







5.2. Application

As mentioned above, we will use video conferencing in the case of a mobile user who needs to take a path where the signal quality drops to an unacceptable level due to a gap in coverage, giving a very low QoS. This can be remedied by using the CR. but problematic arises: WHEN and WHY to use cognitive radio?

Most research related to the QoS of video conferencing, take into account throughput as pertinent parameter. For this reason, we choose the "Throughput" as a single pertinent parameter for our application. For this, a throughput classification is required, and as we play the role of the expert, we have created our own database following certain rules in order to apply our approach. The database was divided into two parts, the first one for learning and the second one for testing. The value of the throughput will change almost ever' time it is measured during the day even on the same route, for this reason, our measures have been taken into account for 5 weeks at 3 different intervals of the day (Sam - 1 lam. 1 lam - 3pm, 3pm - 5pm) excluding weekend.

We proposed to affect throughput on three classes:

- Gold for samples where the throughput is greater than 384 Kb/s, ensuring 100% quality satisfaction of the user.
- Silver for samples where the throughput is between 160 Kb/s and 384 Kb/s, of acceptable quality.
- Bronze for samples where the throughput is less than 160Kb/s. This means that video conferencing is not satisfactory, and it is the class that interests us because this is when we use the CR.

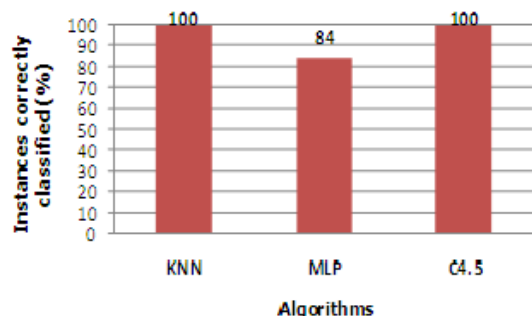


Figure 2: Database Classes

5.2.1. First Question "WHEN"?

For the data classification, we used three different algorithms derived from the field of machine learning.

- The k-nearest neighbor's algorithm (K-NN) which is a supervised classification algorithm.
- The multilayer perceptron algorithm (MLP of neural networks).
- The C4.5 algorithm of decision trees.



We remark that the multi layer perceptron has not given satisfactory results compared with the other algorithms: however we did not get 100%. Despite the change of parameters such as the number of hidden layers, the number of epochs and the learning rate.

We remark also that the two other algorithms (KNN and C4.5) have given results of 100%, however if we take a look at the generated tree, we see is not exactly what we want because for him the GOLD class starts from a rate of 380 Kb/s instead of 384 Kb/s, this can generate after mistakes with misclassifying some instances.

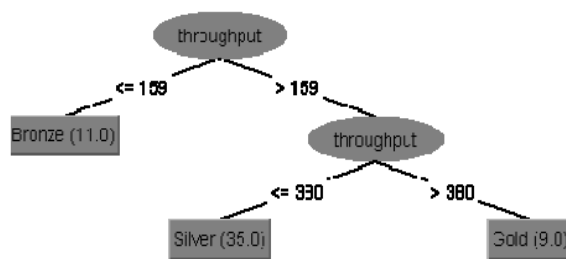


Figure 4: Decision rate

Concerning the K-NN algorithm, it was tested with several values of K on the test database and each time the result was different, but it is satisfactory until the value of K=6.

For our approach, we chose this algorithm, as it gave the best results (shown below) in terms of reliability and Clustering.

Table 1: Classification obtained with KNN

Values of K	Instances correctly classified		Misclassified Instances	
	Count	Percentage	Count	Percentage
K=1	20	100%	0	0%
K=2	19	95%	1	5%
K=3	19	95%	1	5%
K=4	18	90%	2	10%
K=5	18	90%	2	10%
K=6	18	90%	2	10%

For K= 1: All the throughput examples were well posted in their appropriate class. Whereas, with K=2, one sample was misclassified, we notice that in the graph.

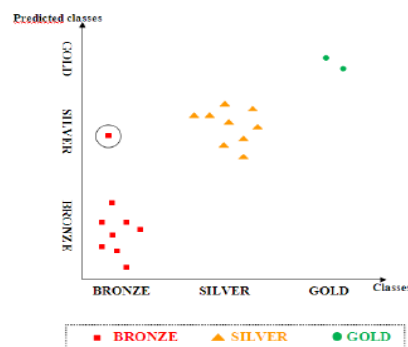


Figure 5: Classification result of KNN with K=2

Note: items close to the threshold of a class are misclassified. For example, the instance which have a

throughput of 159Kb/s was ranked on Silver class instead of the Bronze one, for the other values of K. the result is more divergent.

*Instance: 13*

*Throughput: 159.0*

*Predicted Class: Silver*

*Class: Bronze*

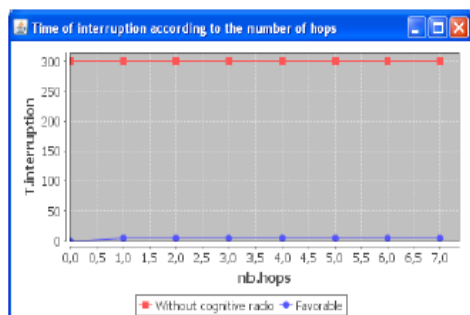
**Report:** according to the results of the classification, the cognitive radio will be activated each first day of week from 8am to 1 pm and from 3pm to 5pm, and every Wednesday from 8am to 5pm because in these intervals the throughput belongs to the Bronze class <160Kb/s. So, through this, the question When is answered.

For more general rules, it would be interesting to consider the other QoS video conferencing parameters and classify these data using other methods of artificial intelligence such as fuzzy logic and genetic algorithms.

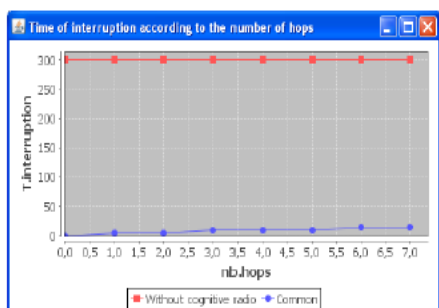
### 5.2.3. Experimentation Results

To support our proposal and to better understand, we compared the QoS with and without cognitive radio (case without CR means that our terminal has only one access technology and works on the same frequency band).

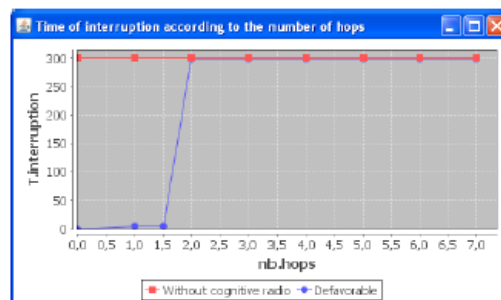
The graphs below illustrate this comparison. For being readable, we supposed that the maximum time of rupture is 5 min and that the tree band set contains 7 bands.



**Figure 6:** Comparison between the favourable scenario and the case without cognitive radio



**Figure 7:** Comparison between the common scenario and the case without cognitive radio



**Figure 8:** Comparison between the unfavourable scenario and the case without cognitive radio

## 6. Conclusions

We presented in this paper a new approach that uses Cognitive Radio to improve wireless communication for a cognitive radio mobile terminal by enhancing the QoS of video conferencing application. Our contribution is positioned in learning from events (machine learning). Our expert role has allowed us to choose the throughput parameter to perform a classification that allows the terminal used to gain experience for future events that means that it will know when and where it will activate the cognitive radio.

The usefulness of cognitive radio is a hypothesis that has been proved based on the required time for a connection to a new frequency band and this, whatever of the number of frequency bands that a terminal used to remedy a failed connection.

In our future work, we think we can improve the wireless links reliability and ensure good quality of service to CR mobile terminals [26] [27] [28] by integrating MAS. We will seek also to reduce the impact of mobility on cognitive radio communications by building predictive models of mobility by referring to previous work such as [29] and [30].

## References

- [1] J. Milola, (1999) "Cognitive radio - model-based competence for software radios". Licentiate Thesis, KTH, Stockholm.
- [2] J. Neel, (2006) "Analysis and Design of Cognitive Radio Networks and Distributed Radio Resource Management Algorithms", Faculty of the Virginia Polytechnic Institute and State University.
- [3] C. Clancy et al, (2007) "Applications of machine learning to cognitive radio networks", IEEE Wireless Communications, vol. 14, no. 4.
- [4] E. Hossain, D. Niyam, Zhu Han, "Dynamic Spectrum Access and management in cognitive radio networks", Cambridge University Press 2009.
- [5] S. Busanelli et al, (2011) "Vertical Handover between WiR and UMTS Networks: Experimental Performance Analysis", International Journal of Energy, Information and Communications Vol. 2, Issue I.
- [6] Z. Daia et al, "Vertical handover criteria and algorithm in IEEE 802.11 and 802.16 hybrid networks", Laboratoire de Motorola Paris.

- [7] Atiq Ahmed el al, (2011) "An agent Based Architecture for Cognitive Spectrum Management", Australian Journal of Basic and Applied Sciences, 5(12): 682-689.
- [8] Chu Tian el al, (2010) "Spectrum Trading in Cognitive Radio Networks An Agenl-based **Model under Demand Uncertainty**", Global Telecommunications Conference.
- [9] Dzikowski Jacek, Cynthia Hood, (2009) "An Agent-Based Simulation Framework For Cognitive Radio Studies", Simutools '09 Proceedings of the 2nd International Conference on Simulation Tools and Techniques.
- [10] Galindo-Serrano Ana, Lorenza Giupponi, (2009) "Aggregated Interference Control for Cognitive Radio Networks Based on Multi-agent Learning". Proceedings of the 4th International Conference On CROWNCOM.
- [11] Gaurav S, Kasbekar, Sarkar S, (2010) "Spectrum auction framework for access allocation in cognitive radio networks", IEEE.ACM Transactions on Networking, vol. 18, pp. 1841- 1854.
- [12] Jiandong Li, Chungang Yang, (2010) "A Markovian Game-Theoretical Power Control Approach In Cognitive Radio Networks: A Multi-Agent Learning Perspective", Wireless Communications and Signal Processing (WCSP).
- [13] Letaief Ben, IC.Wei Zhang, (2009) "'Cooperative Communications for Cognitive Radio Networks", Proceedings of the IEEE.
- [14] Li H, (2009) "Multi-agent Q-Learning of Channel Selection in Multi-user Cognitive Radio Systems: A Two by Two Case", Systems, Man and Cybernetics.
- [15] Lim Kok et al, (2011) "Achieving Context Awareness and Intelligence in Distributed Cognitive Radio Networks: A Payoff Propagation Approach", Workshops of International Conference on Advanced Information Networking and Applications.
- [16] Mir Usama, Leila Merghem-Boulahia, Dominique Gaili, (2010) "A Cooperative Mulliagent Based Spectrum Sharing", Sixth Advanced International Conference on Telecommunications.
- [17] Mir Usama, Leila Merghem-Boulahia, Dominique Galti, (2010) "Dynamic Spectrum Sharing in Cognitive Radio Networks: a Solution based on Mulliagent Systems". International Journal on Advances in Telecommunications, vol 3 no 3 & 4.
- [18] Usama Mir, LeilaMerghcm-Boulahia, Dominique Gaiti. (2010). COMAS: A CooperaliveMulliagent Architecture for Spectrum Sharing. EURASIP Journal on Wireless Communications and Networking, Volume 2010, Article ID 987691.
- [19] Mir Usama, Leila Merghem-Boulahia, Dominique Gaiti, (2009) "Utilization of a Cooperative Mulliagent System in the Context of Cognitive Radio Networks". © Springer-Verlag Berlin Heidelberg 2009, LNCS 5844, pp. 100-104.
- [20] Mir Usama, Leila Merghem-Boulahia, Dominique Gaili, (2010) "Mulliagent Based Spectrum Sharing Using Petri Nets". Y. Demazeau el al. (Eds.): Trends in PAAMS, AISC 71, pp. 537-546. [springerlink.com](http://springerlink.com) © Springer-Verlag Berlin Heidelberg.
- [21] Qi Zhao, Qin shi, Wu Zhijie, (2011) "Self-Organize Network Architecture for Multi-Agent Cognitive Radio Systems", International Conference on Cyber-Hnabled Distributed Computing and Knowledge Discover)'.  
[22] Raiyn J, "Toward cognitive radio handover management based on social agent technology for spectrum efficiency performance improvement of cellular systems Personal", IHKH 19th International Symposium.
- [23] Trigui Hmna, Moez Hssegghir, Leila Mcrghcm Boulahia, (2011) "Gestion dynamique du spectre entre terminaux radio cognitive mobiles", CHIP 2011 - Colloque Francophone sur I Ingenierie des Protocole s.
- [24] Wu Cheng el al, (2010) "Spectrum management of cognitive radio using multi-agent reinforcement learning", AAMAS '10 Proceedings of the 9th International Conference on Autonomous Agents and Multiagent Systems.
- [25] Xie Jiang, Howitt Ivan, Raja Anita, (2007) "Cognitive Radio Resource Management Using Mulli Agent Systems", Consumer Communications and Networking Conference.
- [26] A. Amraoui, W. Baghli and B. Benmammam, (2012) "Improving video conferencing application quality for a mobile terminal through cognitive radio", 14th IKHK International Conference on Communication Technology (ICCT2012). Chengdu, China, November 9th-1 Ith. to appear.
- [27] B. Benmammam, A. Amraoui and W. Baghli, (2012) "Performance improvement of wireless link reliability in the context of cognitive radio". UCSNS International Journal of Computer Science and Network Security, Vol. 12, No. I, Pages: 15-22.
- [28] A. Amraoui and al, (2012) "Toward cognitive radio resource management based on multi-agent systems for improvement of real-time application performance", 5th IFIP International Conference on New Technologies, Mobility and Security (NTMS 2012). Istanbul, Turkey, May 7th-IOth. to appear.
- [29] N. Samaan, B. Benmammam, K Krief and A. Karmouch. "Prediction-based Advanced Resource Reservation in a Mobile K Environment". 18th IKHK Annual Canadian Conference on Electrical and Computer Engineering. CCHCH05, May 1-4, 2005, Saskatoon Inn, Saskatoon, Saskatchewan Canada.
- [30] B. Benmammam and F. Krief. "MQoS NSLP: a mobility profile management based approach for advance resource reservation in a mobile environment". Proceedings of the 7ih IFIP IEEE International Conference on Mobile and Wireless Communications Networks (MWCN 2005). Marrakech, Morocco. September 19-21,2005