

Design of IoT Centric Algorithms: Selection & Performance Assessment

Soumya Banerjee ^{1,2}

Samia Bouzefrane¹

¹Le Conservatoire National des Arts et Métiers (France),²Birla Institute of Technology, (India)

INTRODUCTION **IoT Centric Algorithm** Challenges **Current work** Internet of Things (IoT) **Definitions :** · Uncertainity in Design of IoT layers with Phase I: Identification of immediate diversified components from 3rd parties Internet of Things (IoT) is a new paradigm neighbouring components in IoT has become an that refers to a world-wide network of design space. emerging technology for · Effective Scheduling time interconnected physical things using sensors based standardised communication protocols to Phase II: Compute the position between applications. However, **Design Optimization Objectives:** due to diversified cyber provide human useful services such as IoT objects, locations, time · QoS, load balancing, self organized or physical systems, the personal health care and green energy stamps, relation and community adaptive learning of similar design structure of IoT could be monitoring. At present, many third party service providers are providing a large · Maximizing the benefits Phase III: Evaluate position of more versatile. This new phenomena drives the number of IoT services. Selection of IoT components Optimally (Hard **Optimization strategy:** services to users, based on their owned problem) with dynamic weights different parameters in Designing of IoT algorithm is hard amount them. schema, objects, has become very crucial for the combinatorial problem, hence to use Machine learning success of IoT. Hence, selection of which must be optimal based models to train better design suitable algorithms also is a prime factor. Phase IV: To find out dynamic connectivity from the design point of similar perspectives towards IoT Especially selection of graph-based (Lapacian Matrix) view. This presentation is structures an initiative to introduce algorithms is popular to IoT applications. We analyse a 2nd order hyper-graph Deployment through distributed formal model Phase V: of connected graph for the 2nd order optimization for learning model for IoT systems, in which each Require artefact of IoT. It Is also hyper-edge connects users, objects, and maximum IoT desian Task proposed to develop a services ۲ parameters. Laver learning paradigm, to feed the parameters and Parameters : Effort is given to foster an automated off-line of · Users, objects, and services can be component IoT. and learning based scheme which finally Hence, as a tool, the modelled as a tripartite graph with Interface Laver can be converted as a generic tool of IoT and system could identify hyperedaes. cyber space design at least to satisfy those similar parameters Any IoT system can be defined as a maximum numbers of design constraints. and will contribute in topple that describes the users U, optimal design. services S, objects O, and the ternary Interaction Layer relation between them. distributed 2nd order Introducing IoT objects : Timestam : Location \triangle optimization for learning IoT objects, CONTACT timestamps and locations. P Semantic service interfaces (User requirements umya Banerjee Possibility of Weighted Hypergraph for Semantically similar -->: Semantically dependent REFERENCES Usage relation interaction layer. Samia Bouzefrane [1]. Mustafa Mısır, Michèle Sebag Alors: An Figure 1. Working scheme algorithm recommender system Artificial Intelligence, Elsevier, 2017, 244, pp.291-Weight value is uncertain i ntercation model of IoT 314. [2].Manuel López-Ibáñe Thomas Stützle Automated affline design of algorithms IoT object (represented as Timestamp (represented as Location (represented as Proceeding GECCO '17 Proceedings of Location (represented as Relation (represented as a l Evolutionary Genetic and the 1. Com munity (repres Computation Conference Companion,

 $\begin{array}{c} \mathsf{o}_{8} & \mathsf{o}_{7} & \mathsf{o}_{8} \\ \mathsf{w}(r_{2}) = 3 \\ \mathsf{o}_{8} & \mathsf{o}_{7} & \mathsf{w}(r_{3}) = 4 \\ \mathsf{o}_{8} & \mathsf{o}_{1} & \mathsf{o}_{1} & \mathsf{o}_{1} \\ \mathsf{w}(r_{3}) = 3 \\ \mathsf{o}_{6} & \mathsf{o}_{6} & \mathsf{o}_{6} \\ \mathsf{o}_{8} & \mathsf{c}_{1} & \mathsf{c}_{1} \\ \mathsf{c}_{8} & \mathsf{c}_{1} & \mathsf{c}_{1} \\ \mathsf{c}_{9} & \mathsf{c}_{0} \\ \mathsf{c}_{9} & \mathsf{c}_{0} \\ \mathsf{c}_{9} & \mathsf{c}_{0} \\ \mathsf{c}_{9} & \mathsf{c}_{0} \\ \mathsf{c}_{9} & \mathsf{c}_{1} \\ \mathsf{c}_{9} & \mathsf{c}_{0} \\ \mathsf{c}_{9} & \mathsf{c}_{1} \\ \mathsf{c}_{9} & \mathsf{c}_{0} \\ \mathsf{c}_{9} & \mathsf{c}_{1} \\ \mathsf{c}_{9} & \mathsf{c}_{0} \\ \mathsf{c}_{1} & \mathsf{c}_{1} \\ \mathsf{c}_{1} & \mathsf{c}_{1} \\ \mathsf{c}_{1} & \mathsf{c}_{2} \\ \mathsf{c}_{2} \\ \mathsf{c}_{1} & \mathsf{c}_{2} \\ \mathsf{c}_{2} \\ \mathsf{c}_{1} & \mathsf{c}_{2} \\ \mathsf{c}_{2} \\ \mathsf{c}_{1} \\ \mathsf{c}_{2} \\ \mathsf{c}$

Figure 2. Interaction layer



15 - 19, 2017.[3]. Jimmy Ba, et. AL Distributed Secondorder Optimization using Kronecker-Factoored Approximations, iclr 2017.

pp,1038-1065, Berlin, Germany - July

- [4]. I. Mashal, O. Alsaryrah, T.-Y. Chung, C.-Z. Yang, W.-H. Kuo, and D. P. Agrawal, "Choices for interaction with things on Internet and underlying issues," Ad Hoc Networks, vol. 28, pp. 68-90, 5// 2015.
- [5]. Perera, A. Zaslavsky, C. H. Liu, M. Compton, P. Christen, and D. Georgakopoulos, "Sensor search techniques for Sensing as a Service architecture for the Internet of Things," IEEE Sensors Journal, vol. 14(2), 2014, pp. 406-420.