

Motivation of the work

Currently, mobile terminals feature multiple interfaces to adapt to the steadily increasing number of available wireless access networks. This provides a suitable ground for offloading data from cellular to different WIFI access points using the integration of WIFI and LTE offered by LTE v.12 and v.13. There is a parallel trend towards network programming relying on centralized controllers, of which the Software-Defined Network (SDN)[1] architecture with the OpenFlow[2] protocol is a clear exponent.

Thesis Objectives

We intend to design and implement a SDN-oriented global network optimization algorithm. This algorithm will use flow steering and will be applied on an SDN[1] architecture in which the end-terminals are integrated with the core network.[3].

Research plan

1. First year

Part 1

- Establishing an essential knowledge of cellular standards.
- Establishing an essential knowledge of network protocols:
 - Network managing protocols: ICMP, SNMP.
 - Network managing flow-based protocols: OpenFlow, NetFlow, sFlow.
 - Remote terminal configuration protocols: SNMP, NetConf, TR-069, OMA LWM2M.
 - Statistics collection daemons: collectd, sFlow.
- Mastering SDN:
 - Applying the SDN approach to control a wireless network using the Mininet test bed.
 - Using the RYU controller to monitor, configure and manage flows in a network.

Part 2

- Design of a network prototype.
- Use the Mininet test bed to emulate a backhaul network based on the designed prototype.
- Control the network using the RYU controller.

2. Second year

Part 1

- Designing a Global Network Optimization Algorithm.

Part 2

- Mile stone: Submitting a paper (June 2016) [6]

3. Third year

Part 1

- Develop a successful flow based access point assignment algorithm and take advantage of the SDN and TSA architectures.

Part 2

- Implement different profilers to help the controller with access point assignment:
 - Flow profiler.
 - Mobility profiler.

Next Year Planning

- Test the efficiency of the algorithms on a real network emulation with real terminals scenarios
- Implementing and testing the user profiler on the network
- Preparing an article to submit to a network journal
- Writing the Thesis

References

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- [2] Agarwal, Sankalp, Murali Kodialam, and T. V. Lakshman. "Traffic engineering in software defined networks." *INFOCOM 2013 Proceedings*. IEEE, 2013.
- [3] Vissicchio, Stefano, et al. "Safe updates of hybrid SDN networks." *Université catholique de Louvain, Tech. Rep* (2013).
- [4] Musoff, Howard, and Paul Zarchan. *Fundamentals of Kalman filtering: a practical approach*. American Institute of Aeronautics and Astronautics, 2005.
- [5] Wolpert, Daniel M., and Zoubin Ghahramani. "Computational principles of movement neuroscience." *Nature neuroscience* 3.11s (2000): 1212.
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- [9] Giraldo-Rodríguez, Carlos, Mhiri, Saber, et al. "TSA, an SDN architecture including end terminals." *Consumer Communications & Networking Conference (CCNC)*, 2016. IEEE, 2016.
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Results & Discussion

We applied Kalman filtering, to predict terminal positions [4] [5].

We managed to predict the right position of the terminal with a success rate of **75%**.

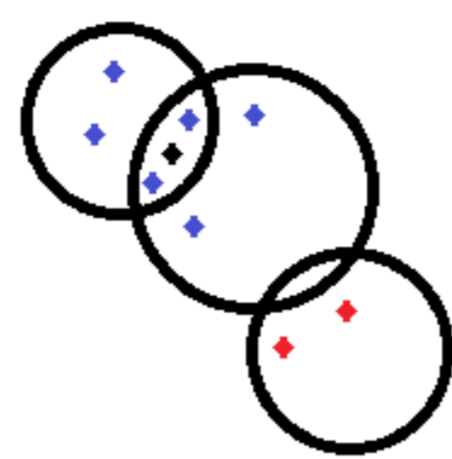


Fig 1 : Terminal position prediction

We developed a flow predictor that is able to predict the next flow type generated by the terminals with an average F1 value of **0,85**. It was presented in the IPIN 2016 [6].

Algorithm

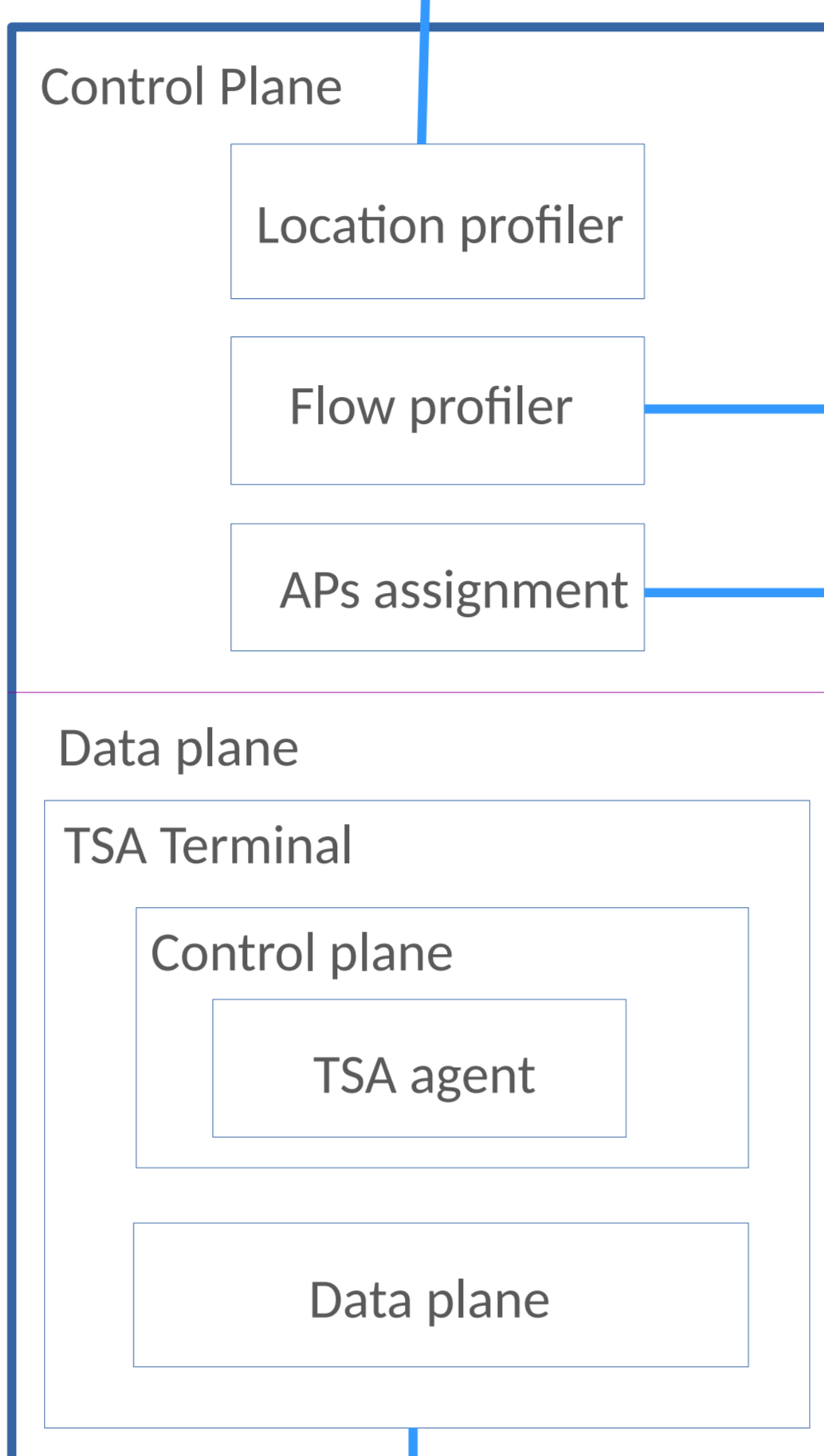
We developed a new algorithm that takes into consideration the flow generated by the users A and the bandwidth available in the network access points B. The algorithm uses a modified genetic algorithm to assign the terminals to the right access points

$$Fit = (\alpha \times A) + ((1 - \alpha) \times B)$$

$$A = \frac{\sum_{i \in AP} N_i}{\text{Total number of requested flows}}$$

$$B = \frac{\sum_{i \in AP} BW_{Assigned_i}}{\sum_{i \in AP} BW_{AP_i}}$$

N_i : number of rightly assigned flow.



TSA is an architecture created by the AtlantTIC research group. it have been tested [9] and published [1][10].

Calibration

Choosing the right α :

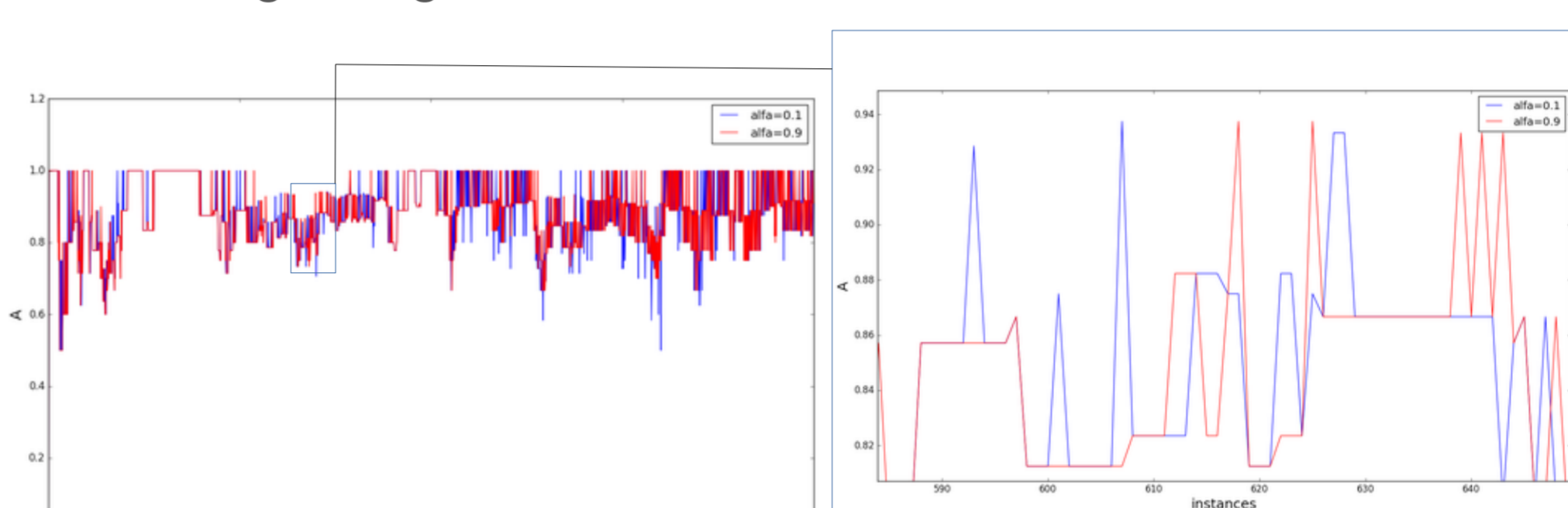


Fig 2 : A values vs. 2000 instances of real data [11] for different α values

We ran the algorithm with multiple values of α and we check A for $\alpha = 0,1$ and $\alpha = 0,9$. For an $\alpha = 0,1$, we obtain an improvement of **81,7%** in A.

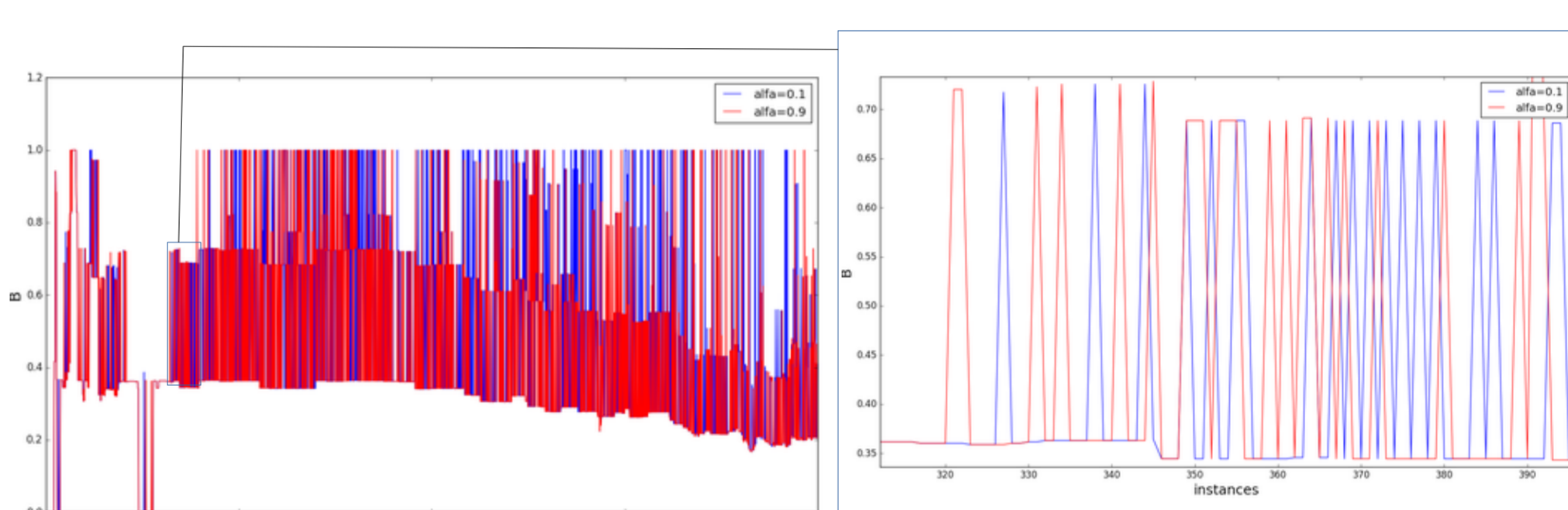


Fig 3 : B values vs. 2000 instances of real data [11] for different α values

We ran the algorithm with multiple values of α and check B for $\alpha = 0,1$ and $\alpha = 0,9$. for $\alpha = 0,1$, we obtain an improvement of **75,95%** in B values.

Choosing the right number of genetic algorithm iterations :

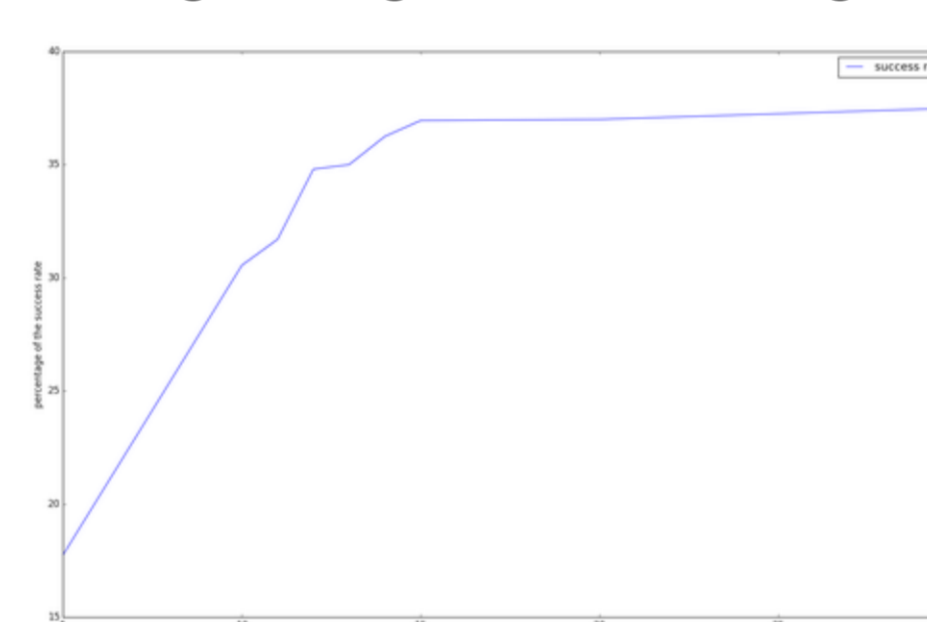


Fig 4 : optimum success rate vs. number of iterations

The algorithm achieves its optimum performance after 15 iterations .

We set a limit of 15 iterations on the genetic algorithm for $\alpha=0,1$.

Results

Using real flow and position data :

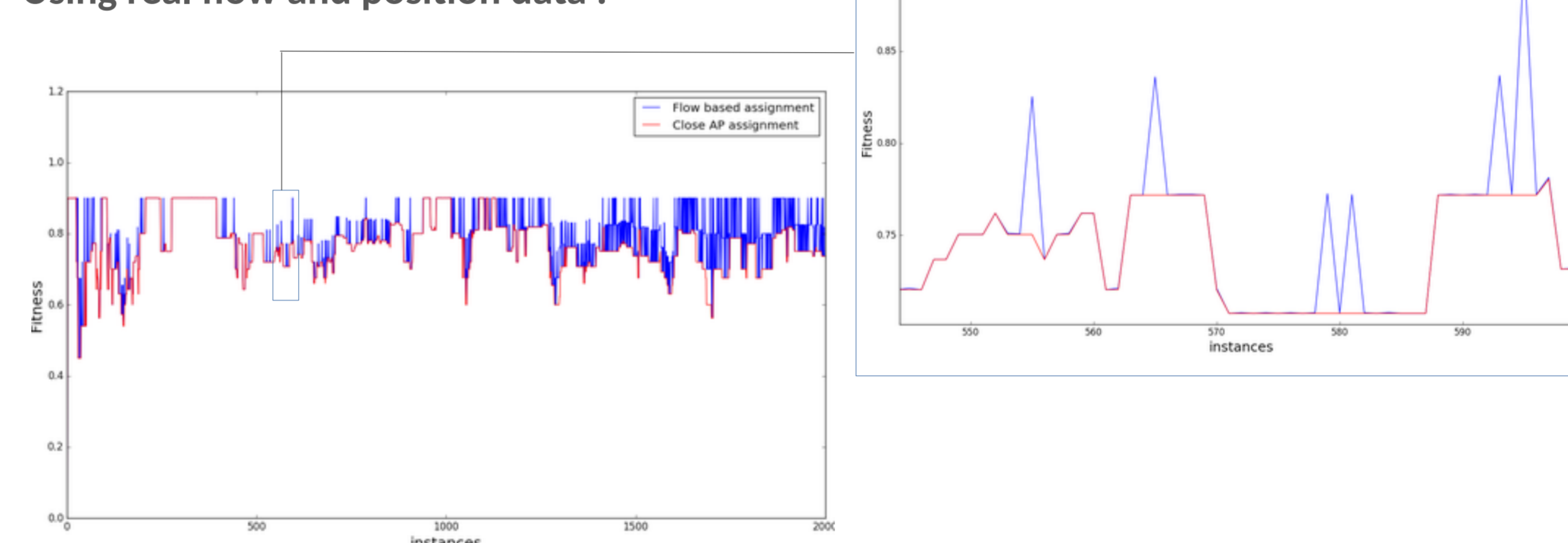


Fig 5 : Fitness values vs. 2000 instances of real data with 15 iterations and $\alpha = 0,1$

Running our flow based access points assignment algorithm with 15 iterations and an $\alpha = 0,1$, our AP assignment algorithm provides a **52,17%** better Fitness value than closest-AP assignment [7] [8] over the first 2000 instances of real data taken from the campus of Rice university[11] .

Overall, our AP assignment algorithm provides a **31,98%** better Fitness value than closest-AP assignment [7] [8] over all the instances of real data.

Using real flow data and predicted position data :

Update interval	Percentage of rightly assigned terminals	Terminal assignment success rate
60s	77,82%	40,59%
45s	78,29%	40,84%
30s	78,55%	40,97%

Fig 6 : Table of assignment success rate with predicted position data

Running our flow based access points assignment algorithm with 15 iterations and $\alpha = 0,1$, our AP assignment algorithm provides **40,8%** better Fitness value than closest-AP assignment [7] [8] over the first 2000 instances of real data with predicted positions.

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