









**Pervasive & Distributed Intelligence** 



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### Outline



01 02 03 04 05

Introduction Objectives Timeline and Research Outcomes Output

### The INNOVATE Team





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Essence: Pervasive & Distributed Intelligence Research Lab

INNOVATE Website: https://sites.google.com/view/mscainnovate/home

https://www.gla.ac.uk/schools/computing/



### Pervasive & Distributed Intelligence

#### **Distributed Intelligence**

Self-organization Algorithms for UxVs Edge-centric Statistical Learning

Funding: H2020/GNFUV



#### **Network-centric Stream Processing**

Delay-Tolerant Data Stream Processing Time-optimized Task Offloading Edge-centric Selective Analytics

Funding: **H2020/MSCA INNOVATE** 

#### **Predictive Computing**

Query-driven Predictive Analytics
Data Relevance: Relevant Data is Big Data
Dataless Explanation & Exploitation of Analytics

Funding: UK EPSRC/CLDS (£3M)



#### Collaboration with Industry & Academia

- Hesso Geneve (CH)
- Repado Ltd (CH)
- inCITES Sarl (LU)
- BMW Group Research (DE)
- BT (UK)

\*

Huawei (CN)

http://www.dcs.gla.ac.uk/essence/

### Research Overview



#### **Query Driven Applications**

Analytics offer the basis for decision making Analytics should be executed on top of multiple data partitions



#### **Queries Management**

Massively allocate queries to distributed datasets Efficiently aggregate multiple query responses Maximize the performance and support time critical applications

#### Management of the Ecosystem

Query Controllers (QCs) manage the incoming queries

Distributed nodes host the data

Query Processors (QPs) execute queries in every node

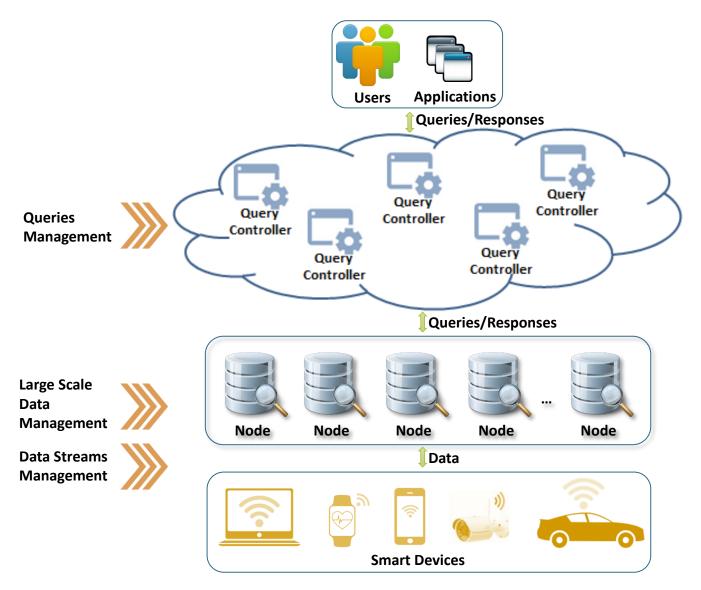


#### **Intelligent Behaviour**

Allocate queries to nodes
Support nodes management
Support data management
Support the behaviour of QCs

### **INNOVATE** Architecture

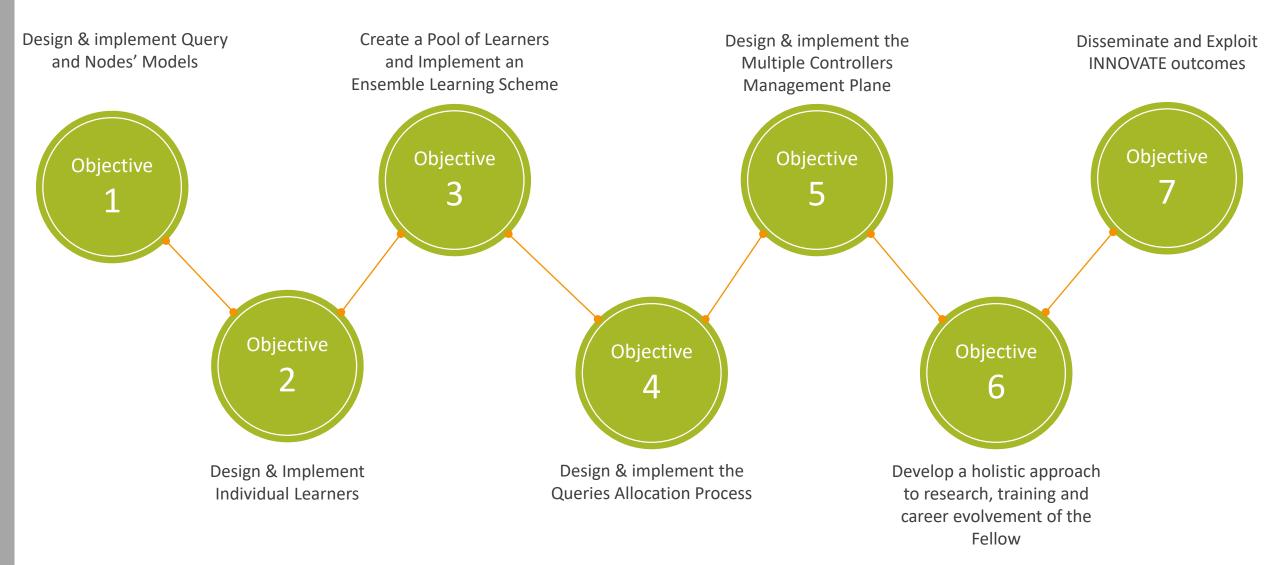




INNOVATE offers intelligent mechanisms for the management of queries, data and distributed nodes

### Research Objectives



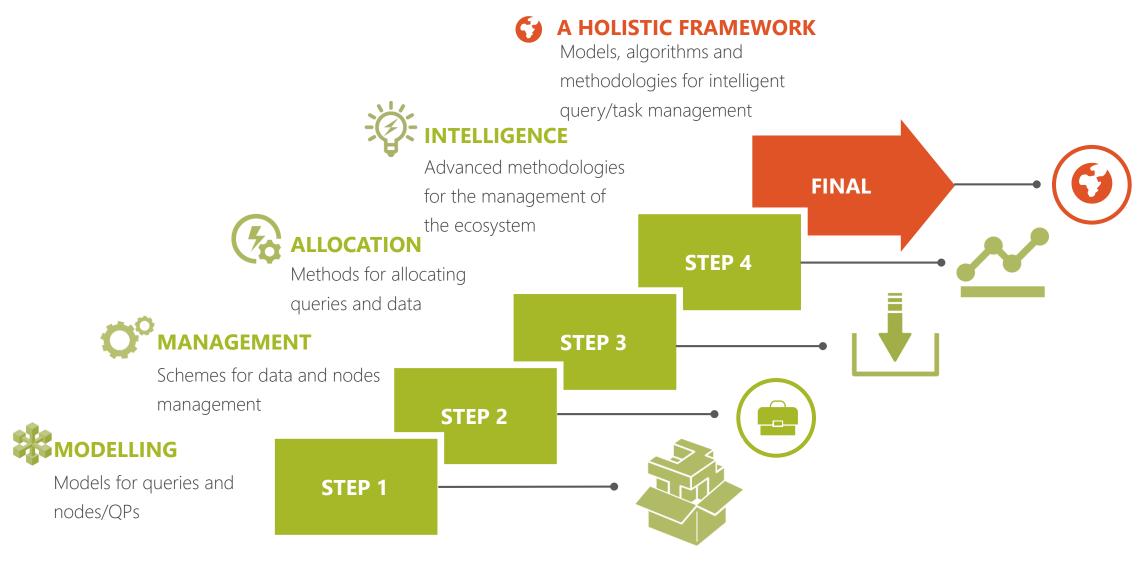


### **INNOVATE** Timeline



### Steps





### Queries and QPs Models





- ✓ We match queries and QPs characterístics
- ✓ Queries
  - ✓ Complexity Class
  - ✓ Deadline
  - ✓ Constraints
- ✓ QPs
  - ✓ Load
  - ✓ Speed of processing
  - ✓ Data present in each node



We propose a model for delivering the complexity class

We propose a *Fuzzy Classification Process* (FCP) The FCP depicts the 'membership' of a query in a pre-defined set of classes



We consider a queue in every node

The size and the rate of the incoming quer

The size and the rate of the incoming queries/tasks affect the load



We adopt IR techniques
We build on an *ensemble similarity scheme*We estimate the number of steps required for executing a query



Based on the contextual information, we build on the *Probability of Allocation* (PoA)

The PoA depicts the 'ability' of a QP to execute a query smoothly

The highest PoA(s) win(s)

### Queries and QPs Models





- ✓ We also focus on additional contextual information
  - ✓ Query/task priority
  - ✓ Available resources
  - ✓ Status of peer nodes
  - ✓ Data present locally and in peers
- ✓ We propose a local decision making mechanism for allocating queries/tasks



We define the query/task contextual vector
We propose a sequential decision making
Every query/task can be executed locally or at
peers



We define the *information vector* for peers We focus on their datasets, the communication cost, the available resources



We propose a *Bayesian classifier* for deciding if a query/task could be executed locally



For selecting the appropriate peer, we adopt a multi-criteria optimization methodology We adopt the VIKOR method

### Multi-criteria Query Allocation





- ✓ We extend our findings taking into consideration:
  - ✓ a more complex decision making scheme
  - √ the 'historical' performance of each node



For deciding a local execution, we adopt a **kNN classifier** 



We provide formulations for estimating the short term and long term load of each node



Peers are selected based on a model retrieved by the *utility theory* 

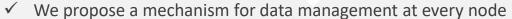


We provide formulations for calculating the probability of a local execution

### Data Management







- ✓ We offer a pre-processing distributed scheme that decides where data should be allocated
- ✓ We focus on the accuracy of data
- ✓ We want to identify and manage the error between the incoming data and
  the available datasets
- ✓ The proposed scheme proactively 'prepares' the data before any query is applied



We define a model that identifies if the incoming data deviate from the ecosystem

If not, data are allocated to the appropriate dataset

If yes, data are rejected



AVDS calculates the probability of a data vector deviates from the ecosystem

We provide formulations for delivering the probability based on a *finite mixture of distributions* 



Our model consists of two parts:

**The accuracy violation detection scheme** (AVDS)

**The Partition identification scheme (PIS)** 



PIS adopts an uncertainty driven decision making
We propose a *Fuzzy Logic controller* for

resulting the appropriate node

### Nodes' Management





- ✓ Nodes convey software and firmware for performing tasks
- ✓ We propose a distributed software update scheme
- ✓ We avoid the disadvantages of legacy, centralized systems
- ✓ Nodes monitor specific KPIs and independently decide when they will initiate the update process



Nodes monitor their internal status (e.g., load, resources)

Nodes monitor the network's performance (e.g.., bandwidth, errors)



We adopt a *time-optimized decision making mechanism*We adopt the principles of the *Optimal Stopping Theory*We build on the expected reward maximization



We consider proportional and non-proportional metrics

We calculate the reward for each metric realization

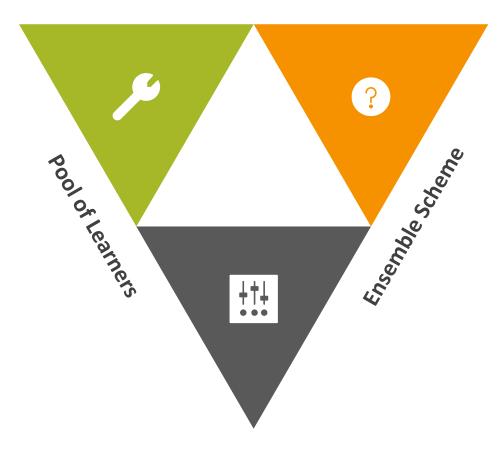


Our model exhibits when to stop the monitoring process and initiate the update

### Ensemble Learning



#### **Individual Learners**



- ✓ We adopt a set of learners
- ✓ They are trained with real and synthetic data
- ✓ We propose a meta-ensemble learning scheme using the following (ensemble) models:
  - ✓ AdaBoost
  - ✓ Stacking
  - ✓ Bagging
- ✓ The (sub-)ensemble schemes are combined with the *One-Over-All (OVA) technique*

#### **Advanced Models**





#### Uncertainty Management

We manage the uncertainty about optimal allocations

We propose the use of Type-2 Fuzzy logic

We combine Fuzzy Logic with a machine learning model



## Automated Knowledge Extraction

We adopt machine learning for generating parts of the Fuzzy Logic model

We automatically deliver the Type-2 Fuzzy Sets and their membership functions

We provide mathematical formulations for the new scheme

#### **Additional Models**

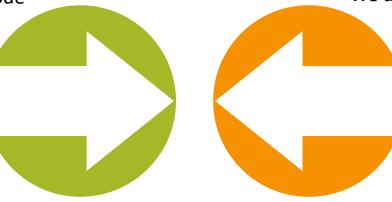


#### A Probabilistic Model for Allocations

We build on our modeling We study the expected load of QPs We propose the concept of the optimal node

#### **Ecosystem Management**

We focus on multiple QCs-nodes/QPs
We apply different types of models
We adopt computational intelligence techniques



#### **Data Synopses Management**

We propose a scheme for sending data synopses to peers

#### **Extension of the Fuzzy Logic Model**

We study the effect of data on the Footprint of Uncertainty (FoU) in Type-2 Systems
We aim to provide a fully automated data driven uncertainty management scheme

### **INNOVATE** Academic Output





**Computing Science** 

#### **Journal Publications**



- Yiannis Kathidjiotis, Kostas Kolomvatsos, Christos Anagnostopoulos, 'Predictive intelligence of reliable analytics in distributed computing environments', Springer Applied Intelligence, 10.1007/s10489-020-01712-5, 2020
- Kostas Kolomvatsos, Christos Anagnostopoulos, 'A probabilistic Model for Assigning Queries at the Edge', Springer Computing, 102, 2020
- Soula, M., Karanika, A., Kolomvatsos, K., Anagnostopoulos, C., Stamoulis, G., 'Intelligent Tasks Allocation at the Edge based on Machine Learning and Bio-Inspired Algorithms', under review, Springer Computing, 2020.
- Kostas Kolomvatsos, Christos Anagnostopoulos, 'An Intelligent Edge-Centric Queries Allocation Scheme based on Ensemble Models', submitted for review in **ACM Transactions of Internet Technology**, 2020
- Kostas Kolomvatsos, Christos Anagnostopoulos, 'Uncertainty-Driven Queries management at the Edge', submitted in Elsevier Fuzzy Sets and Systems, 2020
- \* K. Kolomvatsos, 'A Distributed, Proactive Intelligent Scheme for Securing Quality in Large Scale Data Processing', Springer Computing, 2019
- \* K. Kolomvatsos, 'An Efficient Scheme for Applying Updates in Pervasive Computing Applications', Journal of Parallel and Distributed Computing, Elsevier, 2019
- K. Kolomvatsos, C. Anagnostopoulos, 'Multi-criteria Optimal Task Allocation at the Edge', Elsevier Future Generation Computer Systems, 2019

#### Conferences/Posters/Book Chapters

- A. Karanika. P. Oikonomou, K. Kolomvatsos, C. Anagnostopoulos, 'An Ensemble Interpretable Machine Learning Scheme for Securing Data Quality at the Edge', in International IFIP Cross Domain (CD) Conference for Machine Learning & Knowledge Extraction (MAKE) ,2020
- Karanika, A., Oikonomou, P., Kolomvatsos, K., Loukopoulos, T., 'A Demand-driven, Proactive Tasks Management Model at the Edge', in IEEE International Conference on Fuzzy Systems (FUZZ-IEEE), 2020.
- Ivanov, H., Anagnostopoulos, C., Kolomvatsos, K., 'In-Network Machine Learning Predictive Analytics: A Swarm Intelligence Approach', in G. Mastorakis et al. (eds.), Convergence of Artificial Intelligence and the Internet of Things, Springer, 2020.
- Kostas Kolomvatsos, Christos Anagnostopoulos, 'Edge-Centric Queries Stream Management based on an Ensemble Model', in Springer "Smart Innovation, Systems and Technologies" series volume, 2020
- Karanika, A., Soula, M., Anagnostopoulos, C., Kolomvatsos, K., Stamoulis, G., 'Optimized Analytics Query Allocation at the Edge of the Network', in 12th International Conference on Internet and Distributed Computing Systems, 2019
- E. Aleksandrova, C. Anagnostopoulos, K. Kolomvatsos, 'Machine Learning Model Updates in Edge Computing: An Optimal Stopping Theory Approach', in 18th IEEE International Symposium on Parallel and Distributed Computing, 2019
- 7. K. Kolomvatsos, C. Anagnostopoulos, 'An Edge-Centric Ensemble Scheme for Queries Assignment', in 8th International CIMA Workshop in conjunction with the 30th ICTAI, 2018
- K. Kolomvatsos, C. Anagnostopoulos, 'In-Network Edge Intelligence for Optimal Task Allocation', 30th International Conference on Tools with Artificial *Intelligence*, Nov. 5-7, Volos, Greece, 2018
- K. Kolomvatsos, C. Anagnostopoulos, 'Intelligent Applications over Large-Scale Data Streams', The Scottish Informatics & Computer Science Alliance (SICSA), DemoFest, Edinburgh, Scotland, Nov. 6th, 2018

<sup>\*</sup> Monograph

### **INNOVATE** Training Activities





University

#### **Training Activities**



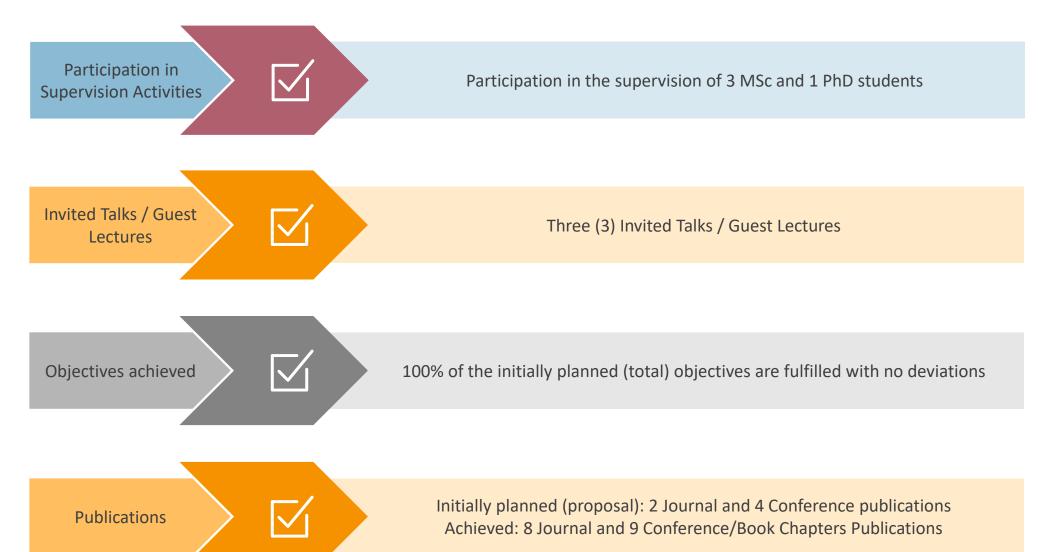
#### **Knowledge Transfer**

- Surviving the Flood of Big Data Streams April 30th, 2018
- Equality & Diversity Essentials May 24th, 2018
- Health, Safety & Wellbeing E-Induction May 31st, 2018
- Research Integrity Training June 1st, 2018
- Coaching Skills For Line Managers (Research & Teaching) June 6th, 2018
- Semantic Intelligent Autonomic Management for Software-Defined Networks June 8th, 2018
- Developing People October 26th, 2018
- Managing Change April 30th, 2019
- Introduction to the General Data Protection Regulation May 1st, 2019
- Information Security Awareness May 1st, 2019
- 11. Managing the Student/Supervisor Relationship June 3<sup>rd</sup>, 2019
- 12. Managing Your Research Data June 3<sup>rd</sup>, 2019
- 13. Research Integrity June 4th, 2019

- K. Kolomvatsos, 'INNOVATE: Pursuing a Successful MSCA Grant & Lessons Learnt', Computing Science Research Away Day 2018, Glasgow, UK.
- K. Kolomvatsos, 'Intelligence over the Edge of the Network', SEEDA-CECNSM Conference, Invited Talk, Kastoria, Greece. September 2018
- K. Kolomvatsos, C. Anagnostopoulos, 'In-Network Decision Making Intelligence for Task Allocation in Edge Computing', 30th International Conference on Tools with Artificial Intelligence, Volos, Greece, 2018
- K. Kolomvatsos, C. Anagnostopoulos, 'An Edge Centric Ensemble Scheme for Queries Assignment', 8th International Workshop on Combinations of Intelligent Methods and Applications, in conjuction with 30th International Conference on Tools with Artificial Intelligence, Volos, Greece, 2018
- K. Kolomvatsos, C. Anagnostopoulos, 'Intelligent Applications over Large-Scale Data Streams', The Scottish Informatics & Computer Science Alliance (SICSA), DemoFest, Edimburgh, Scotland, 2018
- K. Kolomvatsos, 'Data Analysis in Semantic and Distributed Databases', Guest Lecture, Postgraduate MSc Programme – Course: Database Theory & Applications, School of Computing Science, University of Glasgow, Glasgow, UK. November 2018
- 7. K. Kolomvatsos, 'Data Analysis in Semantic and Distributed Databases', Guest Lecture, Postgraduate MSc Programme – Course: Database Theory & Applications, School of Computing Science, University of Glasgow, Glasgow, UK. December 2019

### **INNOVATE** in Numbers





# Thank You

Essence: Pervasive & Distributed Intelligence

http://www.dcs.gla.ac.uk/essence/