

ITU-ML5G-PS-014: Build-a-thon(PoC)

### Autonomous Resource Allocation for Emergency Network Slice

TEAM: AUTOMATO

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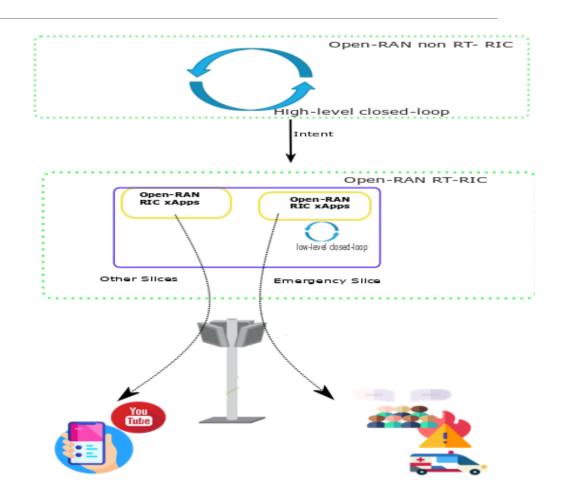
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HOST: ITU FOCUS GROUP AUTONOMOUS NETWORKS (FG-AN)

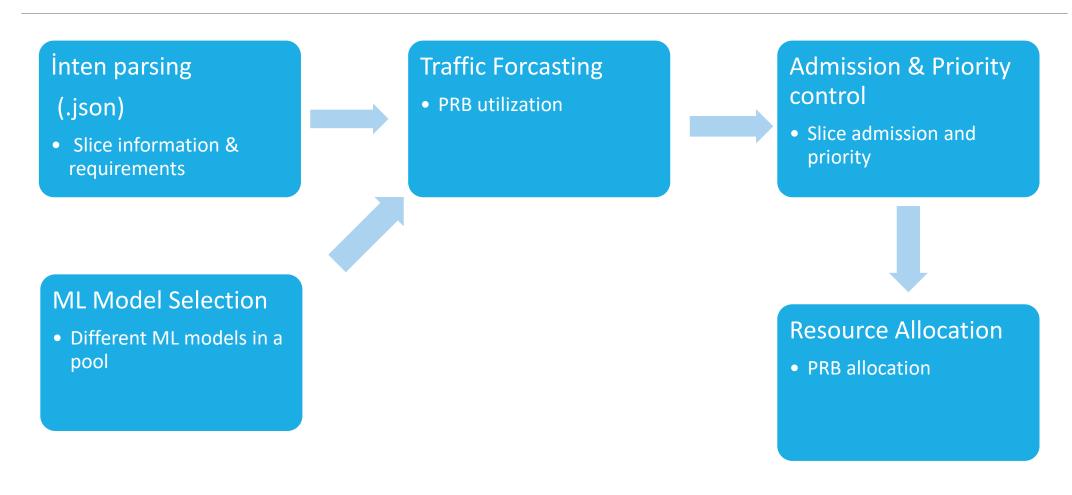
## Problem Definition

Make resource allocation decision for **emergency slice in an autonomous way:** 

- ☐ Get & parse high-level intent
  - Requirements for ES
  - Information regarding data, ML models
- Make traffic analysis
- ☐ Admission & priority control
- Make resource allocation

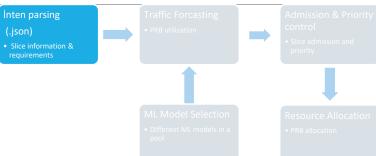


# **Building Blocks**



# Intent Parsing

- We get intent from higher-loop (Open-RAN NonRT-RIC).
- Indicates if there is an emergency case and monitoring xApp is triggered.
  - Information about where data and ML models are.



# Traffic Forecasting

■ We monitor RAN resource utilization Checks for available PRBs for emergency slice • We apply Gaussian Process Regression (GPR) for time-series forecast of PRB utilization ■ We develope this building block as a RIC xApp. This is our first xApp for this problem. Upper bound for ☐ Real data (PRB utilization) taken from paper [1]. the prediction over W samples used for training (last **P** samples) PRB utilization of a given slice Trained with 1000 samples Matern kernel is selected Retrain GPR at every ~15 min.

PRB utilization (%) forecasting-Mean absolute error (MAE) MAE for future MAE for future 4000 points 2000 points

7.725 4.313

# Priority Control

- RAN is shared with other possible slices that have certain number of PRBs dedicated to them. How much resource should we allocate to emergency slice (ES)?
- Level of emergency
  - Protect the existing slices first (ES is the second priority)
    - Algorithm 1 (ALG 1)
  - Always give the first priority to ES
    - Algorithm 2 (ALG 2)



## Resource Allocation

- ALG 1 and ALG 2 decide on resource allocation
  - How many PRB should we allocate for ES?
- Resource Allocation block is impelemented as **RIC xApp. This is our second RIC xApp.**



### ALG 1

**IDEA**: Amount of PRBs allocated to other slices are not always used (underutilization). Predict unused PRBs of other slices (leftover from other network slices) and allocate them to ES.

#### For each other slice n

Step 1: Train GPR with the latest P training data

Step 2: Forecast PRB utilization over the next W samples with GPR -> Un

Step 3: Calculate maximum possible PRB utilization-> Cn = Un + on

Step 4: Calculate forecasted PRB usage over next W samples -> Bn = TnCn

**End** 

Step 5: Calculate available PRBs for Emergency Slice ->PES =  $T - \sum_{n=1}^{N} T_n C_n$ 

Step 6: Allocate PRBs to ES -> PES

Resource allocation (ALG 1)

**Example: Two slices** 

Total system PRBs = 100

Amoun of PRBs allocated to first slice is 40, utilization is %80 (predicted)-> 8 PRBs unused

Amoun of PRBs allocated to second slice is 60, utilization is %90 (predicted)-> 6 PRBs unused

PRBs allocated to emergency slice is **8 +6 = 14** PRBs

## ALG 2

**IDEA**:Emergency slice needs **E** amount of PRBs (fixed). Allocate the available PRBs to ES first. If it is not enough borrow PRBs from other slices. Aim is to always give E PRBs to ES. ES has the first priority.

Resource allocation (ALG 2)

### Solve an optimization problem:

Total PRBs given to slice n

PRB needed for slice n at time

Number of PRB taken from slice n at time t to be used for emergency slice

$$\min \sum_{t=1}^{W} \sum_{n=1}^{N} \max\{0, x_n(t) - (T_n - y_n(t))\}$$
s.t.  $0 \le y_n(t) \le T_n \ \forall n$ 

s.t. 
$$0 \le y_n(t) \le T_n \ \forall n$$

$$\sum_{n=0}^{N} y_n(t) \ge E$$

Guarantee that emergency slice gets enough PRBs

We damage slice n this amount by taking PRBs from it

Transform this problem to a solvable integer problem

<sup>[2]</sup> X. Foukas et al., "Orion: RAN Slicing for a Flexible and Cost-Effective Multi-Service Mobile Network Architecture," in ACM MobiCom, 2017.

<sup>[3]</sup> Armin Okic, Lanfranco Zanzi, Vincenzo Sciancalepore, Alessandro Redondi, Xavier Costa-Pérez, "π-ROAD: a Learn-as-You-Go Framework for On-Demand Emergency Slices in V2X Scenarios", IEEE INFOCOM, 2020.

## ALG 2 (cont.)

**IDEA**: Emergency slice needs **E** amount of PRBs (fixed). Allocate the available PRBs to ES first. If it is not enough borrow PRBs from other slices. Aim is to always give **E** PRBs to ES. ES has the first priority.

Solve an integer programming using auxiliary variable  $u_n$ :

allocation (ALG 2)

Resource

$$\min \sum_{t=1}^{W} \sum_{n=1}^{N} u_n(t)$$

$$\tilde{x}_n(t) + o_n(t) - (T_n - y_n(t)) \le u_n(t)$$

$$0 \le y_n(t) \le T_n \ \forall n$$

$$\sum_{n=1}^{N} y_n(t) \ge E$$

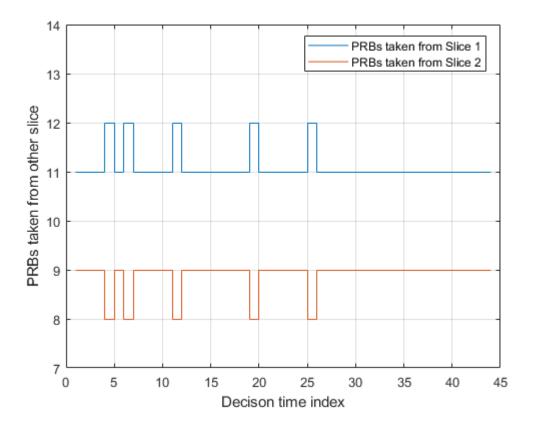
$$u_n(t) \ge 0$$

$$x_n(t) = \tilde{x}_n(t) + o_n(t)$$
 Actual PRB usage at time t in future. This cannot be known in advance Estimated PRB usage with GPR Estimation error. Upper bound can be used provided by GPR.

We have implemented ALG1 & ALG2 and next we will show a demo for ALG2

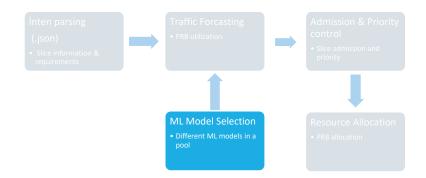
## Experiment: resource allocation

- Two other slices:
  - T40 and 60 PRBs allocated these two slices
  - he network has 100 PRBs (ES should borrow from others)
- Emergence slice needs **E = 20** PRBs
- Result from the Figure: at every decision-time, sum of these PRBs is always 20, which ES needs.



## ML model selection for forecasting

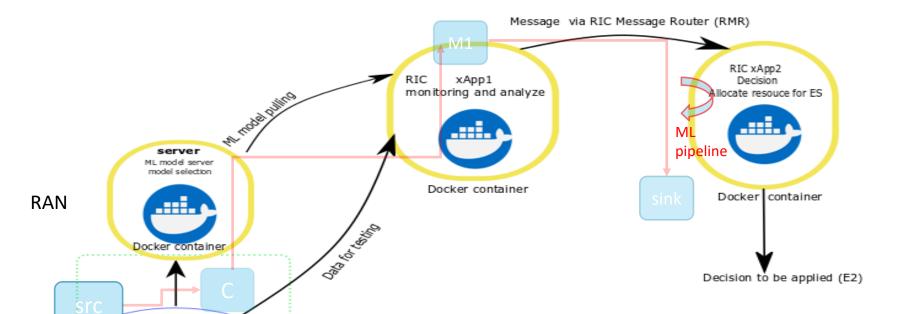
- Different ML models with different performance, comlexity and cost can be available in a ML pool.
- We implement this scheme as a microservice (docker).
  - Webserver stores different ML model
  - Possible to fetch different ML models when forecasting



## Our implementation

PRB data

Local repository



- ☐ Implemented RIC xApp1, xApp2 and server as a microservice.
- ☐ These xApps can communicate through a RIC (O-RAN messaging) message.
  - e.g., xApp1 can send prediction to xApp2 over the network

### PoC Demo

We have a quick demo:

- Ubuntu 20.04 Works fine
- Need docker and docker compose

https://drive.google.com/file/d/1ouqhouvHlZiYOotnKqFHyIB8J UDdAFIF/view?usp=sharing

```
...
                      meldor@meldor-UX305FA: -/enki/openran/ITUChallenge_BuildaThon_Activity&
                    Predicted PRB util. for Slice 1: 71.7
                    Predicted PRB util, for Slice 2: 71.7
                    Results for ALG2
                    Number of PROs taken from slice 1: [7.0]
Number of PROs taken from slice 2: 1 [13.0]
decision app 1
                    NEW PREDICTION TIME: (1611)
                    Predicted PRB util, for 5lice 1: 55.5
                    Predicted PAB util. for Stice 2: 55.5
                    Results For ALG2
                    Number of PMBs taken from stice 1: [7.0]
                    Number of PREs taken from slice 2: : [13.0]
                   NEW PREDICTION TIME: [[63]]
                    Predicted PRB util. for Slice 1: 67.0
                    Predicted PRB util. for Slice 2: 67.0
                    Results For ALGZ
                    Number of PREs taken from stice 1: (7.8)
                    Number of PROS taken from slice 2: : [13.0]
                    Results For ALG?
                    Number of PRBs taken from slice 1: [7.0]
                    Number of PRBs taken from stace 2: 1 [13.0]
                   NEW PREDICTION TIME: [1631]
                    Predicted PRB util, for Slice 1: 72.1
                    Predicted PRB util. for Slice 2: 72.1
                    Results For ALG2
                    Number of PRBs taken from alice 1: [7.0]
decision_app_1
                    Number of PREs taken from stace 2: : [13.0]
                   NEW PREDICTION TIME: [[64]]
prediction_app_1
                    Predicted PRB util, for Slice 1: 72.8
prediction app 1
                    Predicted PRB util, for Slice 2: 72.8
                    Results For ALG2
                    Number of PRBs taken from since 1: [7.8]
                    Number of PRDs taken from slice 2: : [13.0]
                    Fetching model busic_prediction_model_1.pkl
                    NEW PREDICTION TIME: [165][
                    Predicted PRS util. for Slice 1: 68.5
                    Predicted PRB util. for Slice 2: 58.5
                    1636876866782 7/RMR | INFO| sends: ts=1636876668 src=meldor-LX385FA:4564 target=127.0.0.1:4564 open=0 si
prediction spp 1 | 1636876061163 7/RMR | INFO| sends: fs=1636876061 srcnmeldor-0X385FA:4564 target=127.0.0.1:4564 open=1 su
cc=16 fail=0 (hard=0 soft=6)
mecision app 1
                  | Amsalts For ALGZ
decision app 1
                   Number of PRDs taken from slice 1: [7.0]
                   Number of PROS taken from stice 2: : [13.0]
decision app 1
```

Check for details: https://github.com/ITU-AI-ML-in-5G-Challenge/ITU-ML5G-PS-014-Build-a-thon-PoC---Team-AUTOMATO-

### Conclusion and Future Works

- ☐ A PoC for autonomous slice management
- Resource allocation handled
  - Prediction with machine learning & linear optimization
- ☐ Flexible and modular O-RAN RIC xApps design

### **Future Works:**

- ☐ Integration of all xApps to testbed, real O-RAN infrastructure
  - Real autonomous closed loops
- Other optimizations & improvements
- Dynamic estimation of emergency slice

### Thank you for listening!

Q & A

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