

# DATA ANALYSIS PROJECT

RAVURU CHIDAKSH (200010046)

`200010046@iitdh.ac.in`

IIT Dharwad

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# 1 Approach towards Problem 1

## 1.1 Calculating mean and variance:

From the given dataset we know that first 1000 data were mixed with the value of random variables which we wantendly assigned them to 0. So, by calculating the **mean** of the first 1000 values in the dataset we get the mean of the noise data. We can also calculate **variance** for the first 1000 samples in our dataset.

So now we know the mean and variance of noise , we can calculate **mean** for the remaining data say **corrupted\_toss\_data** and also the **variance** of the same data.

## 1.2 Code explained briefly:(explanation is written as comments)

```
1 import numpy as np
2 import pandas as pd
3 #we are reading data from excel file using read_excel
  function into a dataframe or a series object.
4 series = pd.read_excel("comp_1_200010046.xls",header=None)
5 #we know first 1000 data are from noise combined with random
  variable (whose value is 0).
6 #hence we can charecterize the noise using first 1000 values
  of the series object.
7 noise = series.values[:1000]
8 # from the above noise we can find the mean and all required
  to charecterize the noise.
9 currrupted_toss_data = series.values[1001:]
10 #which contains noise with the value of random variable
11 # so we are basically finding expected value of the random
  variable by ,
12 #E[X] = E[currrupted_toss_data] - E[noise]
13 #we can find bias by dividing by,
14 #E[X] = (0* num_tails + 5*num_heads)/10000
15 #bias = num_heads/10000
16 bias = (np.mean(currrupted_toss_data) - np.mean(noise))/5
17 # same as E[X] = 0*(1-p) + 5*p; (where p = bias)
18 #please check the .ipynb file for the clear and complete code
```

Listing 1: Calculating bias

## 2 Calculating Error from the calculated bias:

We discuss about the error because sometimes even when the get a toss and transmit 0, the receiver might receive that signal as head due to noise.

We get error in two cases when we toss a head but receiver receives as a tail and vice-versa. So, we need to calculate both the errors and we can do that with the given information that thy noise is guassian.

Since when we send a tail we get  $N+S$  (where  $S$  is signal) as the same guassian distribution with the expected value  $\mu$  ( say  $\mu$  is the expected value of noise) and when  $S = 5$  the expected value will be  $\mu + 5$  but still remains guassian.

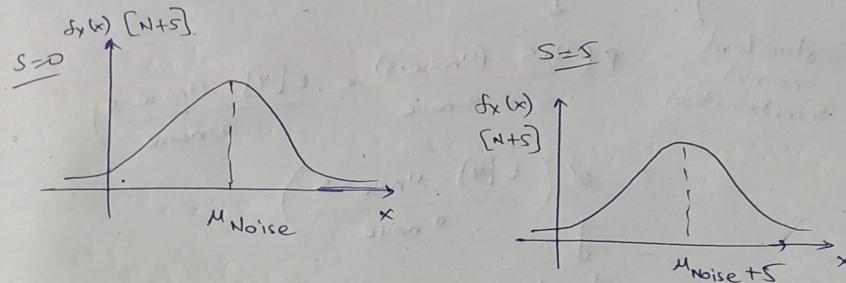
Hence, we can treat the recieving signal as head if  $N+S \geq E[\text{currrupted\_toss\_data}]$  as a head and the data where  $N+S < E[\text{currrupted\_toss\_data}]$  as tail.(corrupted\\_toss\\_data is defined in [section1.2](#))

Since we know the guassian graph , it's like finding area after the certain point and adding it up with the help of  [\$\phi\$  - table](#).

### 3 Images explaining calculation of error

Finding error,

when  $N+S \geq \text{computed\_data\_mean} \rightarrow E[Y]$  (say)



$P(\text{Error})$  where toss is received as head

$$\Rightarrow P(N+0 \geq E[Y])$$

$$= 1 - P(N \leq E[Y])$$

$$= 1 - P\left(\frac{N - \mu_{\text{noise}}}{\sigma_{\text{noise}}} \leq \frac{E[Y] - \mu_{\text{noise}}}{\sigma_{\text{noise}}}\right)$$

standard normal transformation

$$= 1 - \phi\left(\frac{E[Y] - \mu_{\text{noise}}}{\sigma_{\text{noise}}}\right)$$

standard normal CDF

If head is received as tail,

$$P(N+S < E[Y])$$

standard  
normal  
transformation

$$P\left(\frac{N+S - (\mu_{\text{noise}} + S)}{\sigma_{\text{noise}}} < \frac{E[Y] - \mu_{\text{noise}} - S}{\sigma_{\text{noise}}}\right)$$

$$= \Phi\left(\frac{E[Y] - \mu_{\text{noise}} - S}{\sigma_{\text{noise}}}\right)$$

standard normal CDF

And the total error will be sum of both the errors , and with the values calculated from the dataset we can substitute the values and get the total error and error percentage. In my case the net error was around **0.0018**. Around 0.18%.

## 4 Note:

- I wrote the code in python in Google Colab , which i'm submitting with extension "ipynb". So, to run the entire file , please upload the file on drive and open it with google colabaratory.
- I worked with .xls file in the entire process which i'm going to zip it with this pdf file . Please upload that .xls file on Google colab before running the cells.