

Development of GNU Radio Blocks for Spectrum Sensing Based on the Analysis of the Autocorrelation of Samples

Héctor Iván Reyes-Moncayo

Universidad de los Llanos

Grupo de Estudio en Redes y Aplicaciones –GERA

Grupo de Investigación Macrypt

Villavicencio-Meta-Colombia

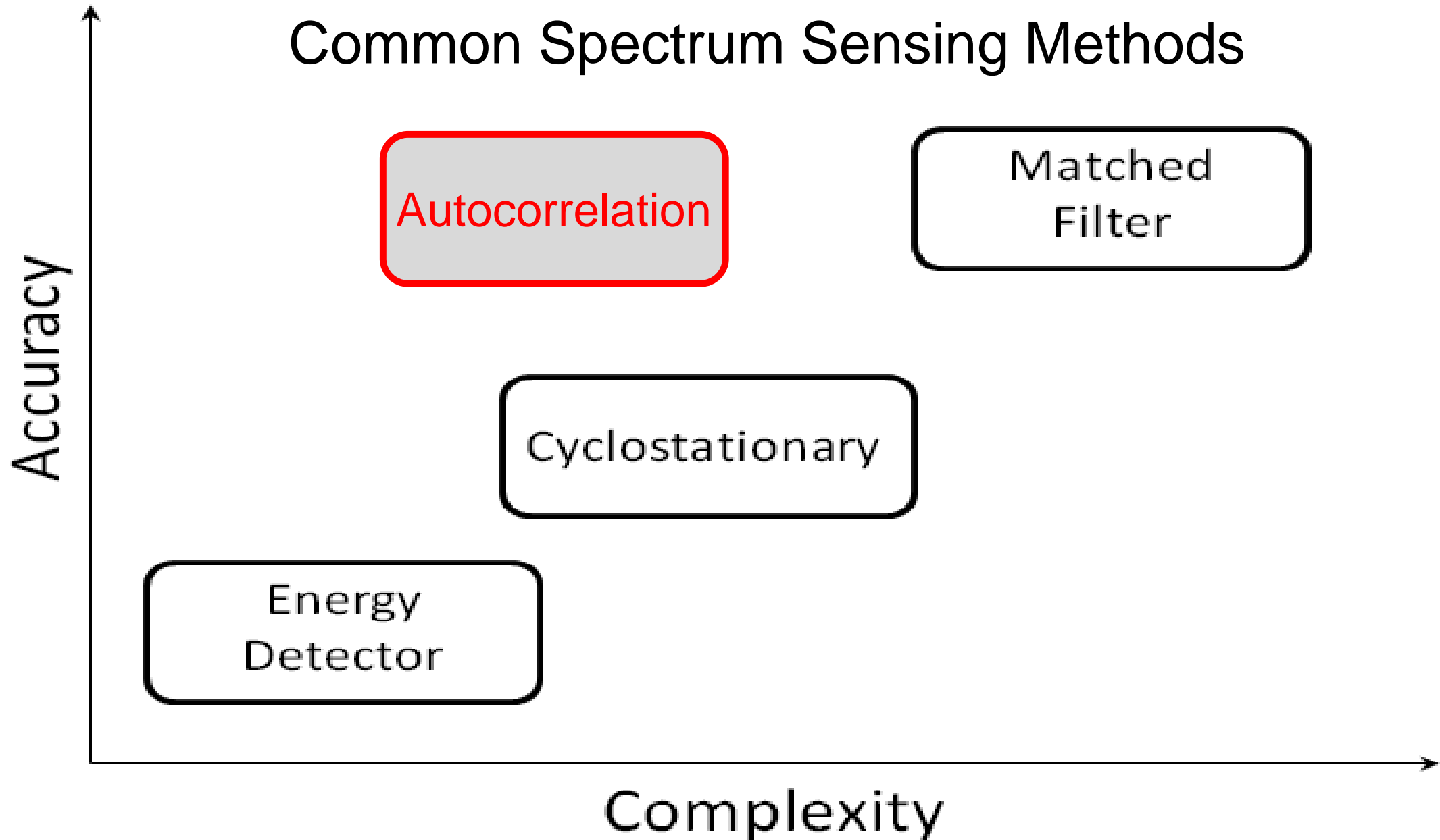


Content

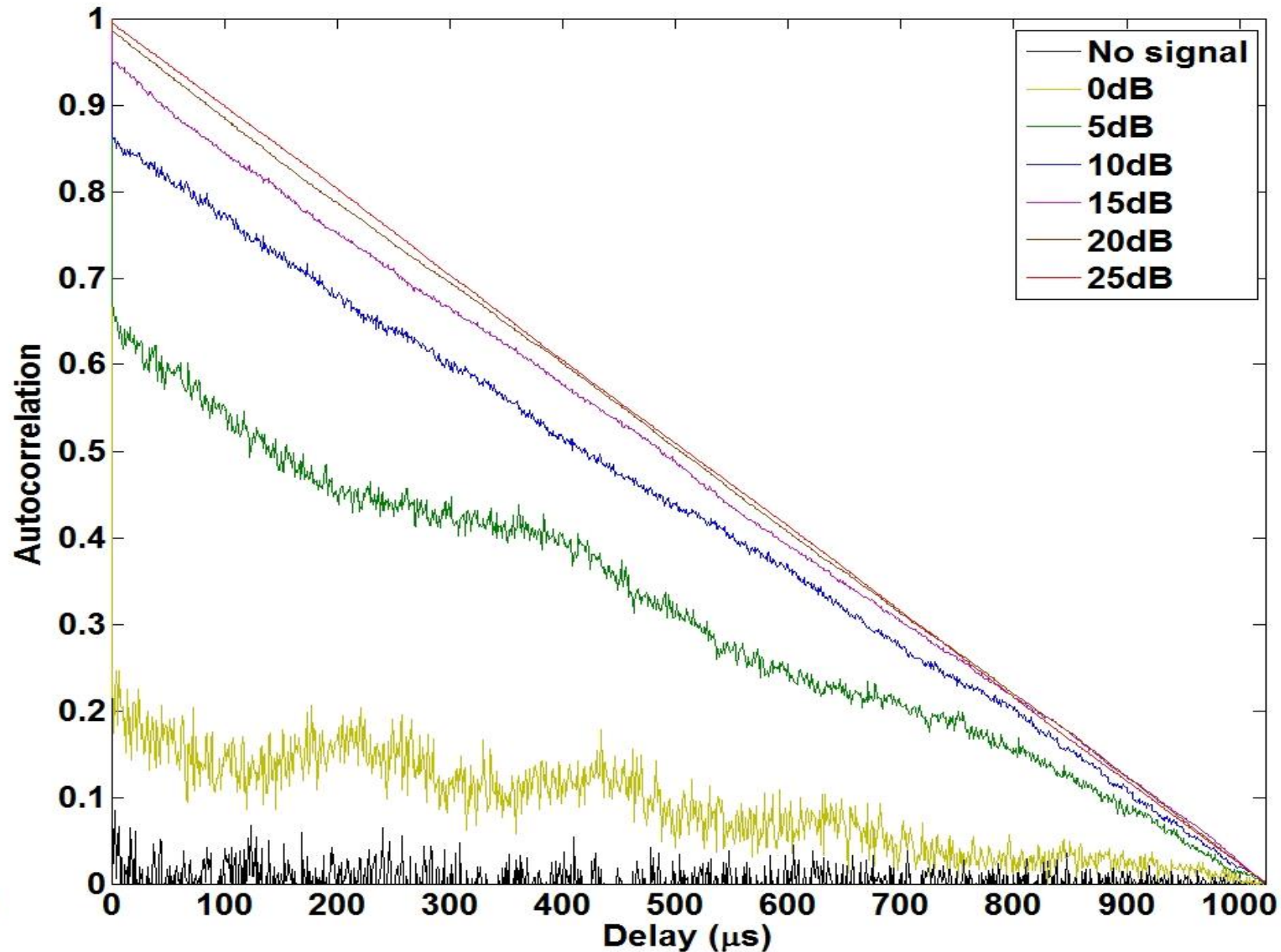
- Common Spectrum Sensing Methods
- The Euclidean method
- The Kmeans method
- The KNN method



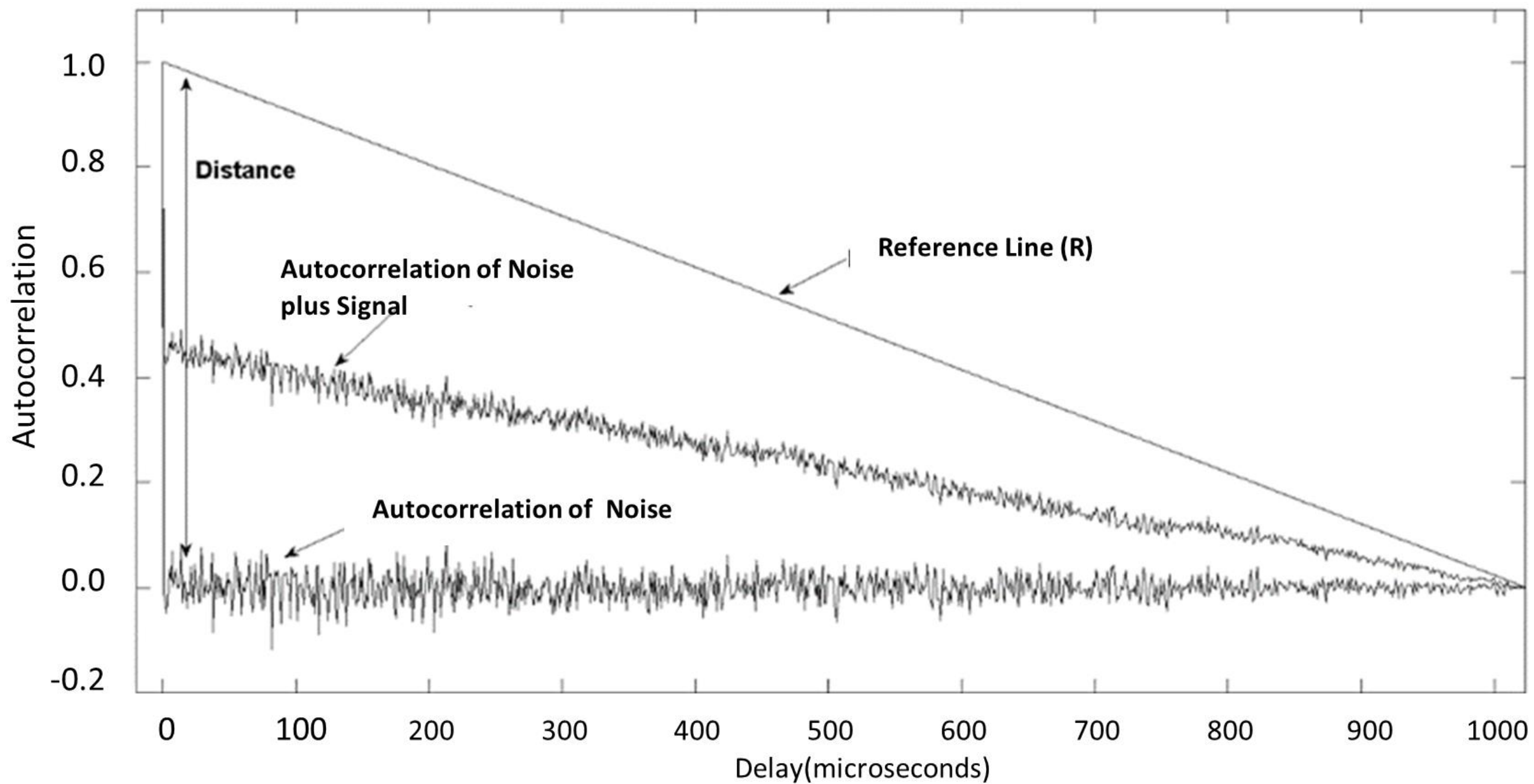
Common Spectrum Sensing Methods



Autocorrelation for Different Conditions



Autocorrelation vs. Delay

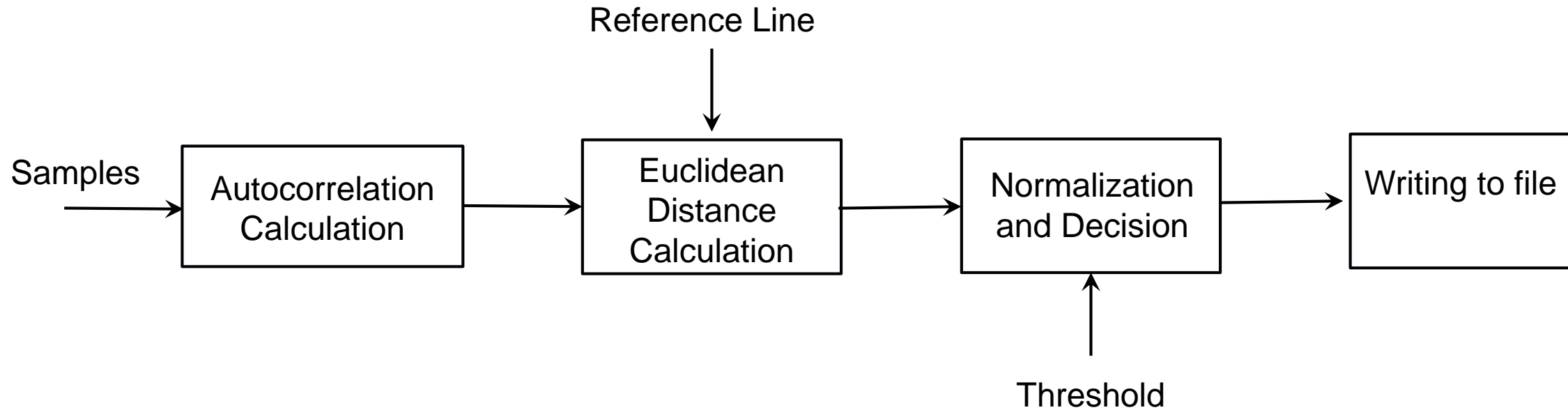


Euclidean Distance Method

$$E(x, y) = \sqrt{\sum_{i=0}^n (x_i - y_i)^2}$$



Euclidean Distance Block Diagram



Euclidean Distance GNU Radio Block

osmocom Source
Sample Rate (sps): 1M
Ch0: Frequency (Hz): 470M
Ch0: Freq. Corr. (ppm): 0
Ch0: DC Offset Mode: Off
Ch0: IQ Balance Mode: Off
Ch0: Gain Mode: Manual
Ch0: RF Gain (dB): 10
Ch0: IF Gain (dB): 20
Ch0: BB Gain (dB): 20

Euclidean_c
Threshold: 940m
Fc: 470M
Location: Villavicencio
Folder: /home/i...CON18/DATA/
Samples: 1.024k

Properties: Euclidean_c

General Advanced Documentation

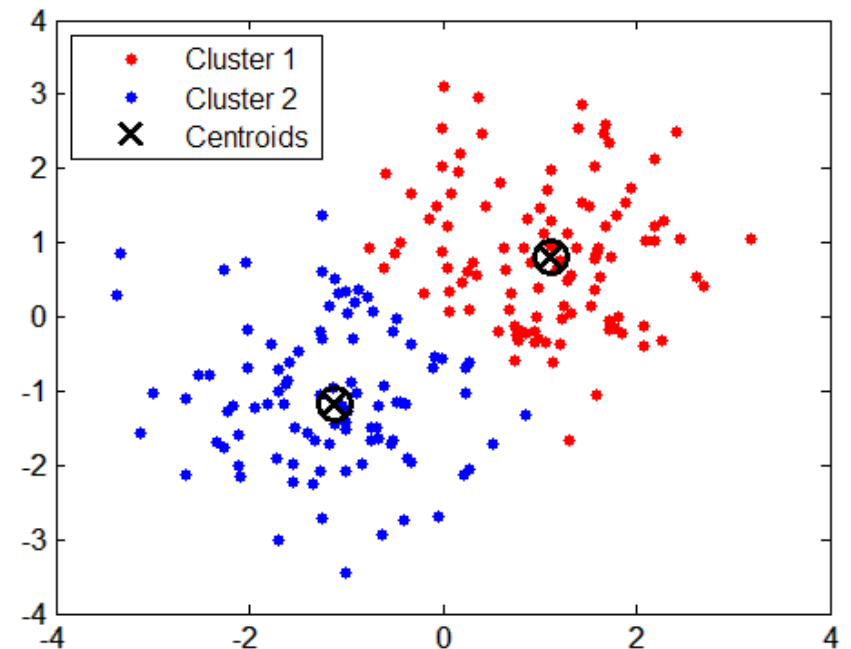
| | |
|-----------|---------------------------------------|
| ID | grcon18_Euclidean_c_0 |
| Threshold | 0.94 |
| Fc | frequency |
| Location | "Villavicencio" |
| Folder | "/home/ivan/Documents/GR_CON18/DATA/" |
| Samples | 1024 |

OK Cancel Apply

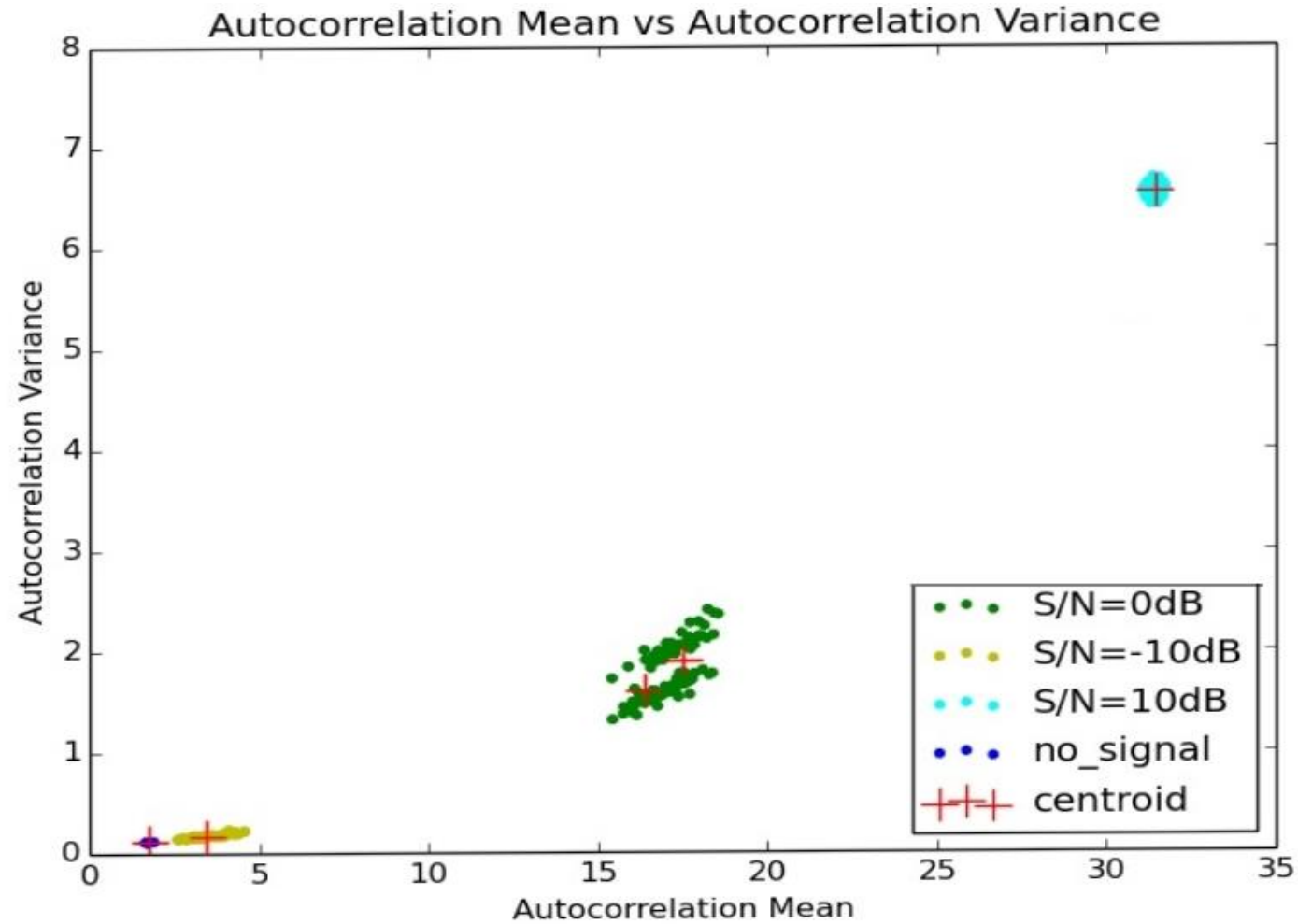
K-Means Clustering Method

$$\sum_{i=0}^n \min_{\mu_j \in C} (||x_j - \mu_i||^2)$$

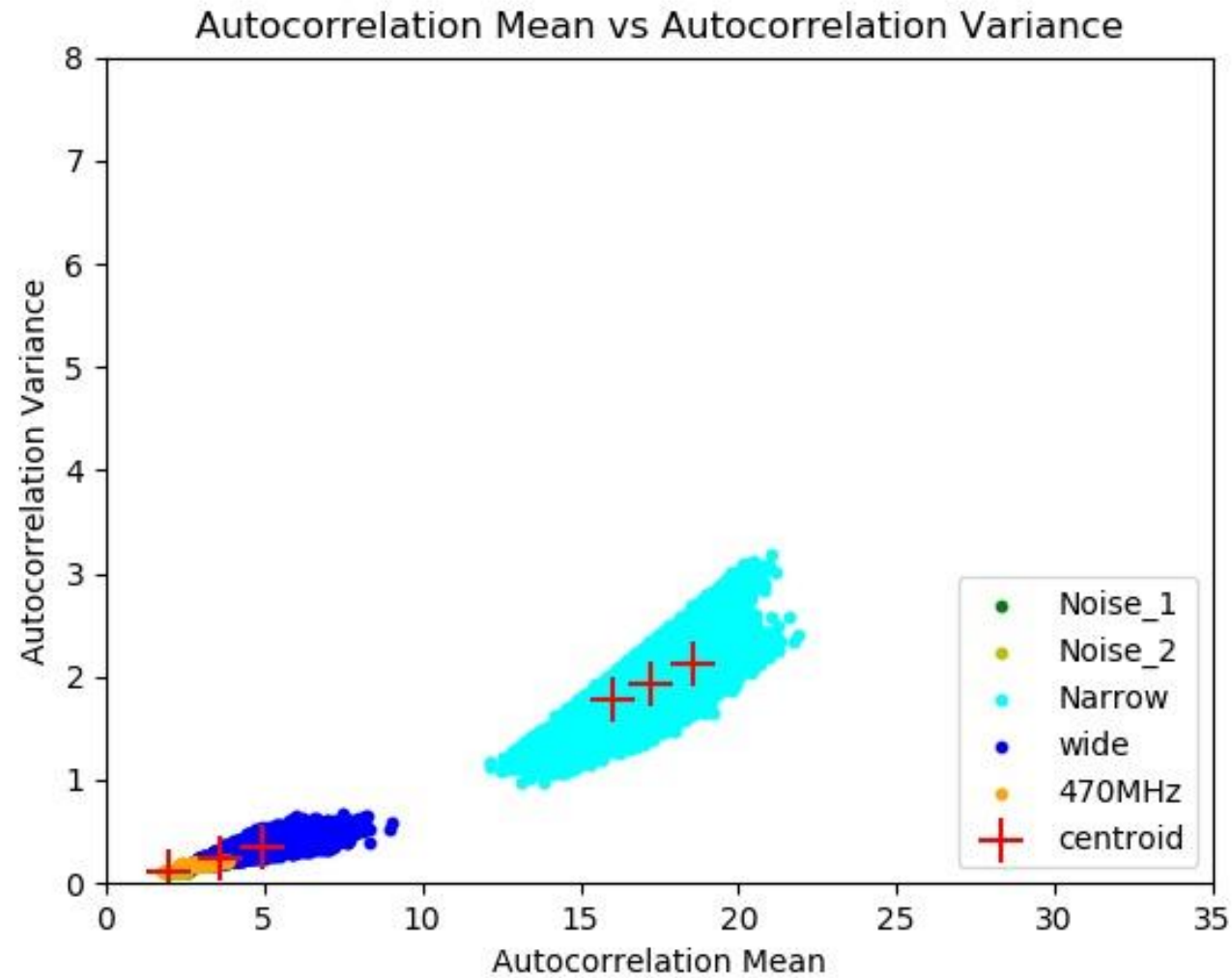
- Pick initial centroids: K samples of the dataset.
- Assign each simple to the nearest centroid.
- Calculate new centroid: mean of the cluster samples.



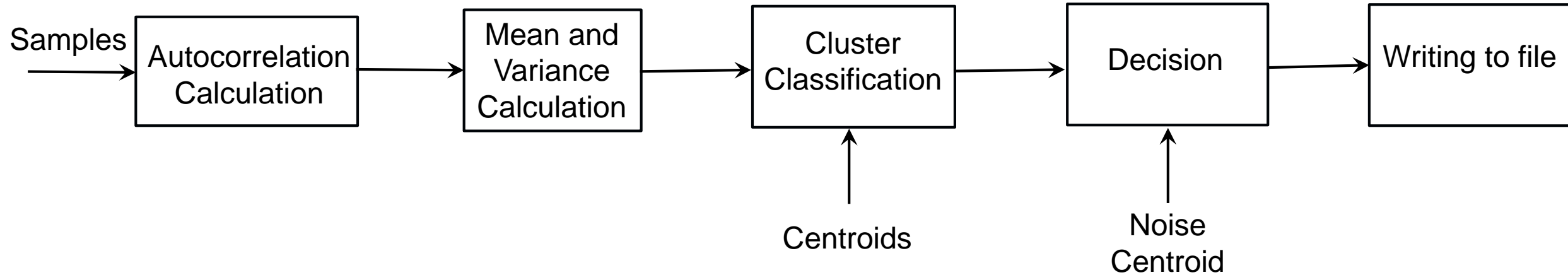
K-Means Clustering and Autocorrelation



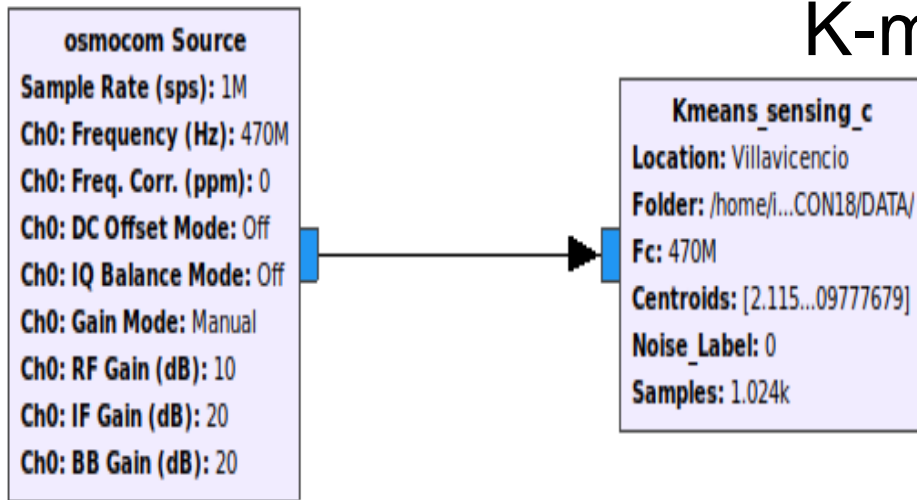
K-Means Clustering and Autocorrelation



K-means Method



K-means GNU Radio Block



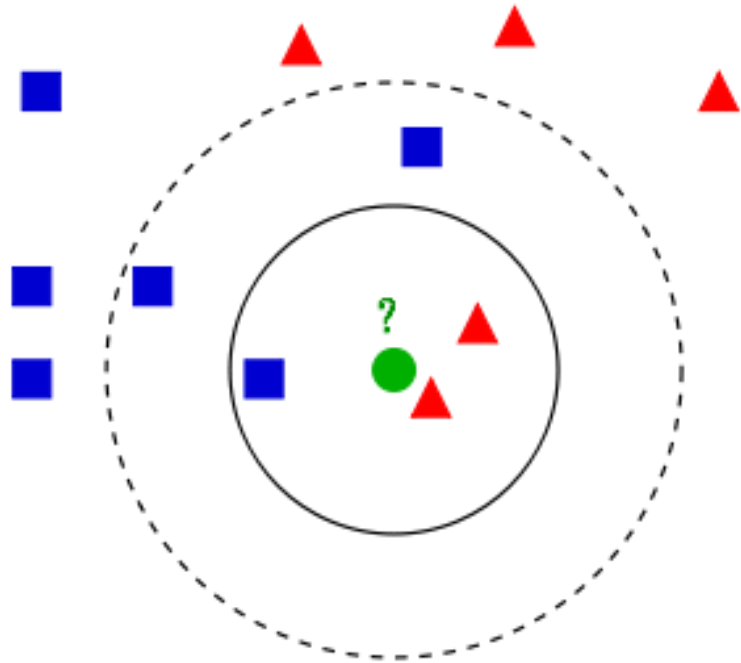
Properties: Kmeans_sensing_c

General Advanced Documentation

| | |
|-------------|---|
| ID | grcon18_Kmeans_sensing_c_0 |
| Location | "Villavicencio" |
| Folder | "/home/ivan/Documents/GR_CON18/DATA/" |
| Fc | frequency |
| Centroids | ([[[2.11531228, 0.11762681], [13.25689781, 1.06694517], [6.86376845, 0.74284188], [27.96604178, 3.92538267], [16.69717163, 1.29444081], [10.19791648, 1.09777679]])]) |
| Noise_Label | 0 |
| Samples | 1024 |

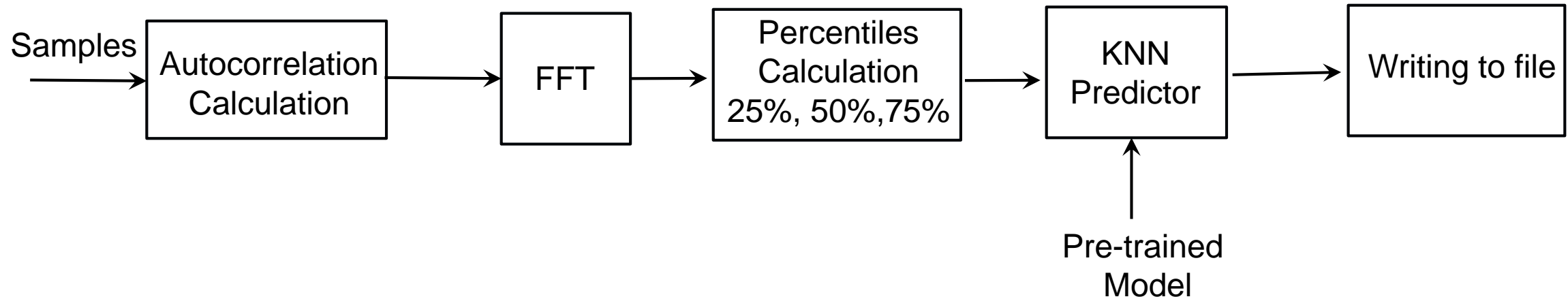
KNN Method

K- nearest neighbor method

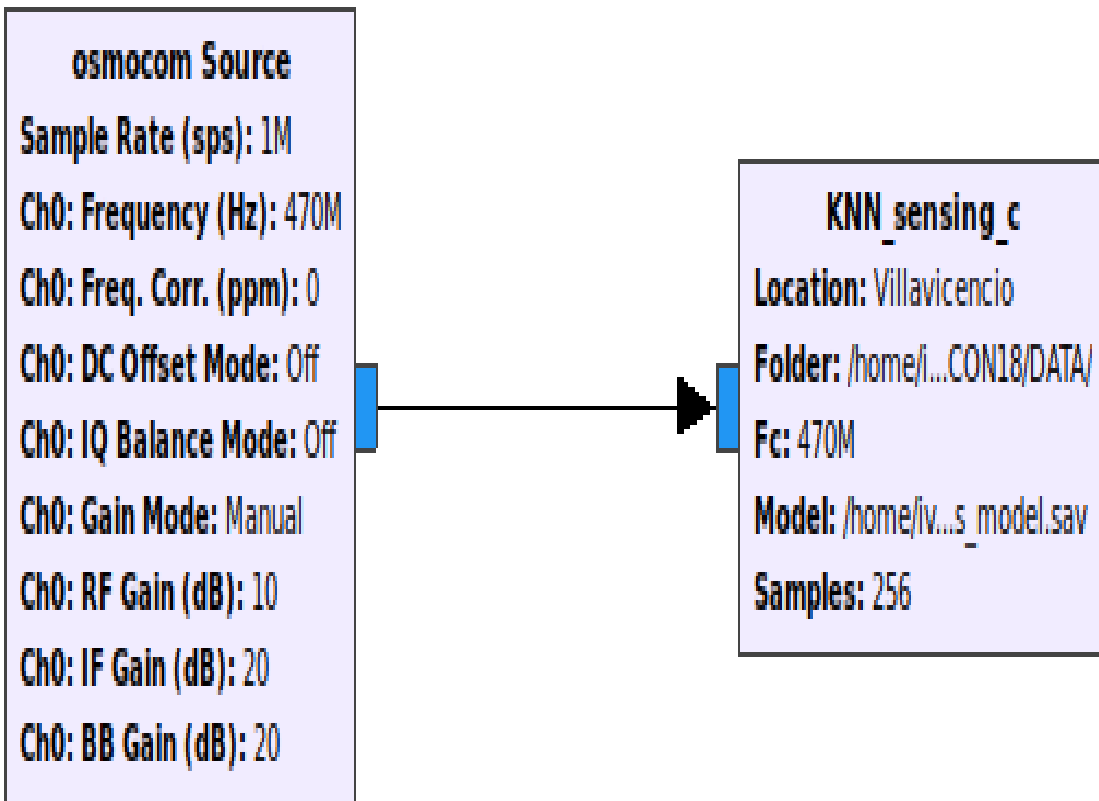


- Training samples with known labels.
- A new sample is given the most common label among its K nearest neighbors.

KNN Method



KNN GNU Radio Block

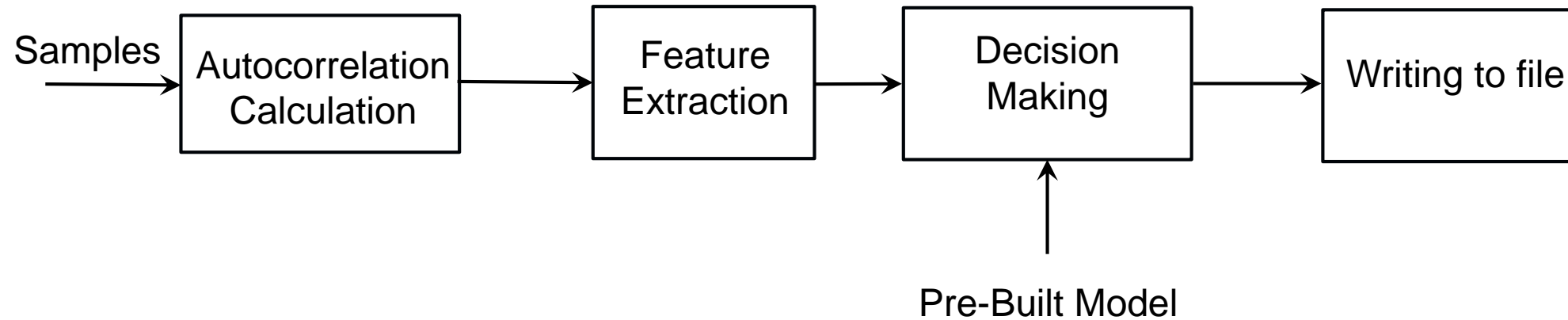


Properties: KNN_sensing_c

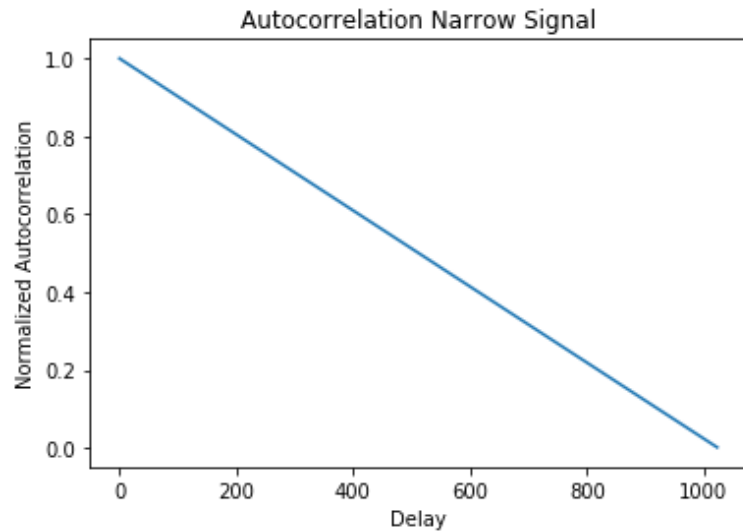
General Advanced Documentation

| | |
|----------|---|
| ID | grcon18_KNN_sensing_c_0 |
| Location | "Villavicencio" |
| Folder | "/home/ivan/Documents/GR_CON18/DATA/" |
| Fc | frequency |
| Model | "/home/ivan/Documents/GR_CON18/Model/quantiles_model.sav" |
| Samples | 256 |

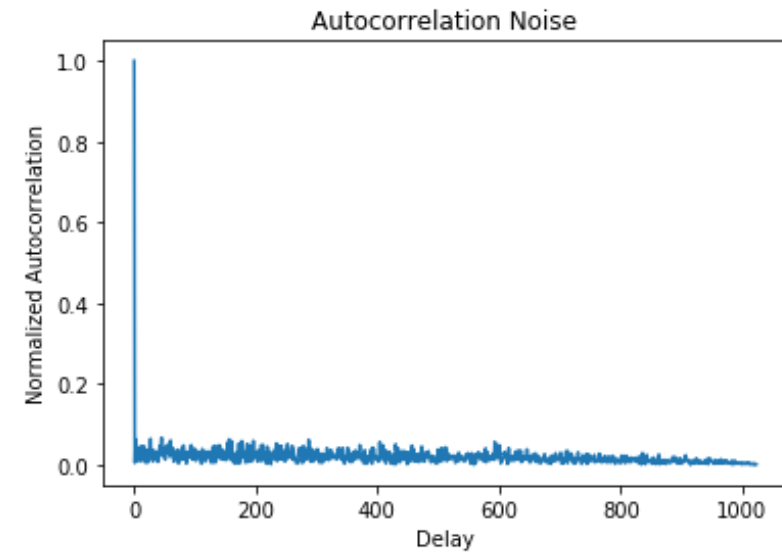
What all the methods have in common



Sometimes it is easy to extract the features



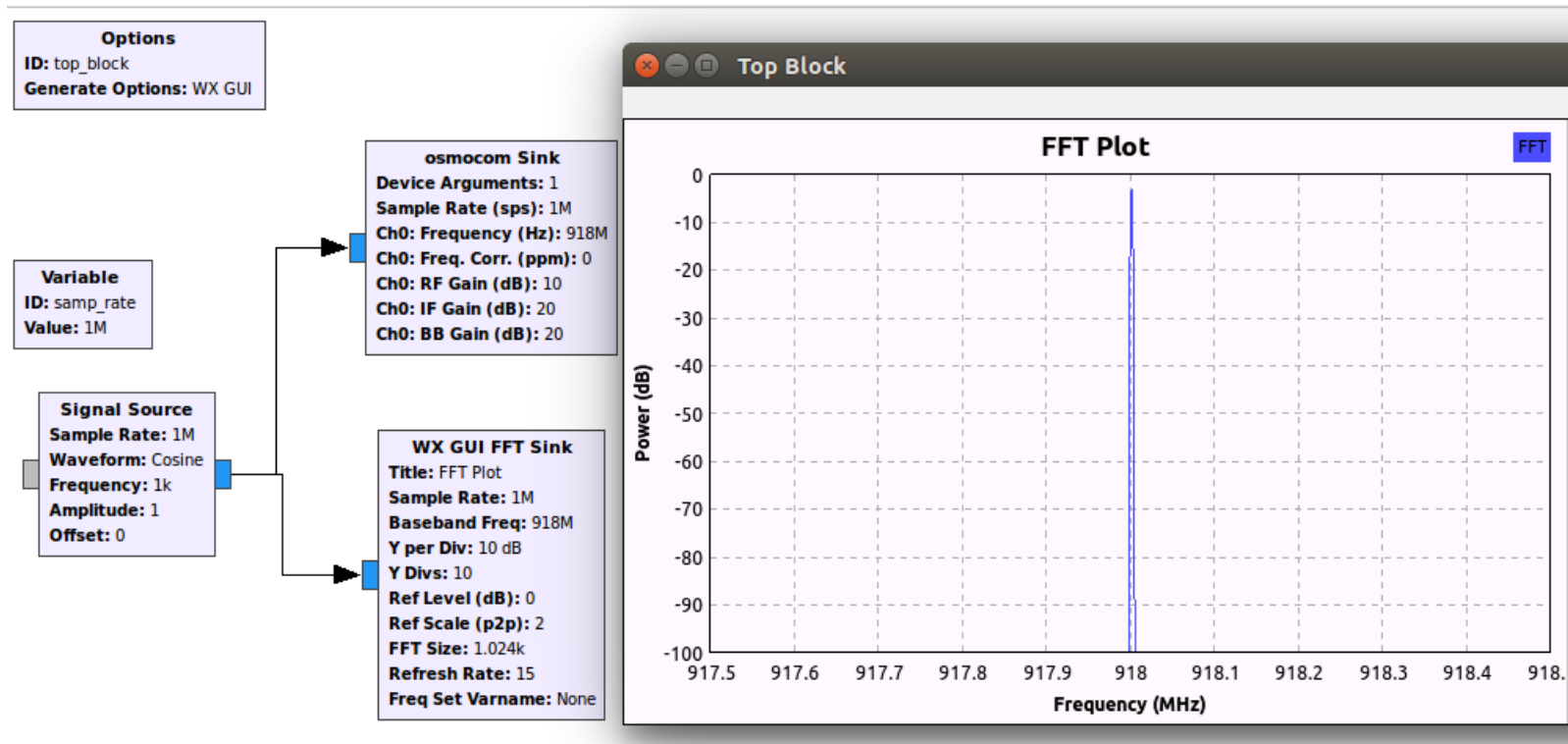
Hack RF + GNU Radio signal



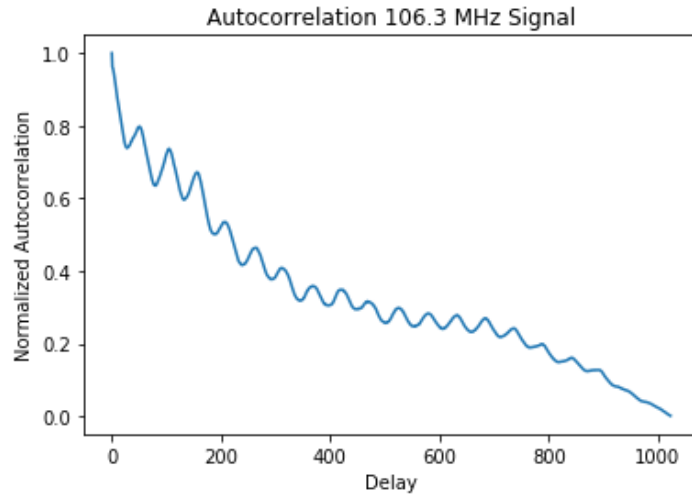
Noise



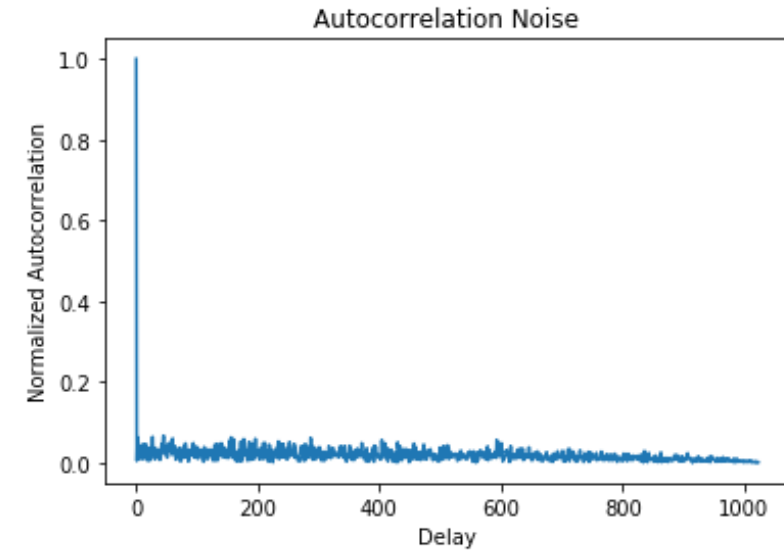
Scheme used for generating a narrow signal



Sometimes it is easy to extract the features



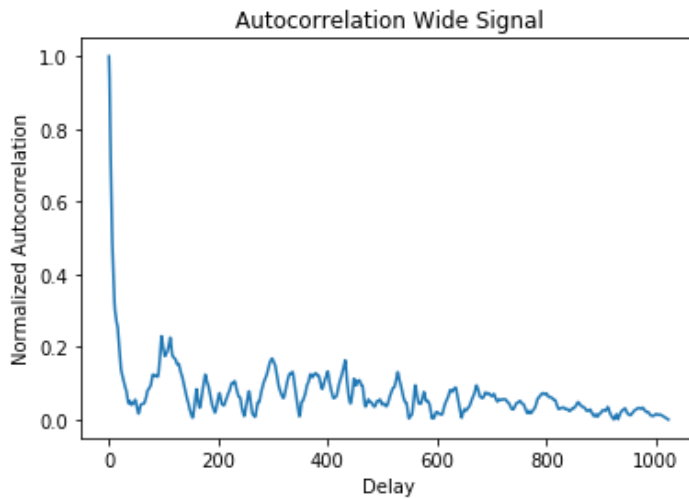
Hack RF + GNU Radio signal



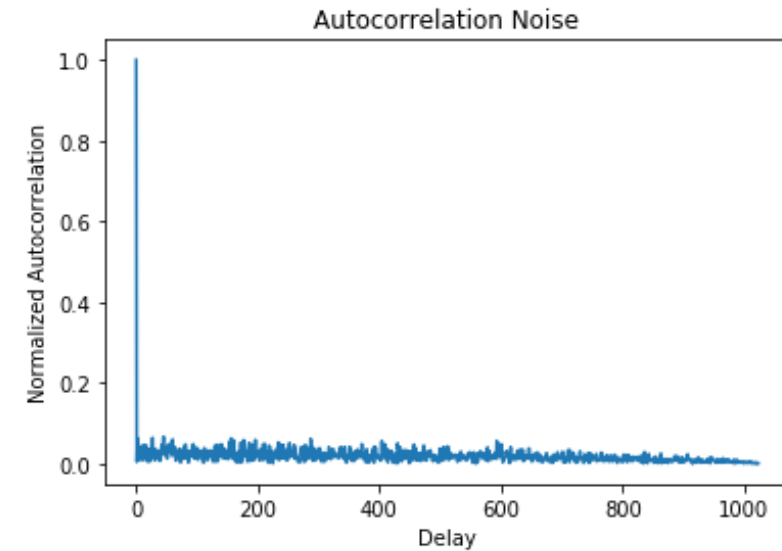
Noise



Sometimes it is a little bit difficult



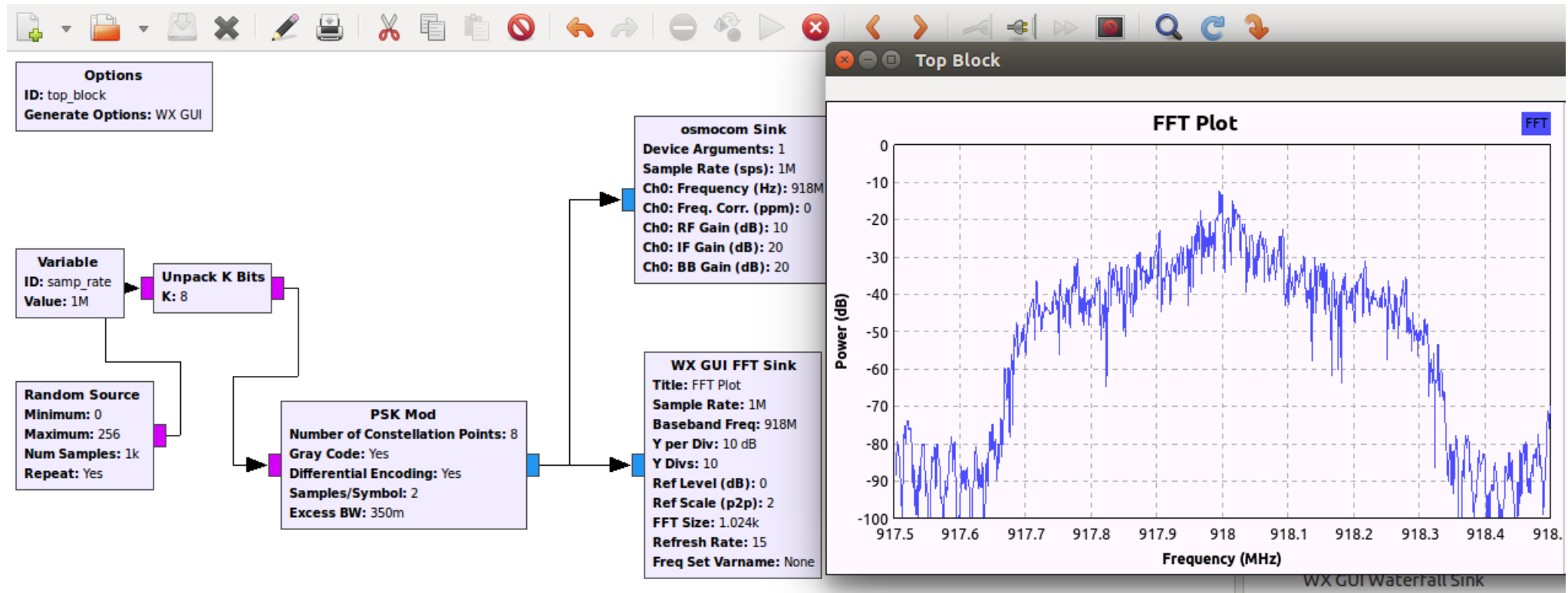
Hack RF + GNU Radio signal



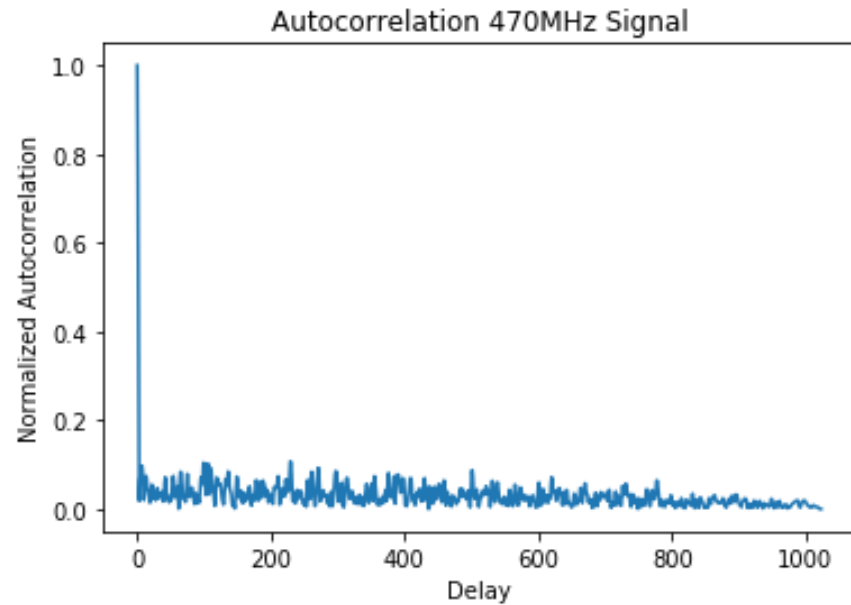
Noise



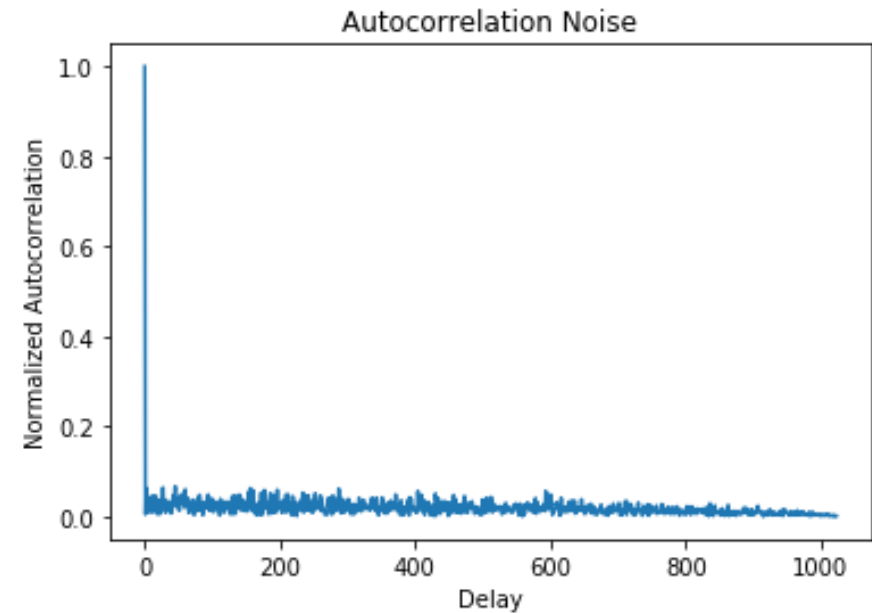
Scheme used for generating a wide signal



Sometimes it is much more difficult



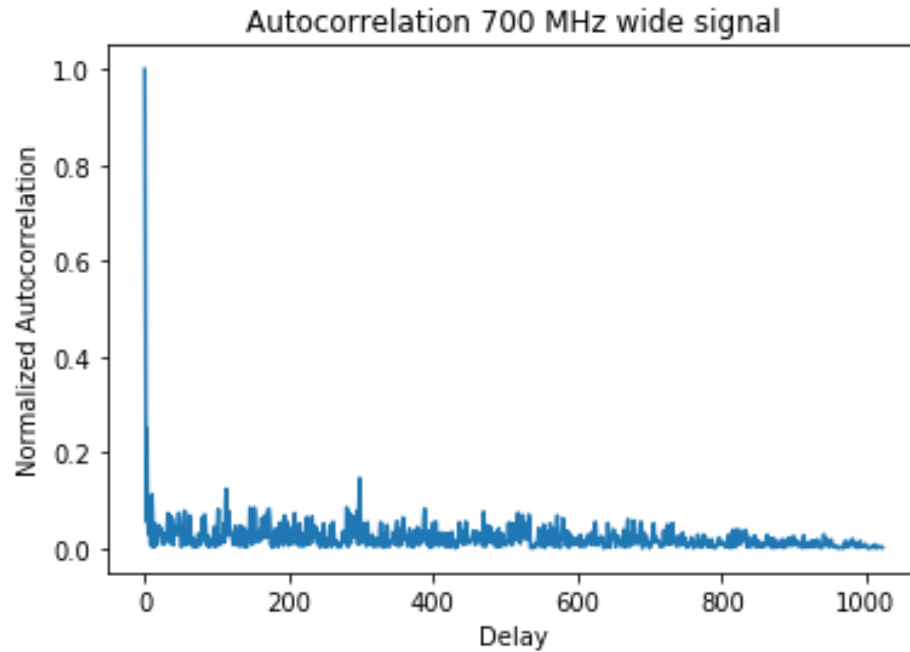
470 MHz signal (Digital TV)
taken from the air



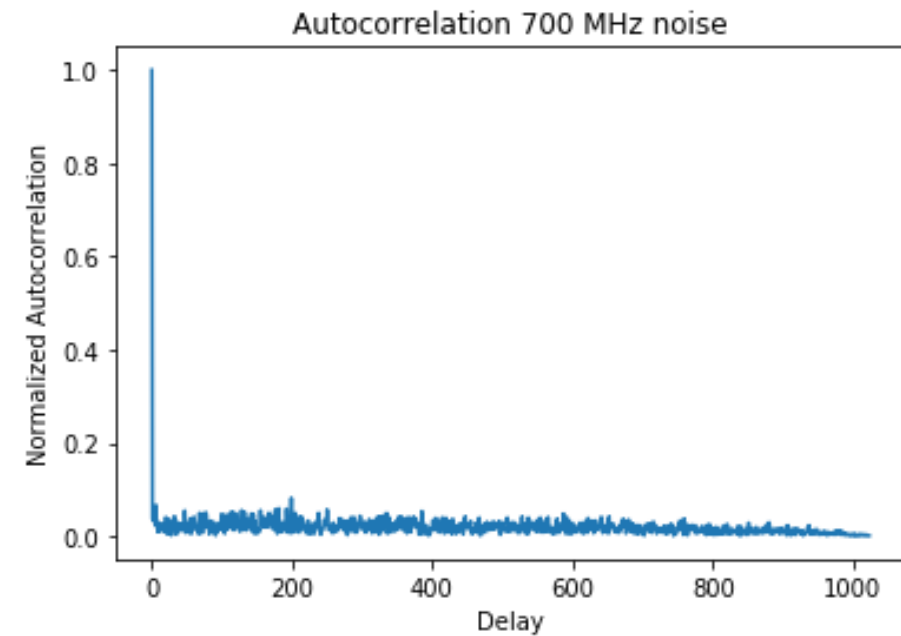
Noise



Sometimes it is a much more difficult



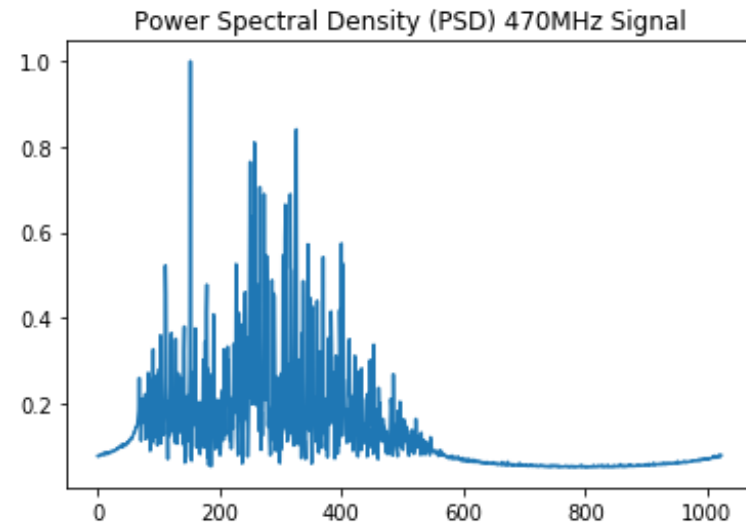
Hack RF + GNU Radio signal



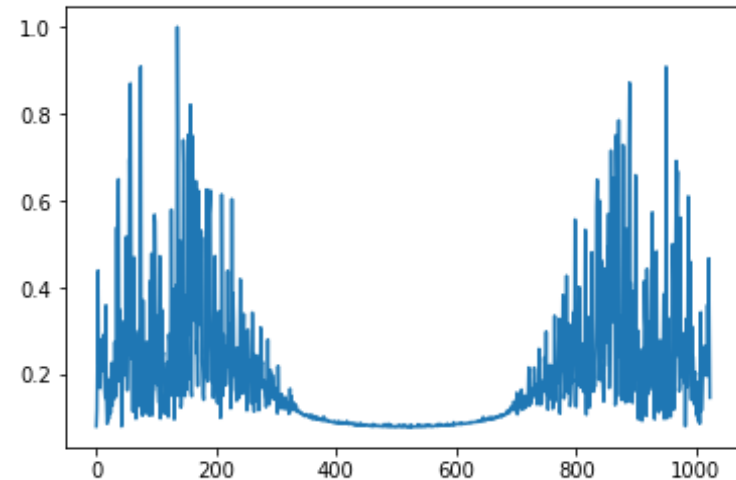
Noise



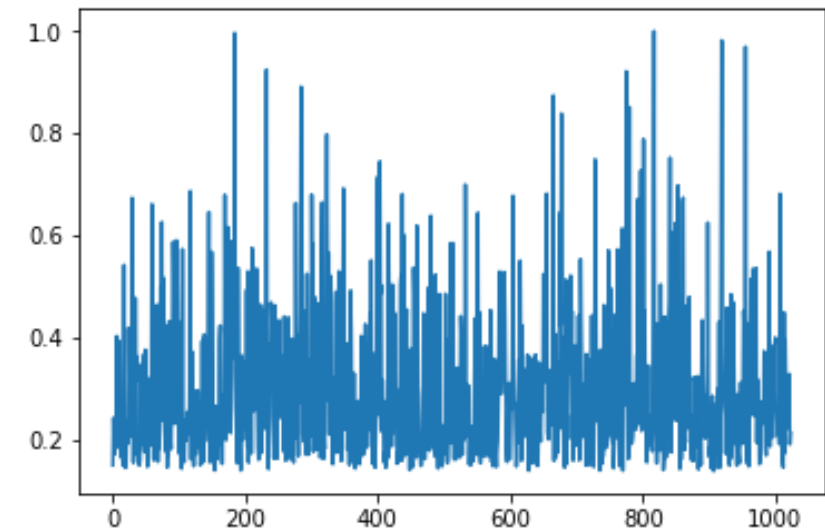
Taking the FFT of the autocorrelation makes the features more visible



470 MHz signal (Digital TV) taken from the air

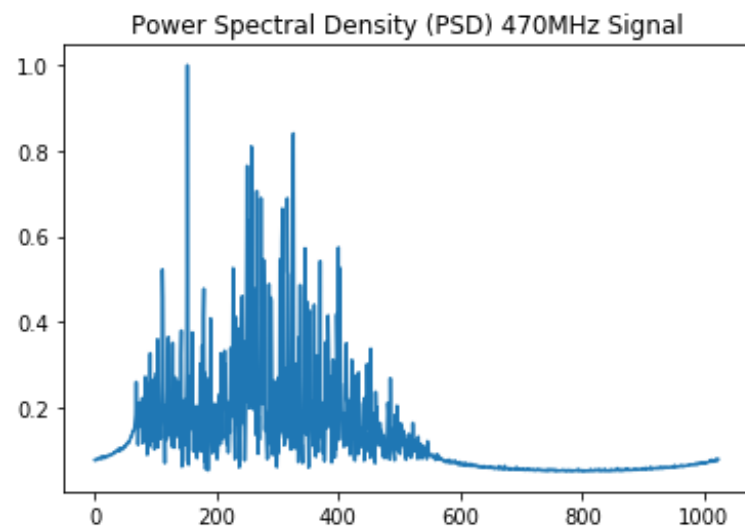


Hack RF + GNU Radio signal

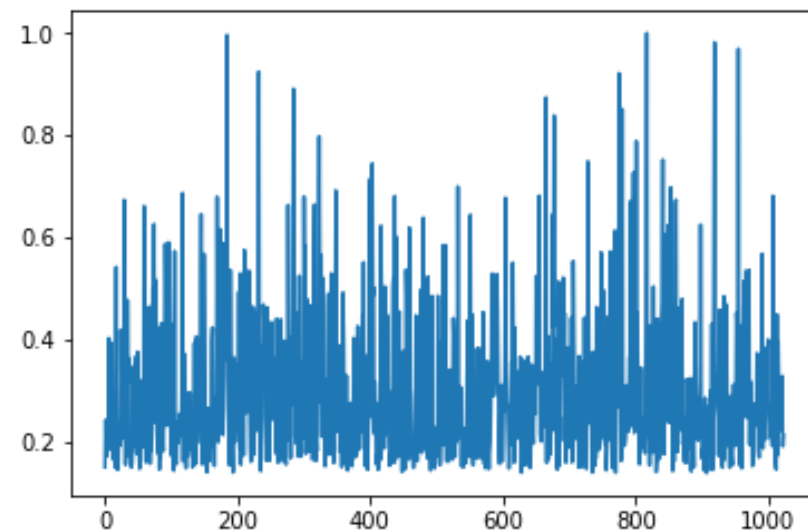


Noise

25%, 50% , 75% percentiles as features



470 MHz signal (Digital TV) taken from the air



Noise

| | p_25 | p_50 | p_75 | PU |
|---------------|----------|----------|----------|-----|
| 119221 | 0.088163 | 0.119343 | 0.193114 | 1.0 |
| 119222 | 0.067656 | 0.094399 | 0.155456 | 1.0 |
| 119223 | 0.116526 | 0.159822 | 0.233984 | 1.0 |
| 119224 | 0.091226 | 0.121970 | 0.190987 | 1.0 |

| | p_25 | p_50 | p_75 | PU |
|----------|----------|----------|----------|-----|
| 0 | 0.140444 | 0.188720 | 0.258878 | 0.0 |
| 1 | 0.181953 | 0.248563 | 0.342750 | 0.0 |
| 2 | 0.147883 | 0.193947 | 0.270266 | 0.0 |
| 3 | 0.176191 | 0.227200 | 0.330292 | 0.0 |

Tests

| | Narrow Signals | | Wide Signals | |
|-----------|----------------|-----------|--------------|-----------|
| Method | False Alarm | Detection | False Alarm | Detection |
| Euclidean | 0% | 100% | 0% | 80,4% |
| Kmeans | 0% | 100% | 2,3% | 95,1% |
| KNN | 0% | 100% | 1,6% | 96,2% |



Conclusions and Future Work

- The KNN method performs better than the Euclidean and K-means method; the more training data the better the performance.
- We can see the autocorrelation and the PSD as images and analyze them by means of machine- learning algorithms.
- Pre-trained models can be shared to be re-used for spectrum sensing.
- Future work includes solving the problem of overflowing (OOOOOO) when running the flowgraph.
- Future work also includes working on multi-label classification using the KNN method.
- Future work also includes calibrating the hyper-parameters of the model for better performance; trying other percentiles or looking for other features to exploit.

Thanks !

hreyes@unillanos.edu.co

