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# DEEP REINFORCEMENT LEARNING-BASED NETWORK SLICING

Yassine Hadjadj-Aoul, and Abdelkader Outtagarts

Inria – Nokia Bell Labs common lab research action

« Analytics and machine learning for dynamic management of mobile virtualised network and service optimisation »

# TEAM WORKING ON SERVICE PLACEMENT

#### **INRIA**

- Yassine Hadjadj-Aoul (Associate prof.)
- Gerardo Rubino (Research director)
- Eric Rutten (Researcher)
- Anouar Rkhami (Phd student)
- Pham Tran Anh Quang (former Post-doc)

#### Nokia

- Abdelkader Outtagarts (Senior researcher)

# PLAN

- Introduction
- Challenges related to network slicing
- On using Deep Reinfocement Learning for network slicing
- Conclusions

## KEY CONCERNS FOR NETWORK OPERATORS

The ability to support current, emerging and future services

**Examples:** Automotive, mMTC, IoT/Factory automation, e-health

Ability to lease infrastructure to third parties without compromising the network and its efficiency

Consequence: Multiple stakeholders in a network

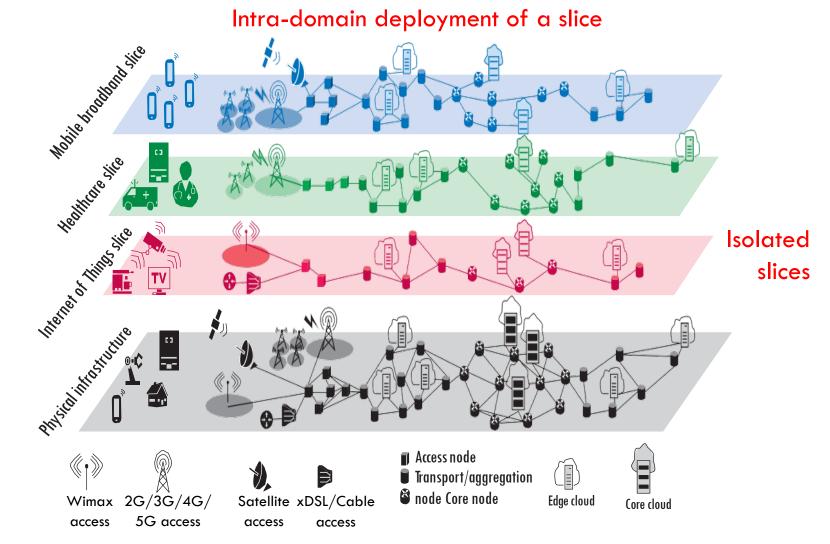
Automation is big deal for Network Operators

Objective: Zero touch networks

# WHY ARE OPERATORS INTERESTED IN NETWORK

SLICING?

Network slicing is seen as one of the most important technologies to meet Network Operators' expectations



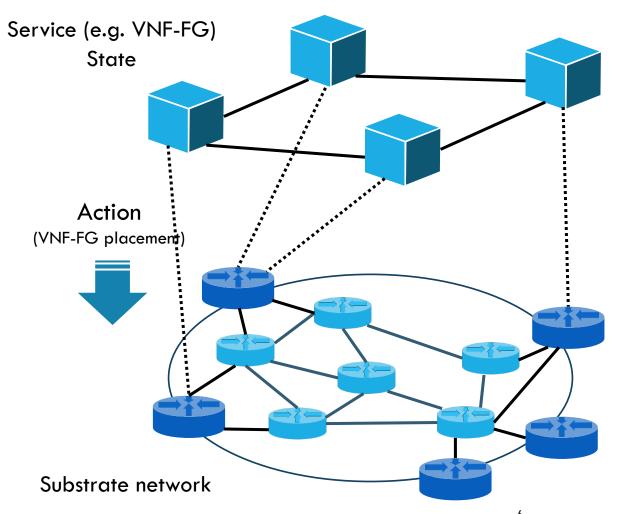
# IN PRACTICE, WHAT DOES SLICING A NETWORK CONSIST OF?

#### Placement of services in a VNF-FG form

- Involves not only the placement of VNFs but also addressing a routing problem
  - Addressing the VNFs placement then the routing ... or simultanuously
- Need to consider several metrics (QoS requirements)
- Intra-domain or multi-domain placement

Extremely large number of possibilities of placement (very large action space)

 Difficulty in finding an optimal placement, excepting for very small instances (i.e. NP-hard problem)



## LIMITS OF EXISTING WORK

- Optimization problems (i.e. MILP) have the limitation of not always being applicable in a real context, given the latency of resolution or unsuitability in a real context
- Meta-heuristics are slow, require a realistic simulation environment ... and there is no learning
- Heuristics are very fast but present some difficulties in finding good solutions (i.e. stuck on local minimums)
- Al-based approaches: as learning progresses, performance becomes more and more important.
  - Safety problem

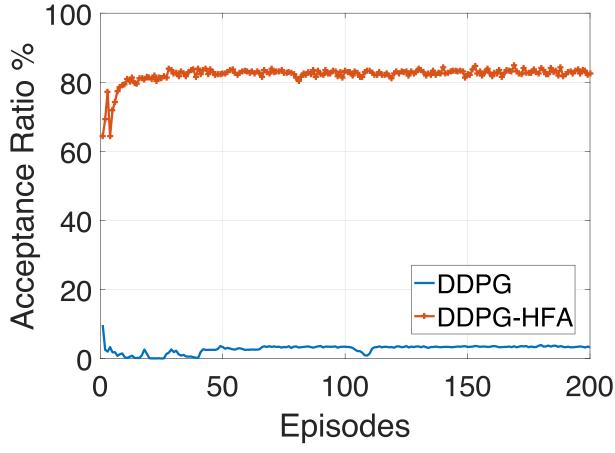
#### Objective

Exploring the potential of deep learning for safe placement of VNF-FG

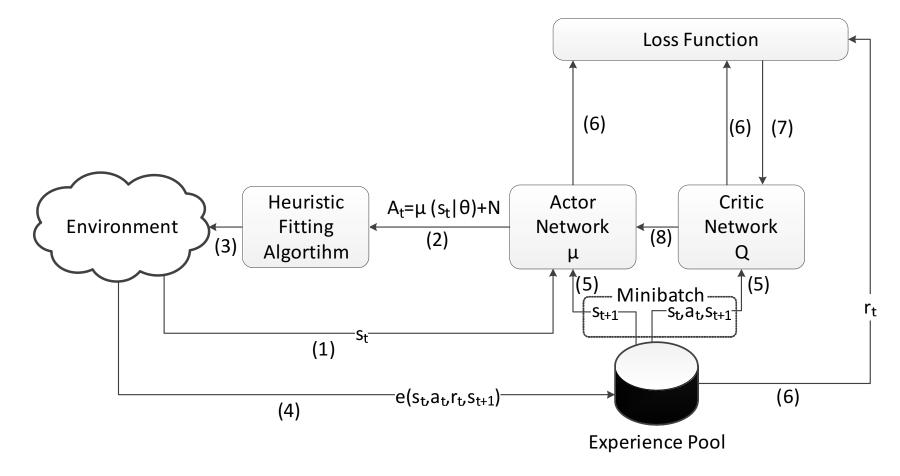
### APPLICATION OF DRL FOR VNF-FG PLACEMENT

- The vanilla DDPG is not suitable for very large-scale discrete action space
- In VNF-FG embedding problem, there are constraints of resources such that some discrete actions are not feasible

Proposing to add to DDPG a
Heuristic Fitting Algorithm (HFA)



# DDPG-HFA



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## **ADVANTAGE**

#### FFPG-HFA allows improving the reward

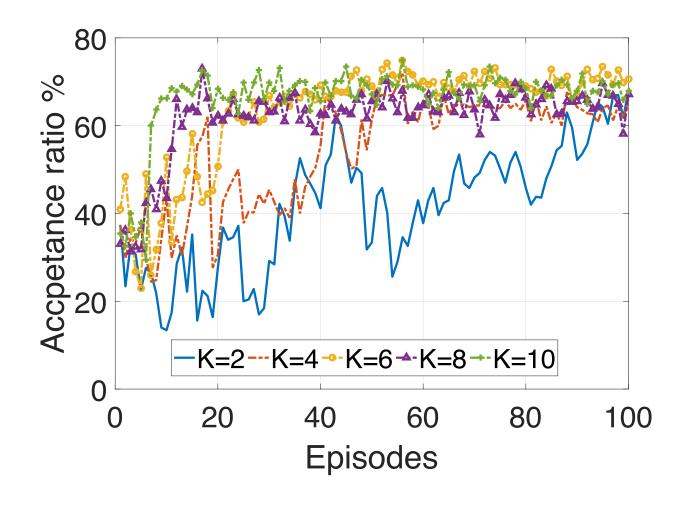
- Convergence to classical heuristic performance is immediate
  - No random behavior at the begining of the learning

Learning is slow, but once learned, execution is immediate

The proposed solution allows

- Improves the heuristic performance
- Makes the placement safer ...

# WHY GO DEEP?



Impact of the number of fully connected Layers

### IMPROVING EXPLORATION WITH EEDDPG

Noise is added to the proto-action to create H noisy actions.

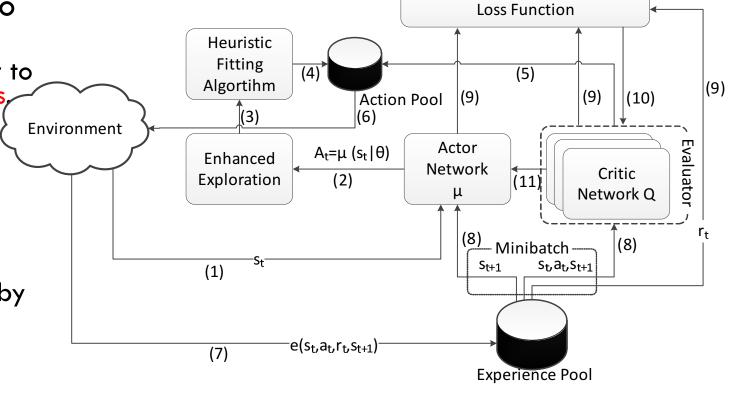
 Each proto action is fed into HFA in order to determine the feasible allocation of VNFs,

#### **Evaluation:**

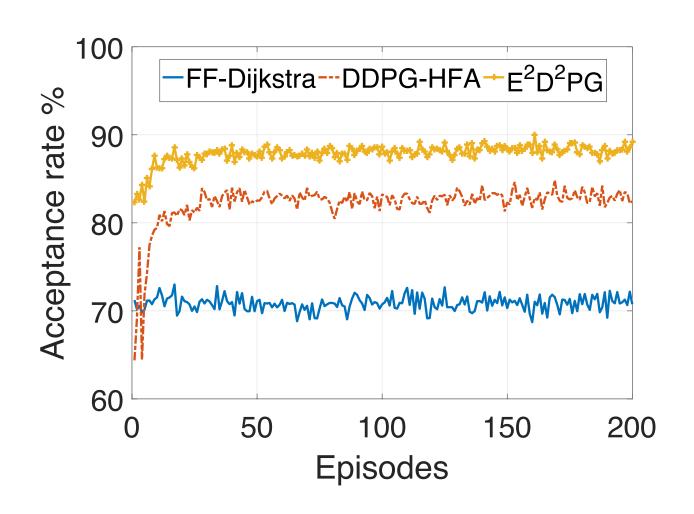
- Multi-critic
  - Different nets due to the arbitrary initialization
  - The best action identified (best Q value) by the evaluator will be executed

#### **Updates of actor:**

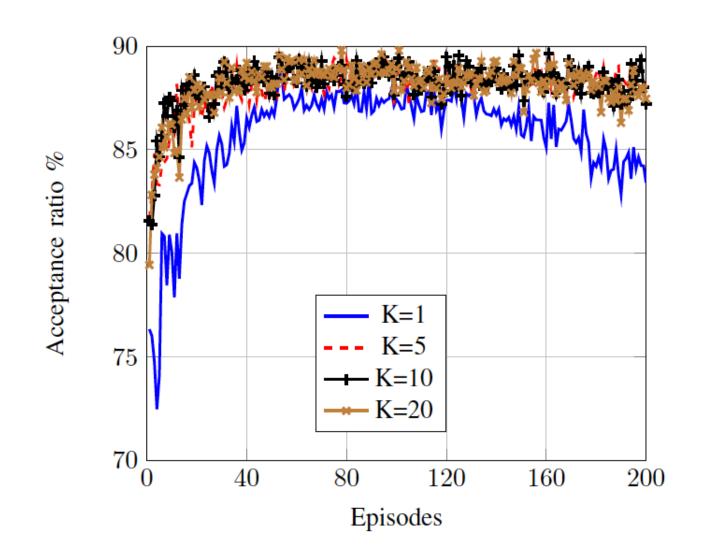
Best critic network



### RESULTS ON THE APPLICATION OF EEDDPG



# IMPACT OF THE NUMBER OF CRITIC NETS



## **ADVANTAGE**

Improving exploration allows to discover better action

which is very efficient in congested systems

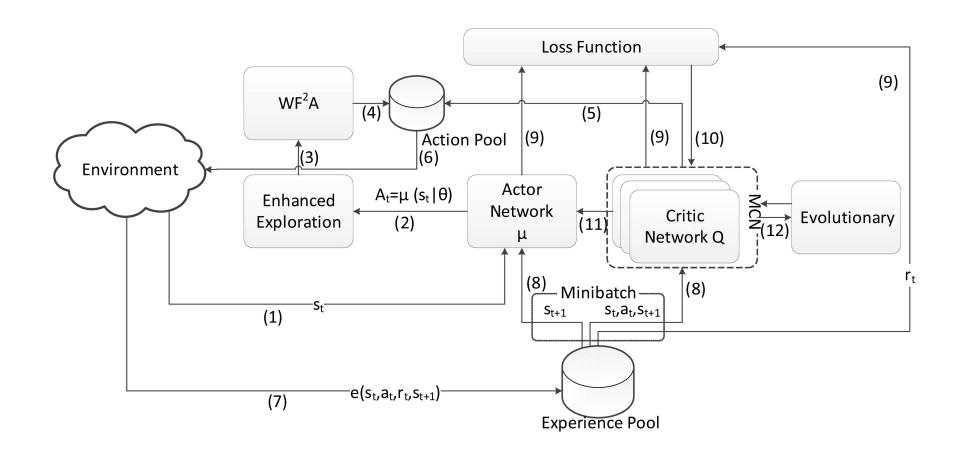
With learning techniques, there is sometimes a degradation of performance

• the use of several CRITIC networks brings more stability

The complexity is not necessarily much more important

 the back-propagation is only applied on one CRITIC network, the best ranked one

## **EVOLUTIONARY EEDDPG**



# **ADVANTAGE**

Discover new and much more interesting states/actions

• Effective in highly congested systems

## CONCLUSIONS

DRL are very efficient in addressing the targeted issue

Lack of explainability

 But combining a heuristic with DRL allows having some guarantees in the placement quality

Efforts are still needed ...

- Ongoing work on using Graph neural nets for services placement
- Combining control theory and DRL for safer placement strategies.

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