

Thresholding-based decision with a single sample

Table of contents

1	Objective	1
2	Theoretical aspects	1
2.1	Decision with a single sample	1
2.2	Matlab snippets and code	2
	Generating random binary data	2
2.3	Counting values of 1 from a binary vector	2
3	Exercises	3
4	Final questions	4

1 Objective

Simulate a binary constant-signal detection system based on thresholding, and characterize the system via the Receiver Operating Characteristic.

2 Theoretical aspects

2.1 Decision with a single sample

Consider a binary message encoded with two constant levels 0 and A , affected by white noise. Taking a single sample of the signal we obtain a value r :

$$r = s + n,$$

where s is the true signal value (0 or A), and n is the sample of noise.

The receiver decides what is the true signal by comparing the sample r with a threshold T , whose value depends on the specific decision criterion used:

- Maximum Likelihood
- Minimum error probability
- etc.

For one decision, there can be **four outcomes**:

- **correct rejection**: signal is 0, detection is 0
- **false alarm**: signal is 0, detection is 1
- **miss**: signal is 1, detection is 0
- **hit** (correct detection): signal is 1, detection is 1

The Receiver Operating Characteristic (ROC) curve is the plot of the probability $P(\text{hit})$ against the probability $P(\text{false alarm})$, for all possible values of T .

2.2 Matlab snippets and code

Generating random binary data

In Matlab, we can generate a vector randomly filled with 0's and 1's in the following way:

- We use the function `rand()` to generate a vector with random floating point numbers between 0 and 1
- We compare the vector with some constant. The comparison result will be 0's and 1's, which are placed randomly.
- If we compare with 0.5, we get an equal amount of 0's and 1's (equal probability).
- In general, if we compare with some threshold $p \in [0, 1]$ we get 0's and 1's with probability p and $1 - p$, respectively.

```
% Generate a random vector with 25\% 0's and 70\% 1's  
v = rand(1,10000) > 0.25;
```

2.3 Counting values of 1 from a binary vector

If we have a vector containing only 0's and 1's, we can count the number of 1's by simply summing the vector.

```
% Count how many 1's are in the vector v generated above
count = sum(v);
```

3 Exercises

1. Simulate threshold-based detection with a single sample, as follows:
 - Generate a vector of 100000 values 0 or A , with equal probability (consider $A = 5$)
 - Add over it a random noise with normal distribution $\mathcal{N}(0, \sigma^2 = 1)$
 - Pick a value of T between 0 and A , and compare each element with T to decide which sample is logical 0 or logical 1 (A)
 - Compare the decision result with the true original vector, and count how many correct detections and how many false alarms have been
 - Estimate $P(\text{hit})$ and $P(\text{false alarm})$ by dividing the above numbers to the size of the vector
2. Wrap the above code into a function `[phit, pfa] = myThreshDet(T)` that returns the two probabilities for a given T . Draw the ROC by running the function for 100 values of T uniformly spaced between 0 and A , and plotting the resulting vector `phit` against `pfa`.
3. Repeat the simulation for BPSK modulation, as follows.
 - Generate a vector of 100000 values 0 or A , with equal probability (consider $A = 5$)
 - Generate a signal `p` with 1 period of a sinusoidal signal, of length 100 samples.
 - Generate the BPSK modulated signal `s`. Use `kron()` for this, explain at whiteboard.
 - Add over it a random noise with normal distribution $\mathcal{N}(0, \sigma^2 = 1)$. Call the result `r`.
 - Plot the signals `s` and `r` on the same figure (original signal and noisy signal)
 - Do BPSK de-modulation as follows:


```
rs = p * reshape(r, 100, []);
```
 - The vector `rs` is similar to the `r` vector from the previous exercise. Continue in the same way as in the previous exercise.
4. Repeat the same simulation in Exercise 1 for two samples per bit:
 - double the values of the starting vector, making two consecutive 0 or A values, e.g.

[00AA00AAAA00AA...]

- the decision now uses **the average value** of the two consecutive samples of a bit
- plot the ROC and compare with the first one. Which is better?

4 Final questions

1. Suppose we make we start with a vector with much fewer 1's than 0's. Should we increase or decrease the threshold T ?
2. In a practical scenario, what could be a disadvantage of using 2 samples for detection, compared to just 1?