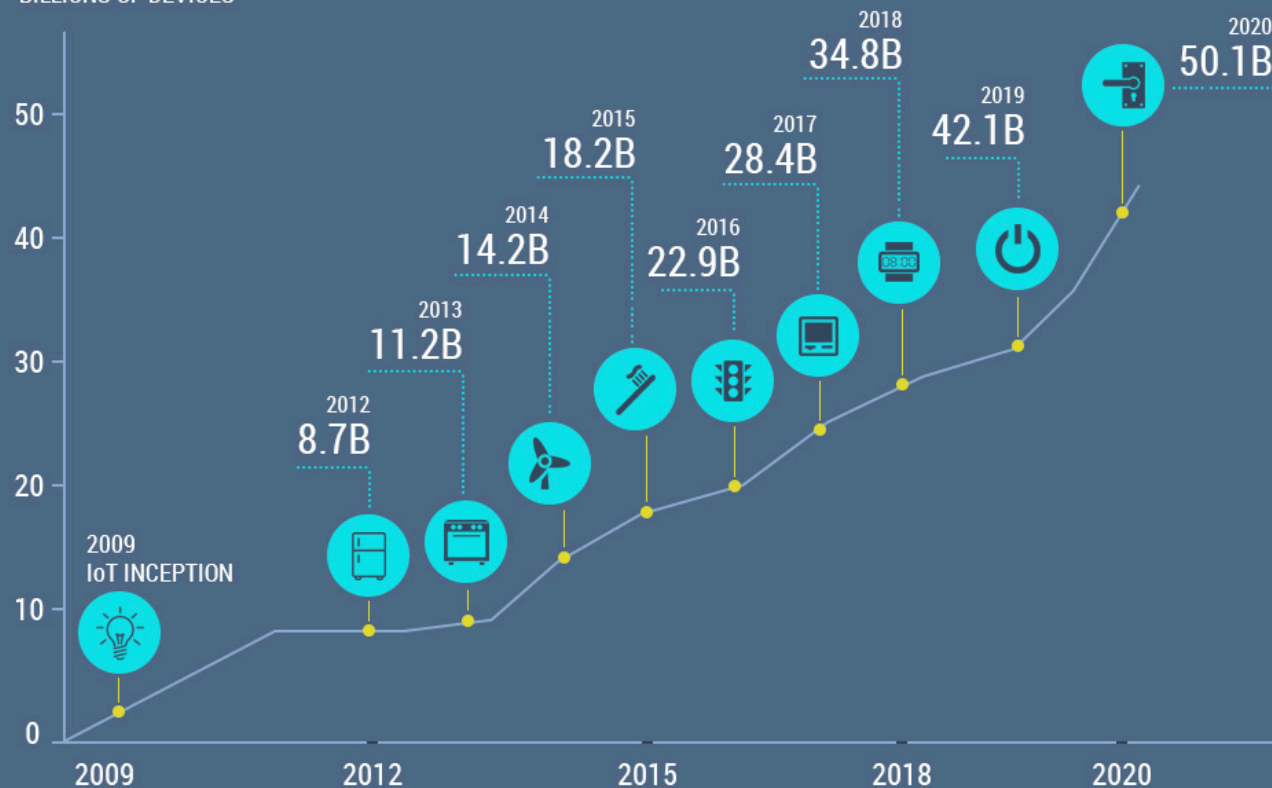


More Connected Devices on the Planet Today Than People

GROWTH OF THE IoT

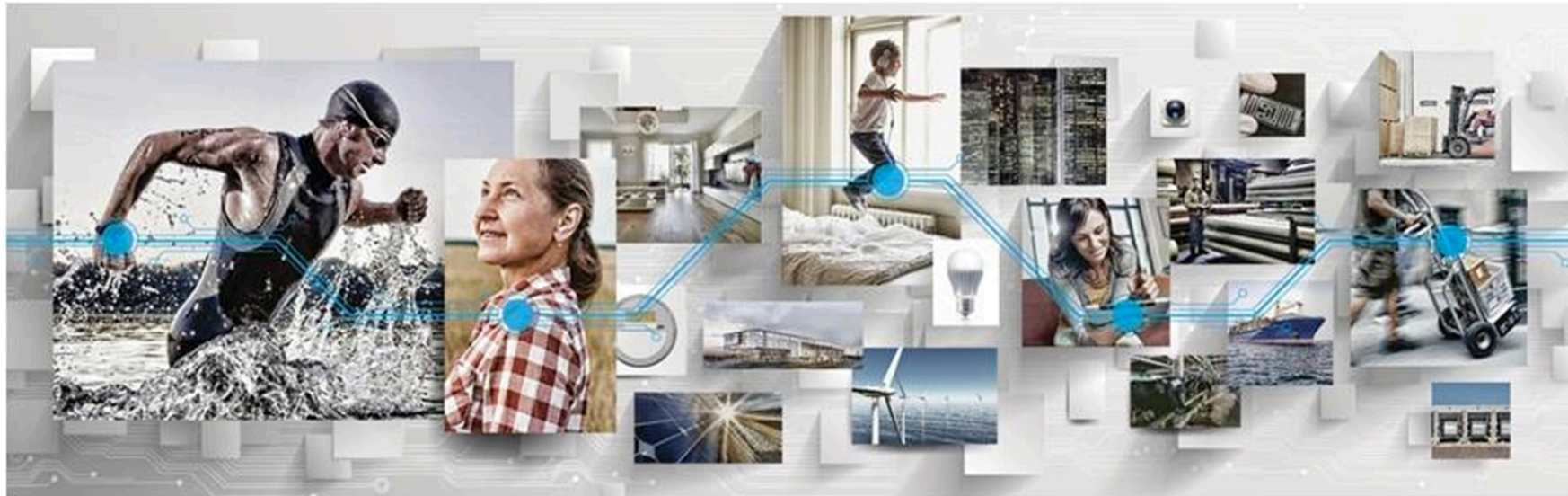
THE NUMBER OF CONNECTED DEVICES WILL EXCEED 50 BILLION BY 2020

BILLIONS OF DEVICES



SOURCE CISCO

Internet of Things Applications



**Smart
Grid**



**Safety
Security**



**Connected
Home**



**Building
Automation**



**Lighting
Control**



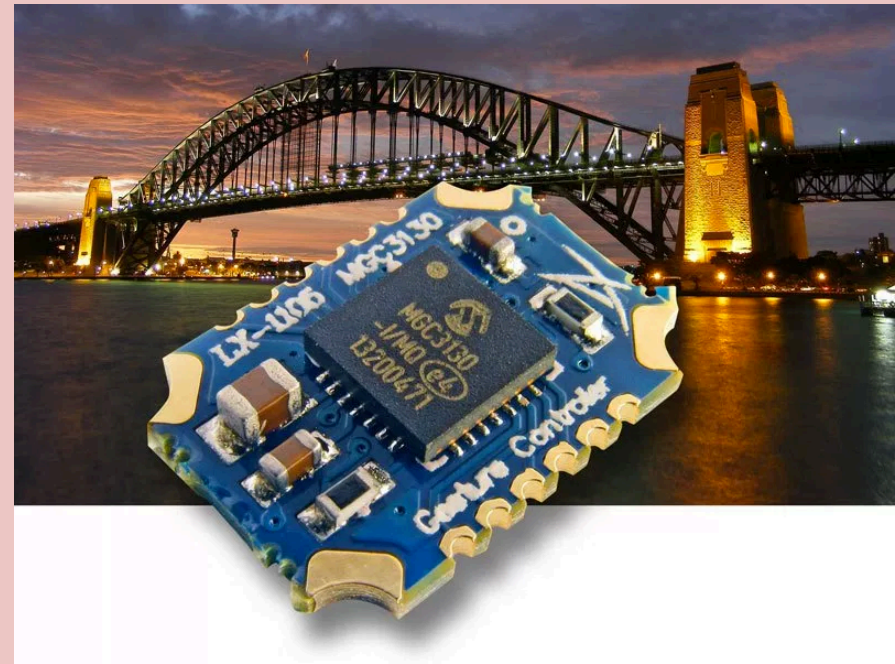
**Smart
Devices**



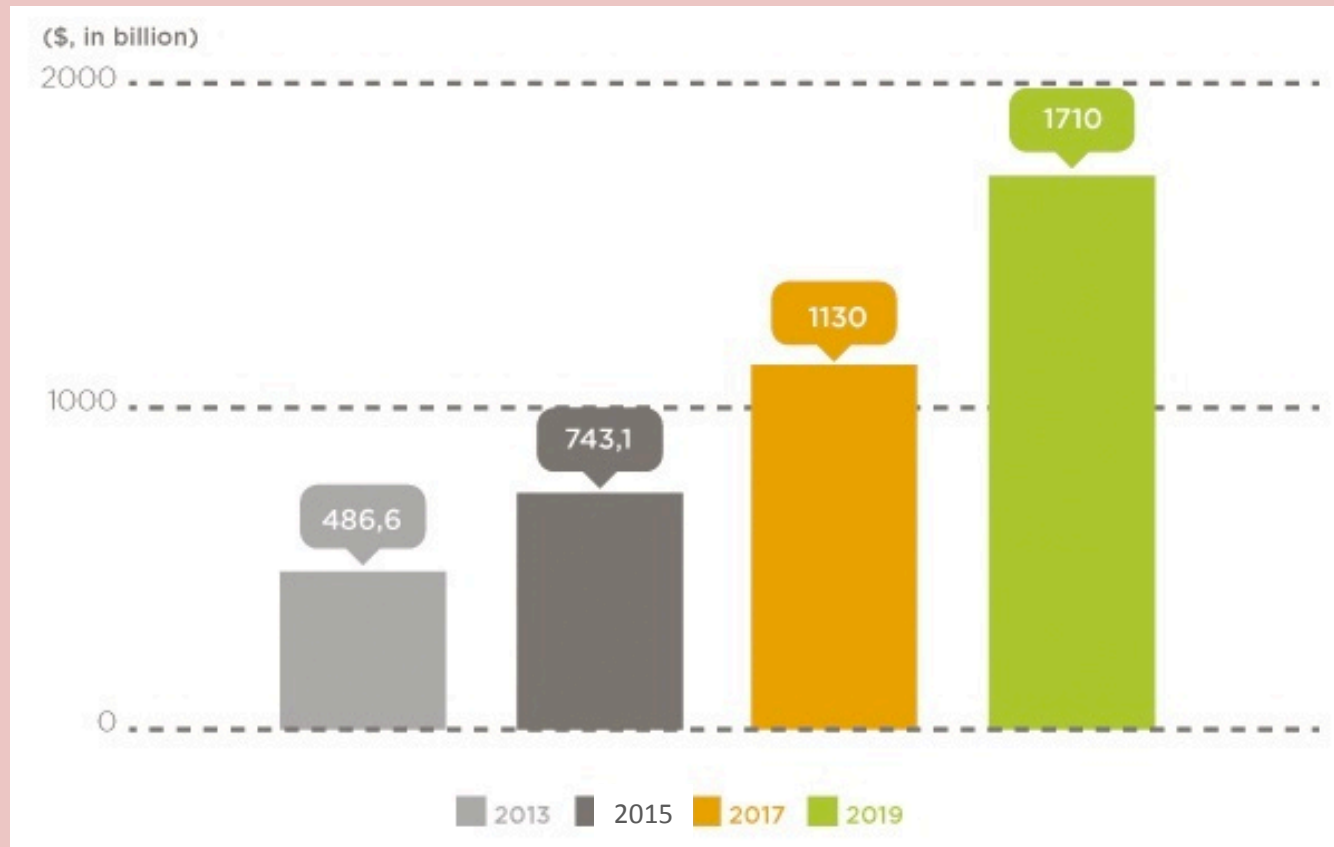
**Health
Fitness**

IoT in Sydney Harbour Bridge

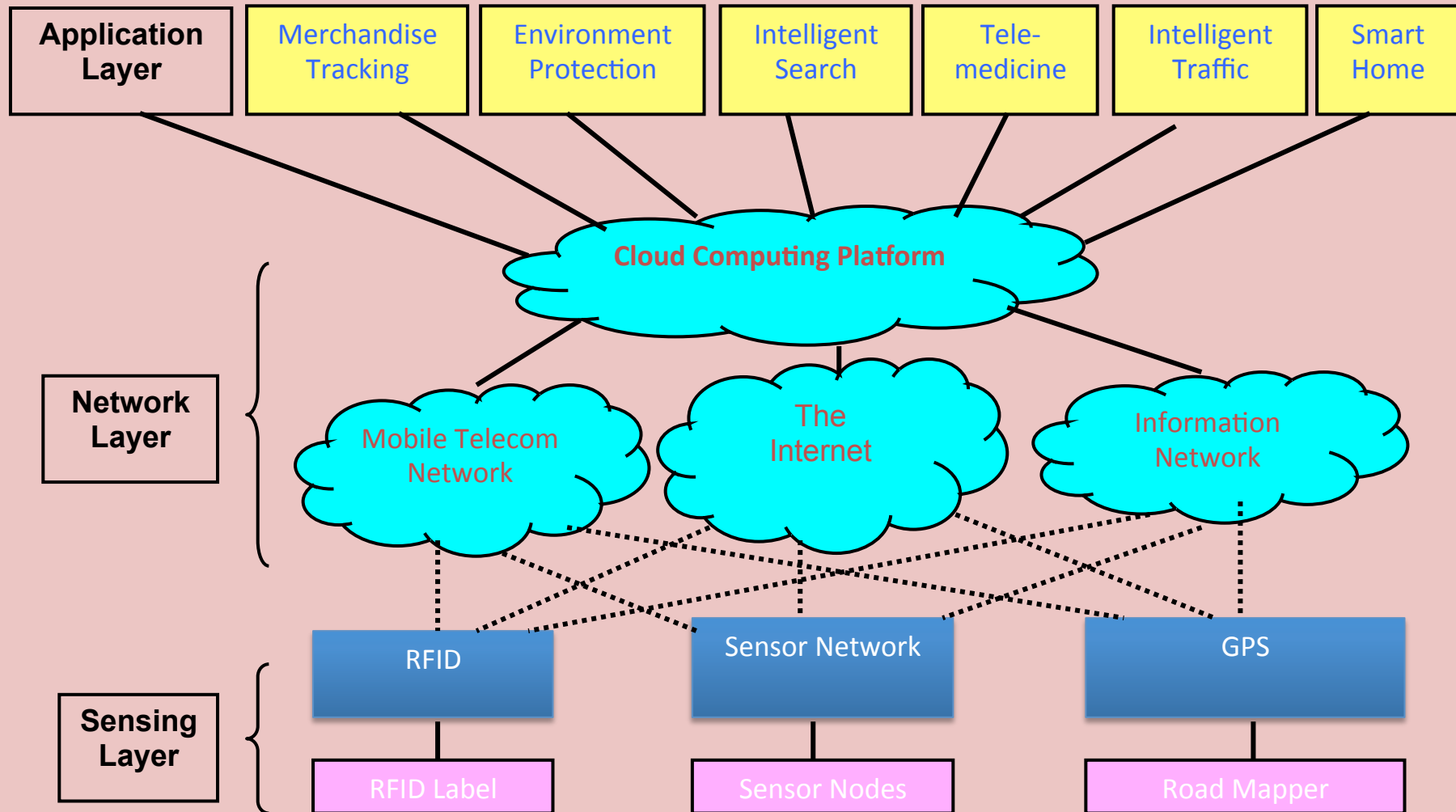
- 1,149m length, 134m height
- 160,000 cars per day
- Moving joints <18cm>
 - 2400 vibration sensors
 - Machine learning for structural monitoring



Global IoT Market



Architecture of the Internet of Things



IoT in Public Clouds

	AWS	Microsoft	IBM	Google	Alibaba
Service	AWS IoT	Azure IoT Hub	IBM Watson IoT	Google IoT	AliCloud IoT
Data Collection	HTTP, WebSockets, MQTT	HTTP, AMQP, MQTT and custom protocols (using protocol gateway project)	MQTT, HTTP	HTTP	HTTP
Security	Link Encryption (TLS), Authentication (SigV4, X.509)	Link Encryption (TLS), Authentication (Per-device with SAS token)	Link Encryption (TLS), Authentication (IBM Cloud SSO), Identity management (LDAP)	Link Encryption (TLS)	Link Encryption (TLS)
Integration	REST APIs	REST APIs	REST and Real-time APIs	REST APIs, gRPC	REST APIs
Data Analytics	Amazon Machine Learning model (Amazon QuickSight)	Stream Analytics, Machine Learning	IBM Bluemix Data Analytics	Cloud Dataflow, BigQuery, Datalab, Dataproc	MaxCompute
Gateway Architecture	Device Gateway (in Cloud)	Azure IoT Gateway (on-premises gateway, beta version)	General Gateway	General Gateway (on-premises)	Cloud Gateway (in Cloud)

Current platforms issues: Not Fully Integrated, No low-latency, and might be Expensive

Service Ecosystem: Fragmented

IoT infrastructure
→ designers need to interact with **many services** from sensors/actuators to data analytic

CLOUD

Cloud: latency

The cloud uses **virtual machines**
→ unnecessary data movement

Data Storage, Analytics

Network: latency, bandwidth and cost

- Large geographical distance → **Higher Latency**
- The aggregated **b/w** of sensors >> network b/w
- Big Data is heavy to move → Higher **Cost and Latency**

Raw data: the size can be **huge** (e.g. camera)

Network & Internet Infrastructure

IoT's

Raw data

Big Data Flow

Raw Data

Feedback

Network of Sensors and Actuators (Physical World)

CYBER - PHYSICAL SYSTEM

Fog Computing

- **The Fog**
 - extends the cloud computing paradigm to the edge of the network,
 - enables a new breed of applications and services
 - an appropriate solution for the applications and services that fold under the umbrella of the IoTs.
- **Benefits**
 - low latency
 - location awareness
 - widespread geographical distribution
 - mobility support
 - the strong presence of streaming and real-time applications
 - heterogeneity

FOG-Engine: Decentralized Hierarchical Big Data Processing on the Edge

CLOUD

Data Processing, Mining, Storage, and Visualization



Data Preprocessing & Analytic

WAN: low b/w (x10MB/s)

FOG

FOG-Engine

FOG

FOG-Engine

FOG-Engine: Data Analytic micro-Engine

Network Access

Raw data



LAN: very high b/w (x1GB/s)

MAN: high b/w (x100MB/s)

Network Access

Raw data

IoT's

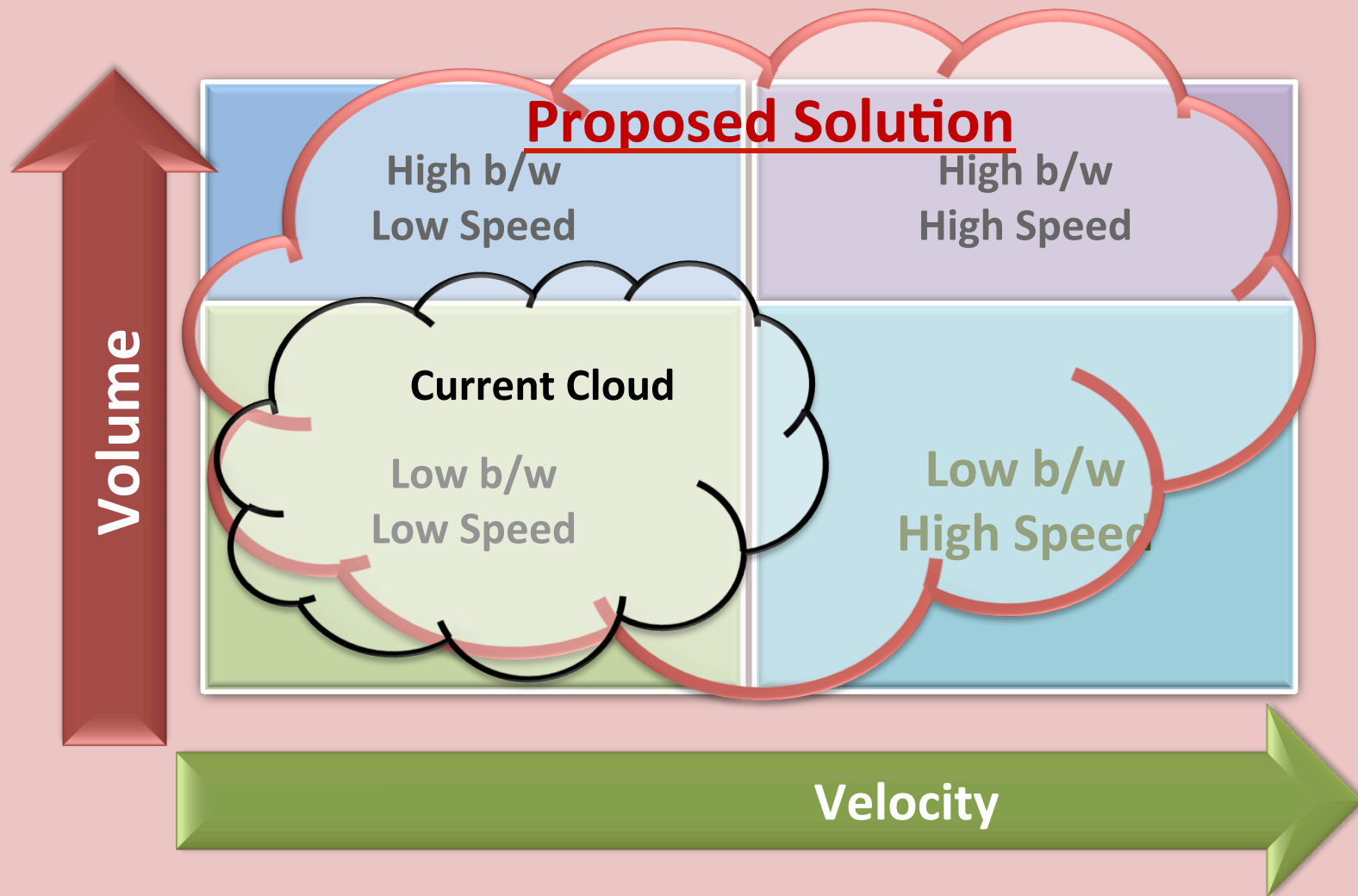


Network of Sensors and Actuators (Physical World)

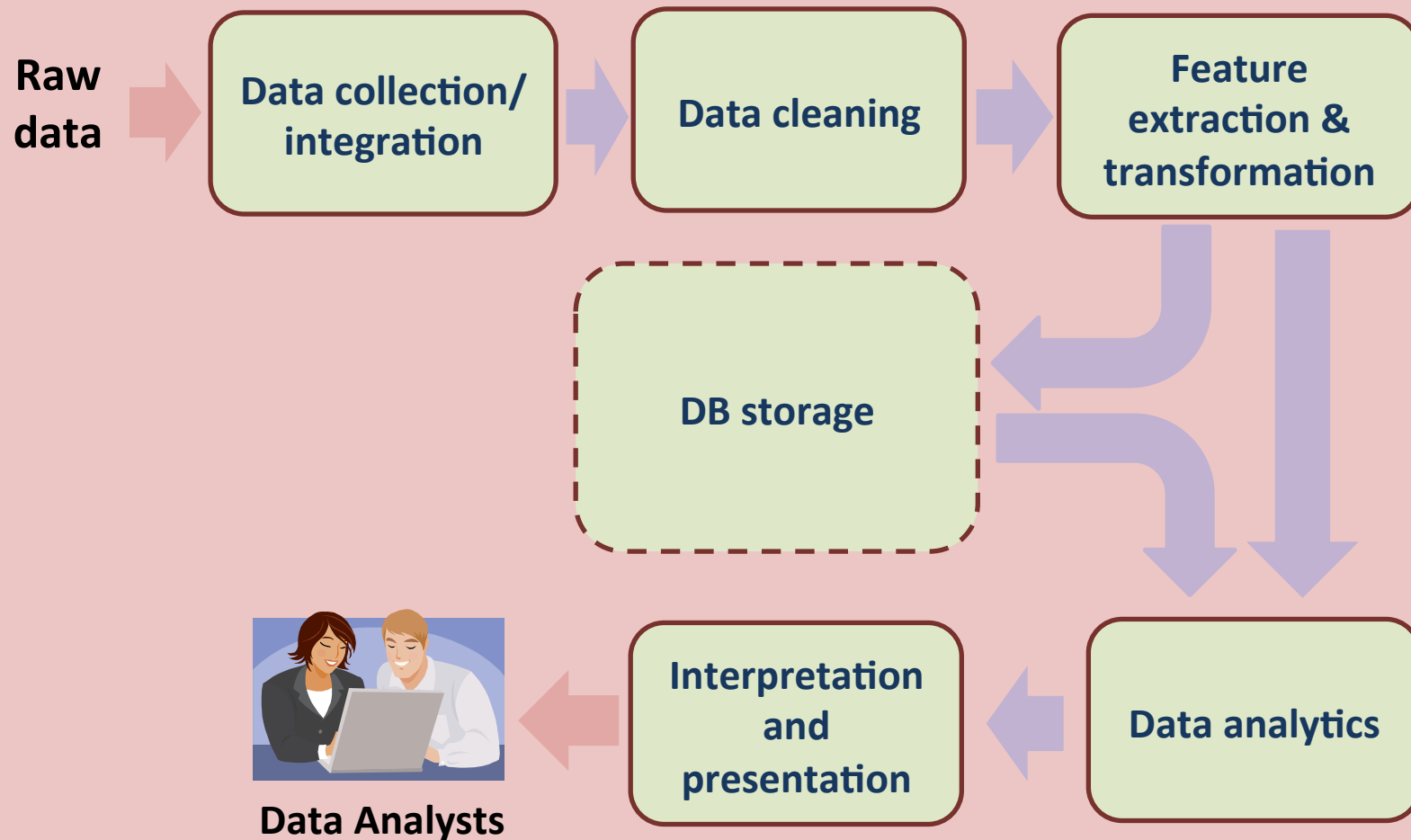
FOG-Engine vs. Cloud

Characteristics	FOG-Engine	Cloud platform
Processing hierarchy	Local data analytics	Global data analytics
Processing fashion	In-stream processing	Batch processing
Computing power	GFLOPS	TFLOPS
Network Latency	Miliseconds	Seconds
Data storage	Gigabytes	Infinite
Data lifetime	Hours/Days	Infinite
Fault-tolerance	High	High
Processing resources	Heterogeneous (e.g. CPU, FPGA)	Homogeneous (Data center)
Versatility	Only exists on demand	Intangible servers
Provisioning	Limited by the number of FOG-engines in the vicinity	Infinite, with latency
Mobility of nodes	May be mobile (e.g. in the car)	None

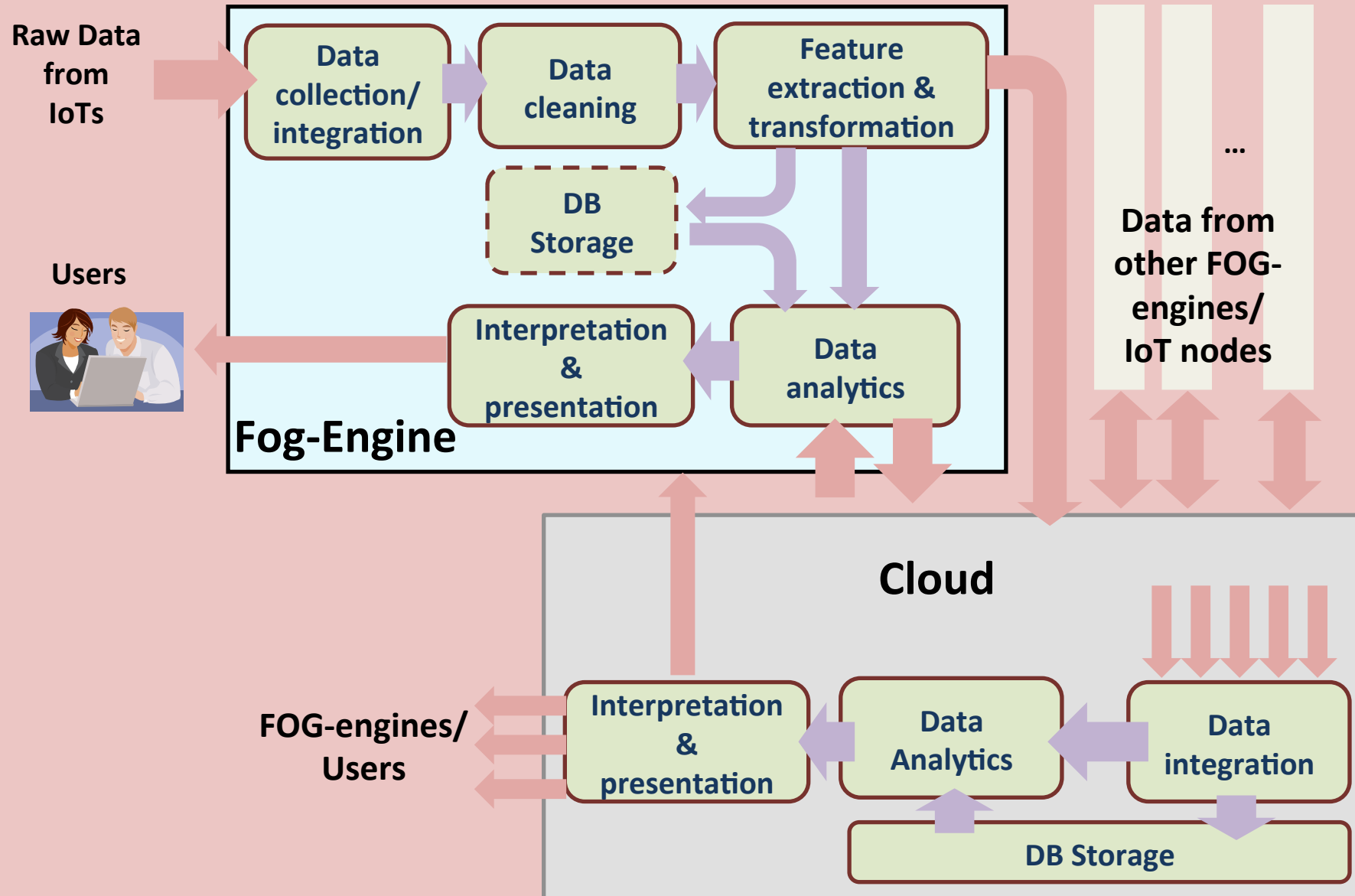
How to Realize Big Data's Vs: Velocity, Volume, ...



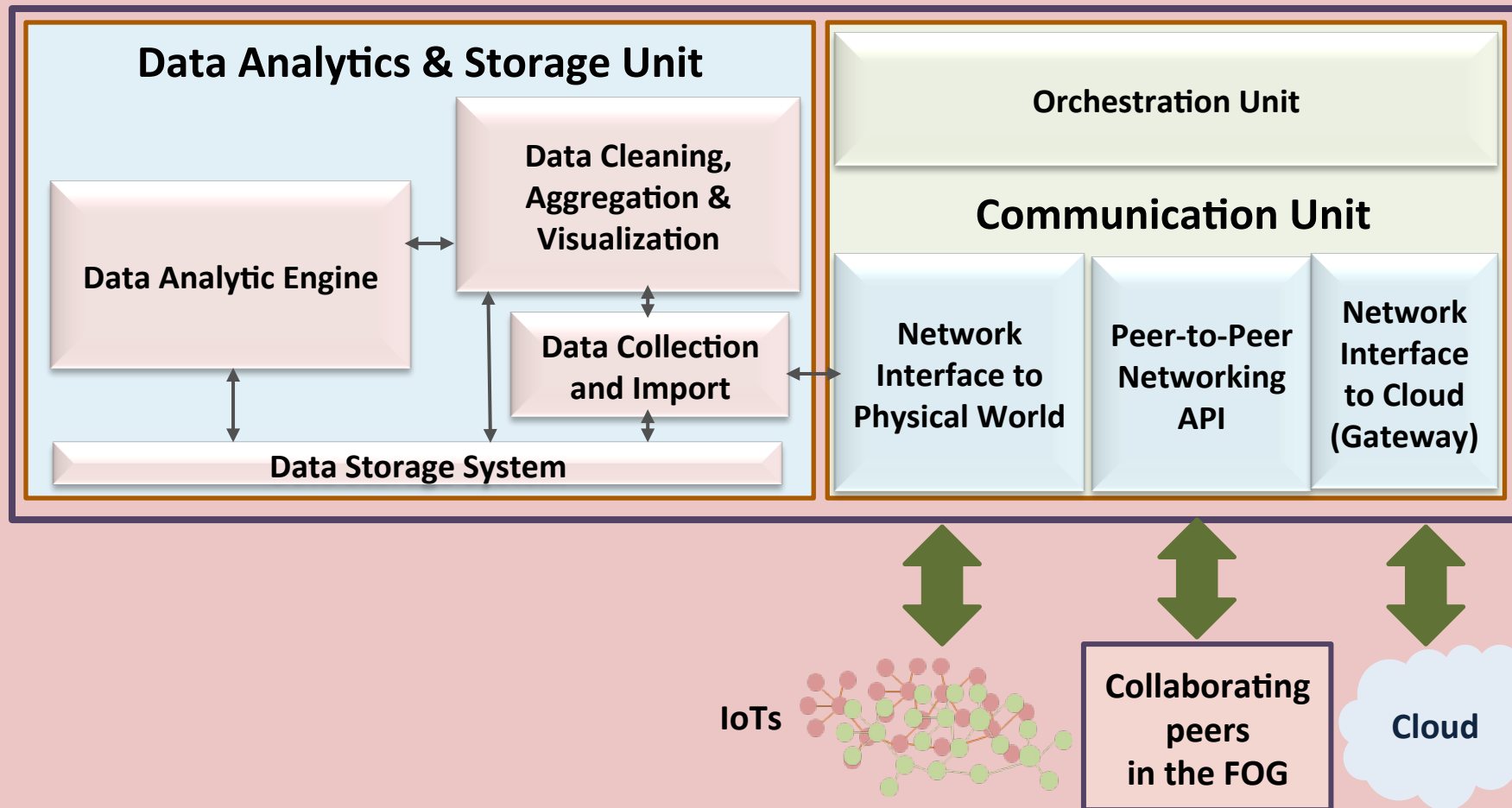
A Typical Data Analytics Flow



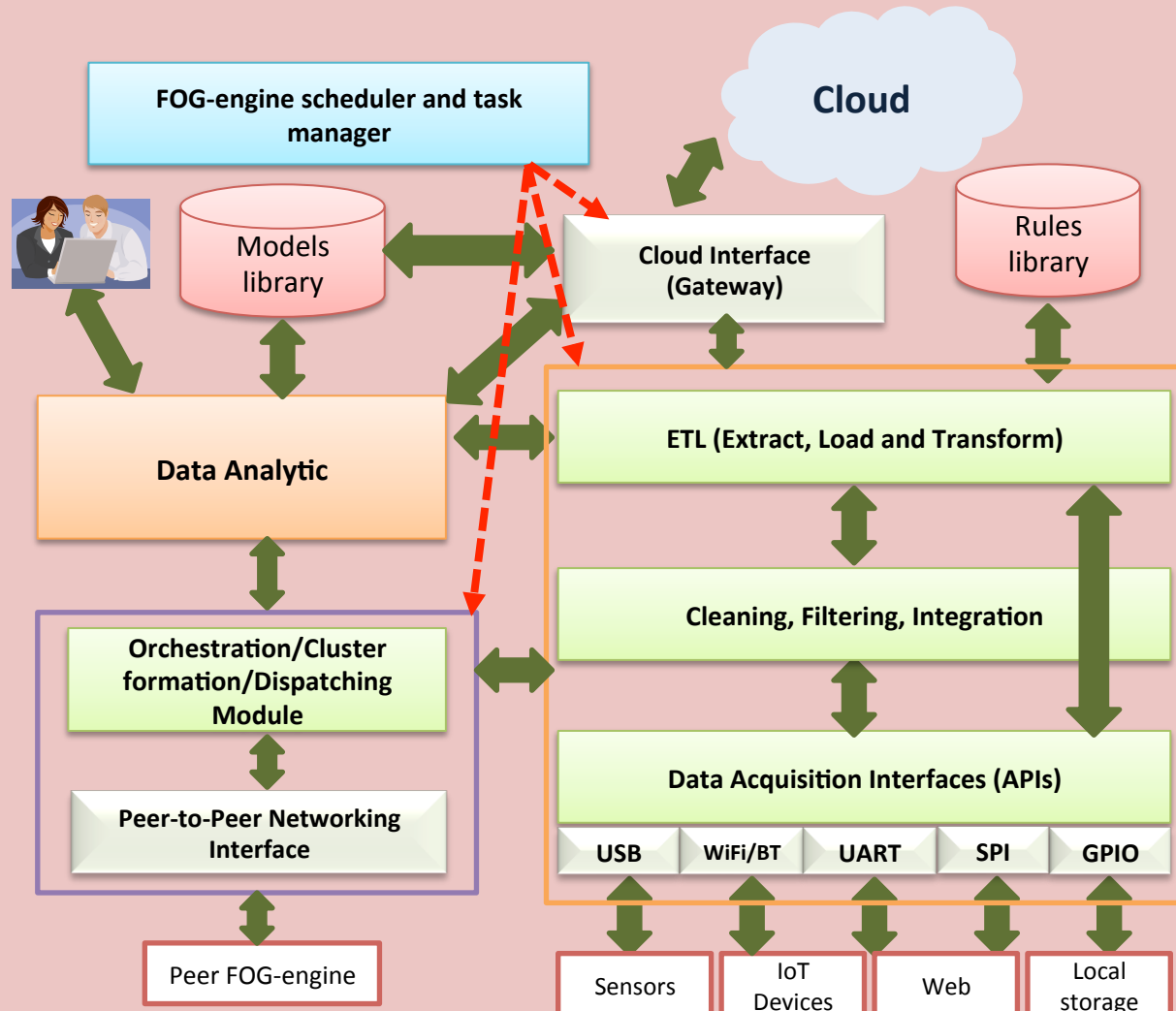
A Modified Data Analytics Flow



General Architecture of FOG-Engine



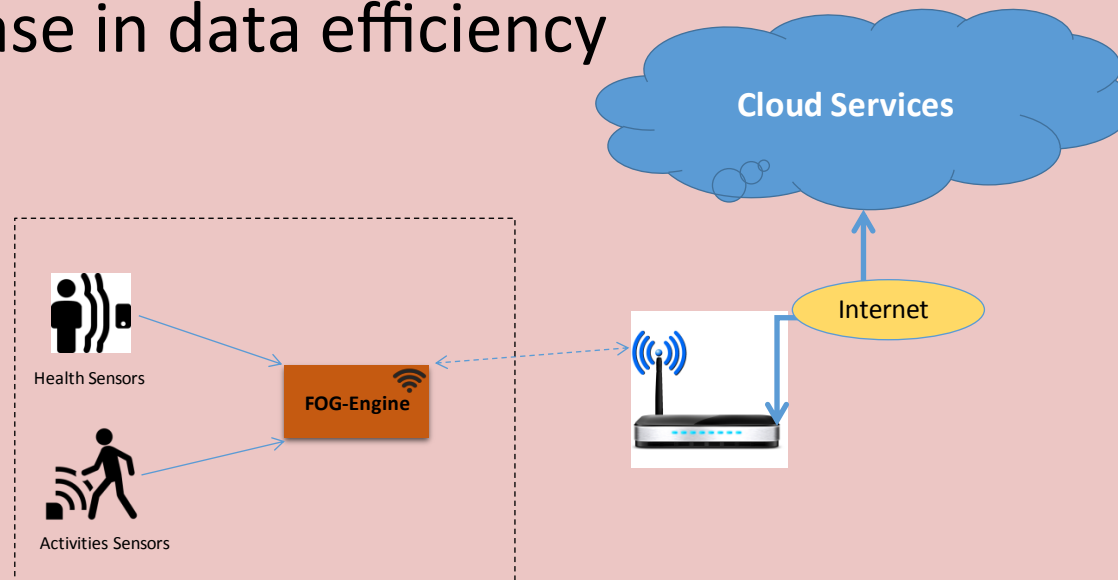
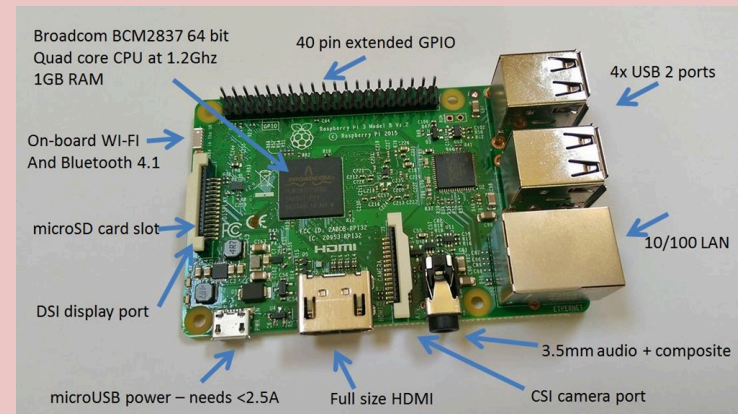
Detailed Architecture of FOG-Engine



USB: Universal serial bus
 BT: Bluetooth
 UART: Universal Asynchronous Receiver/Transmitter
 SPI: Serial Peripheral Interface Bus
 GPIO: General-purpose input/output pins

FOG-Engine Prototyping

- Raspberry Pi 3.0
- Python Library
- Results
 - 40% reduced in data size
 - 60% increase in data efficiency



Smart Nutrition Monitoring System using Fog Computing

OBESITY AND OVERWEIGHT INCREASING WORLDWIDE

37

Percentage of the world's adult population that is overweight or obese

0

Number of countries succeeding in decreasing obesity in last 33 years

14

Percentage of overweight or obese children and adolescents worldwide

62

Percentage of the world's obese living in developing countries

3.4
million

DEATHS CAUSED
by overweight
AND OBESITY



Obesity and overweight
INCREASED
27.5% IN ADULTS
47.1% IN CHILDREN
SINCE 1980

Middle Eastern countries experiencing some of the largest increases in obesity globally:
SAUDI ARABIA, BAHRAIN, EGYPT, KUWAIT, AND PALESTINE



THE US ACCOUNTS FOR **13%** OF THE NUMBER OF OBESE PEOPLE GLOBALLY BUT **JUST 5%** OF THE WORLD'S POPULATION

OBESITY AND OVERWEIGHT CONTRIBUTE TO:



• **CARDIOVASCULAR DISEASE**



• **DIABETES**



• **CANCER**



• **JOINT PAIN**

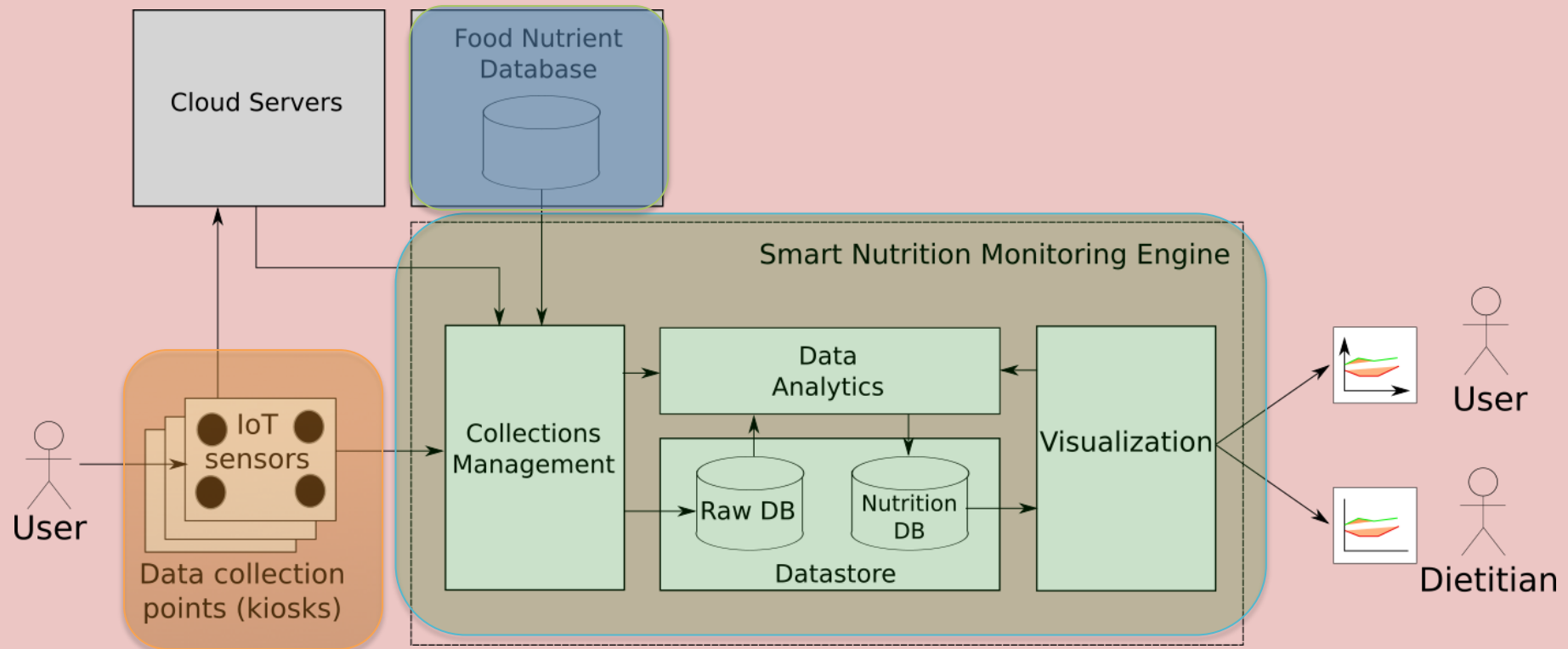
- Overweight and obesity in Adults
 - US: 70%
 - \$200B
 - Australia: 63%
 - \$14B
 - India: ~20%
 - 3rd most obese country in the world

Smart Nutrition Monitoring System

- Project Scope
 - Take away food (50M meals out each week in Australia)
- Non-invasive
 - Minimizing the amount of direct input and actions from users
- High data accuracy and reliability
 - Heterogeneous IoT sensors
- Scalability
 - Cloud and Fog Computing

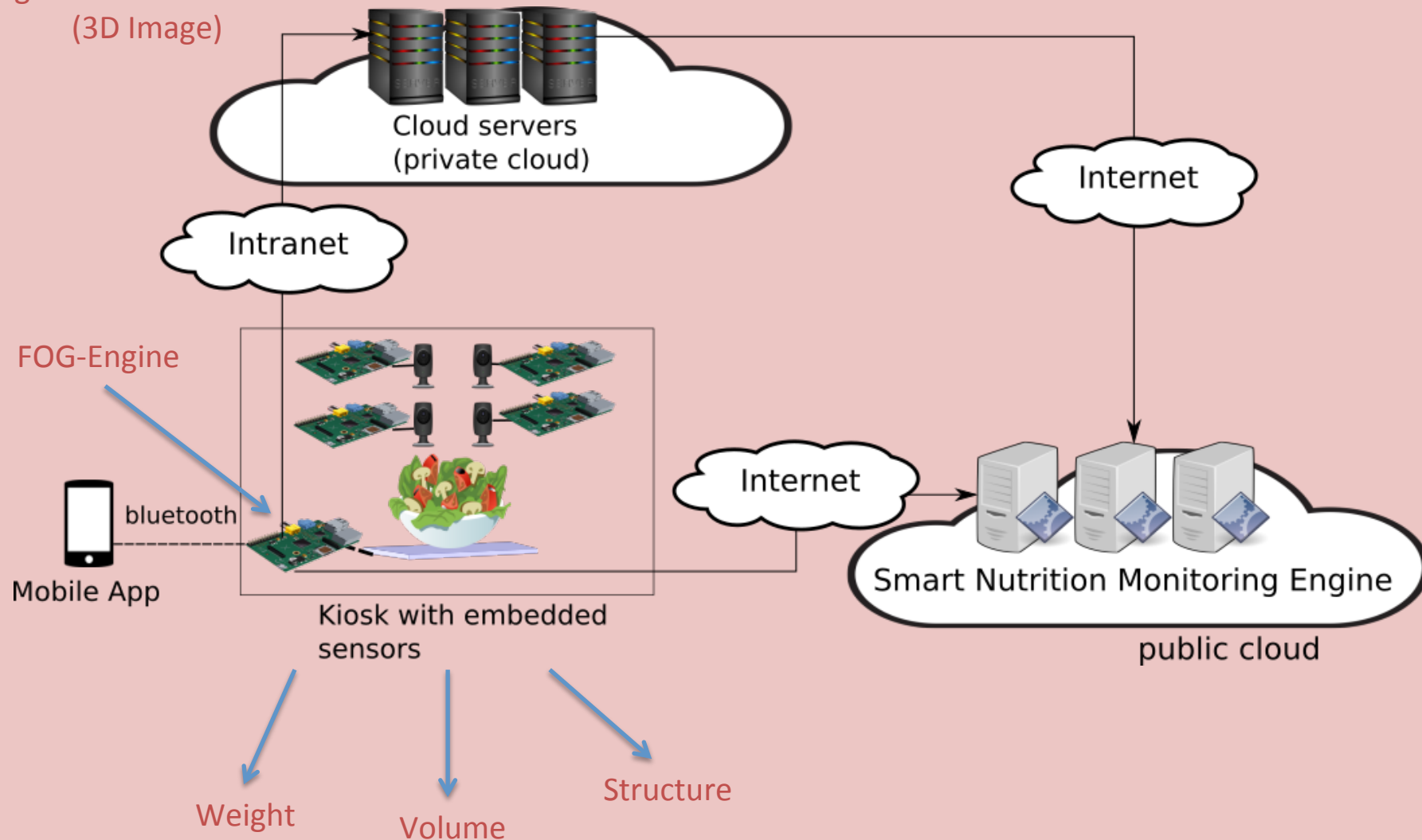


Architecture of Smart Nutrition Monitoring System

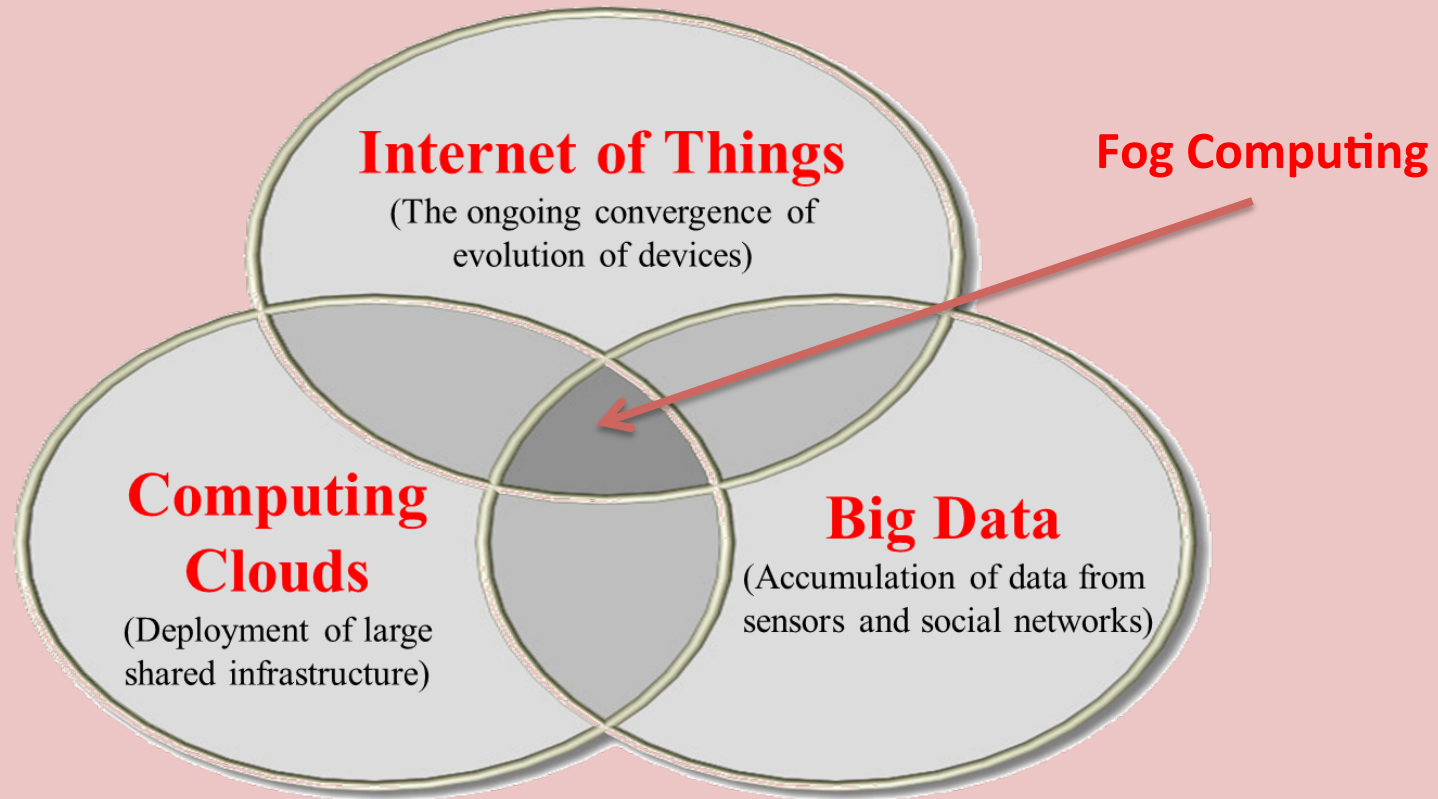


System Prototype

AgiSoft PhotoScan Pro
(3D Image)



Open Issues and Challenges



- Innovative Machine Learning
- Decentralized Resource Scheduling
- Reliability and Power Efficiency
- Information and Security Management

References

- Yogesh Sharma, Bahman Javadi, Weisheng Si, Daniel Sun, “Reliable and Energy Efficient Resource Provisioning and Allocation in Cloud Computing”, 10th IEEE/ACM International Conference on Utility and Cloud Computing (UCC 2017), Austin, US, 2017.
- Bahman Javadi , Rodrigo N. Calheiros, Kenan Matawie , Athula Ginige , Amelia Cook, “Smart Nutrition Monitoring System Using Heterogeneous Internet of Things Platform”, The 10th International Conference Internet and Distributed Computing System (IDCS 2017), Fiji, December 2017.
- Rekha Nachiappan, Bahman Javadi, Rodrigo N. Calheiros, Kenan M. Matawie, “Cloud storage reliability for Big Data applications: A state of the art survey”, Journal of Network and Computer Applications, 97 (2017) 35-47.
- Farhad Mehdipour, Bahman Javadi, Aniket Mahanti, “FOG-engine: Towards Big Data Analytics in the Fog”, The 2nd IEEE International Conference on Big Data Intelligence and Computing (IEEE DataCom 2016), Auckland, New Zealand, August 2016.
- Gubbi, Jayavardhana, Rajkumar Buyya, Slaven Marusic, and Marimuthu Palaniswami. "Internet of Things (IoT): A vision, architectural elements, and future directions." Future generation computer systems 29, no. 7 (2013): 1645-1660.

Thank You



Home Page: <http://staff.scem.uws.edu.au/~bjavadi/>

Email: b.javadi@westernsydney.edu.au