

5G PPP White Paper “AI/ML for Networks”

5G PPP Technology Board Workshop
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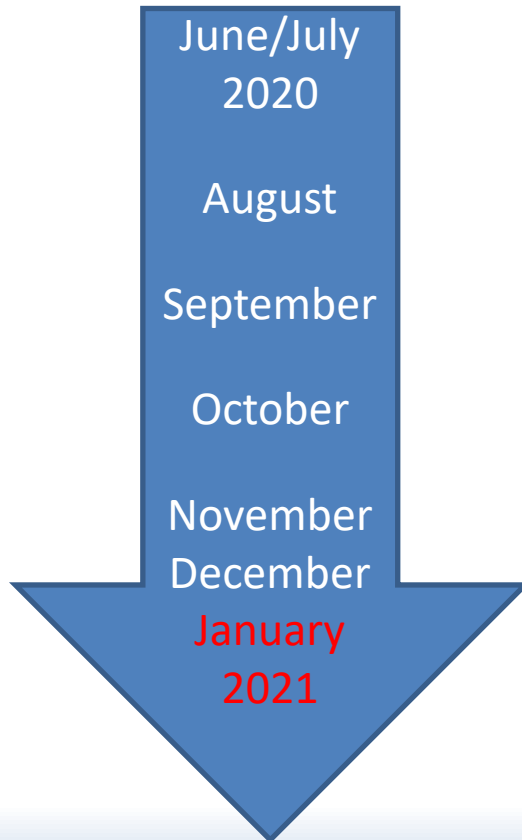
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Scope of the White Paper

AI/ML – Point of Interests from the Communications Network Perspective

- Identification of areas of AI/ML applications optimizing **operation and design** of the future 5/6-G networks
 - addressing different network aspects, communications layers, and specific functions
 - possible cross-optimization opportunities
 - elaborating AI/ML methods from various areas of science and live, to identify the most suitable solutions for the networks
 - including discussion on availability and (re-)usability of data sets
 - management of appropriate AI/ML models and processes
- practical implementation issues and overall impact and perspectives



Time Plan

- Initiative and announcement
- Collection of interests to participate and initial ideas on potential contributions
- Launch by editorial group, mailing list and repository, clarification of expected contributions
- The main contribution/writing period
- **Finalization**
 - First round of consolidation finished
 - First complete draft to be released soon
 - Fine tuning of contribution – request for clarification / last inputs
 - Final draft -> January 2021

Involved 5G PPP Projects

- 5G-PPP / 5G IA
- SliceNet (phase 2)
- IoRL
- 5Growth
- 5G-VINNI
- LOCUS
- 5G-TOURS
- 5G-HEARTH
- 5GZORRO
- 5GENESIS
- TERAWAY
- MonB5G
- 5G-COMplete
- INSPIRE-5Gplus
- 5G SOLUTIONS
- 5G DRONE
- 5G-CLARITY (WP editors)
- ARIADNE (WP editors)

Initial Structure

- Motivation
- Optimization issues requiring AI/ML intervention
 - Network planning
 - Network operation and optimization
 - Network control functions
 - Specific network and service functions
 - Network transmission (low-layer) functions
- AI/ML methods and requirements for network optimization
 - Suitable AI/ML methods
 - Data-sets – availability and usage
 - Keeping control on networks
 - Cross-optimization
 - Management of ML models / functions
- Evolution of ecosystem for AI/ML optimized 5G

- 1) Introduction
- 2) **AI/ML Methods for Network Optimization – An Overview**
 - Neural Networks, Reinforcement Learning, Hybrid Solutions, Further Specific Methods
- 3) **Optimization Issues / Use Cases**
 - Network planning, Network diagnostics/insights, Network optimization and control
- 4) **Architectural Aspects**
 - Data-sets – availability and usage, Management of ML models / functions, Keeping control on networks
- 5) Conclusions/outlook

AI/ML Methods for Network Optimization – **Scope**

- Overview on AI/ML methods in consideration
- Already in use or to be used (promising solutions) for network design, planning, and optimization
- No intention to provide yet another classification
- No intention to include/exclude AI/ML methods
- **Focus is on overview**
 - Reference to literature
 - Reference to networking use cases (Section 3)

AI/ML Methods for Network Optimization – **An Overview**

- Neural Networks
 - **Feed-forward** neural networks
 - **Deep neural** networks
 - **Recurrent neural** networks
 - **Convolutional neural** networks
- Reinforcement Learning
 - Basics/overview
 - **Deep** Reinforcement Learning
- Hybrid Solutions
 - Combined analytical and Machine Learning modeling
 - Expert knowledge aided Machine Learning
- Further Specific Methods
 - Generative adversarial networks
 - Kalman type filtering
 - Is it AI?
 - Unsupervised learning and clustering

Use Cases

Use Case Overview

	AI/ML domain		
Network Domain	Planning	Forecast & Diagnostics	Optimization and Control
Radio Access Network			
Non real time			
Near real-time			
Real-time			
Transport Network			
Fronthaul			
Programmable Switches			
Path computation, Traffic matrix			
NFV infrastructure			
Dynamic Resource Allocation			
MEC & NFV			
E2E slicing			
Service assurance, slice config.			
Admission control & resource allocation			
Security			
Application and vertical domain			
Positioning			

- Network element placement**

- Optimal base station placement will become harder in B5G systems with high frequency APS, relays or RIS
- Domain experts can pre-identify set of potential AP locations
- RL techniques used to identify AP locations while keeping deployment costs bounded

- BBU cluster dimensioning**

- Employs NN to estimate processing requirements of different BBU functions according to load
- Example: Based on PUSCH SINR measurements predict % users with given MCS → Cluster compute requirements

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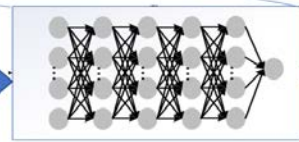
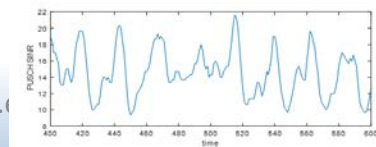
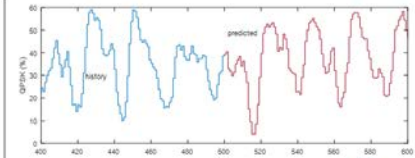
# Neural net construction
def build_model():
    model = Sequential()
    model.add(Dense(1, input_dim = X_train_stdsc.shape[1], activation = 'tanh'))
    model.add(Dense(1, activation = 'tanh'))
    model.add(Dense(1, activation = 'tanh'))
    model.add(Dense(1, activation = 'tanh'))
    model.add(Dense(1))

    model.compile(loss = loss_function, optimizer = optimizer)

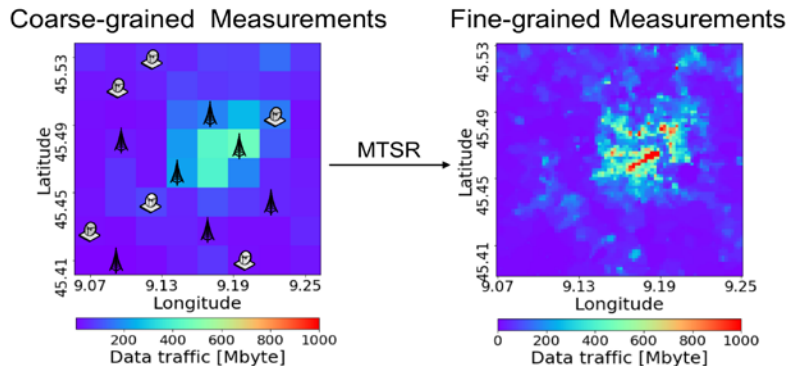
    return model

model = build_model()
model.summary()

# Train the Model
model.fit(X_train_stdsc, Y_train_stdsc, epochs = training_epochs, batch_size = batch_size)
    
```

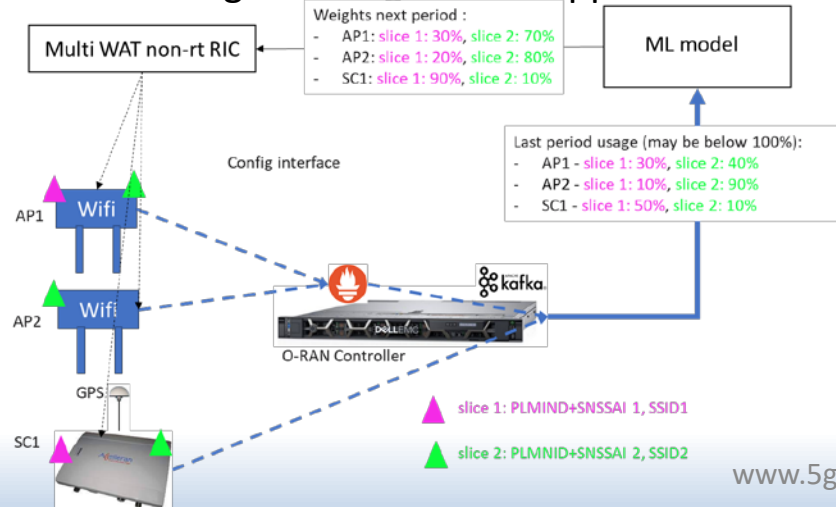


- **Forecasting traffic dynamics**
 - Goal is to provide high accuracy predictions from coarse grained traffic probes
 - Super Resolution or GAN techniques are used
- **Real-time network attack detection**
 - Target is DDoS
 - Traffic classified in conversations (flows)
 - AI technique determines if conversation constitutes an attack

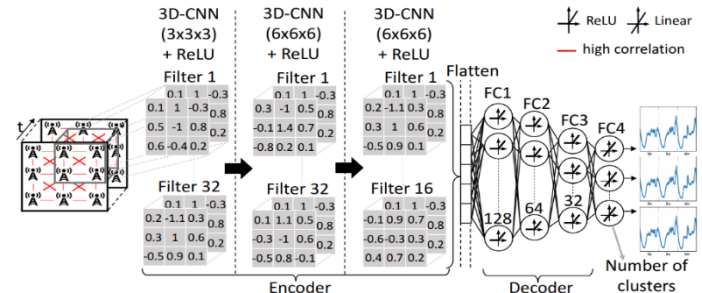




- **RAN slicing in multi-tenant networks**
 - Allocate slice weights based on geographic SLA
 - Using Multi-Agent DRL approach



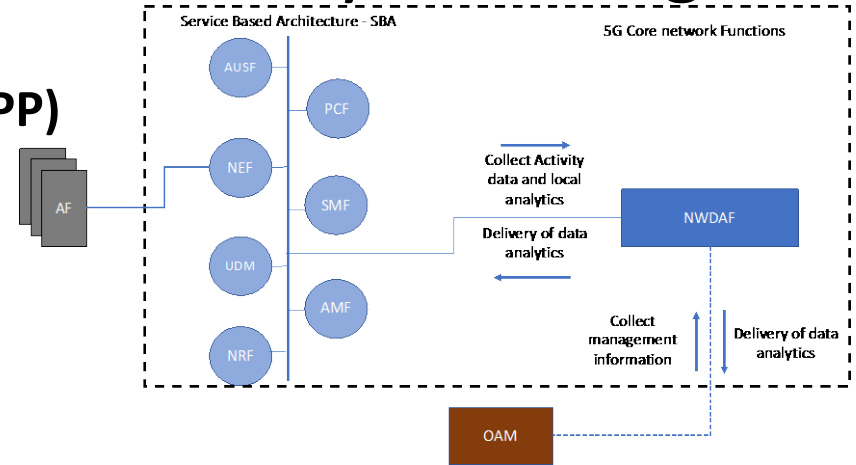
- **Joint slice based demand prediction and resource allocation**
 - Goal: Allocate computation resources to slice software functions (BBU, 5GC, applications)
 - Approach: DNN forecasts computational demand generating a feedback signal that drives the MANO resource allocation function



Architecture aspects

Data sets – availability and usage

- **Network Data Analytics Function (3GPP)**



- Experiential Networked Intelligence (ETSI ENI)
 - Five categories - Infrastructure Management, Network Operations, Service Orchestration and Management, Assurance, Network Security
- Management Data Analytics Service
- Widely available data-sets and open source solutions

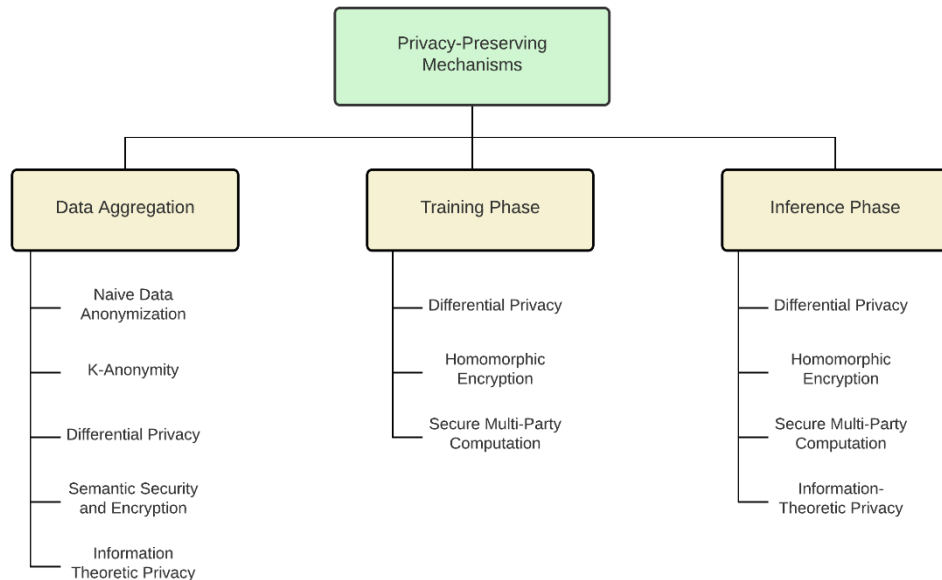
- Requirements for model management
- **Model lifecycle management**
 - Model training
 - On-boarding / deployment of models
 - Monitoring, evaluation and reconfiguration
- Interface management

ITU-T Y.3172 (Architectural framework for machine learning in future networks including IMT-2020) REQ-ML-MNG-004 states that “The ML architecture is required to support an orchestration functionality to manage all the ML functionalities in the network”.

According to REQ-ML-MNG-004, ML model management includes (i) training, (ii) monitoring, (iii) evaluation, (iv) performance-based selection and (v) reconfiguration / updating of the ML models.

Keeping control on networks

- Zero Trust Management
 - **Privacy concerns**
-
- Anomaly detection
 - Cross-optimisation
 - Scaling operations in a network service orchestration platform
 - Enabling AI/ML in vertical domains



© <https://arxiv.org/abs/2004.12254>

Thank you for your attention

Thank you for your contributions!!