

Lecture 19

For example : $X: N(0, \sigma_x)$, $Y: N(0, \sigma_y)$, $Z = X + Y$, X and Y are independent.

Question: $f_z(z) = ?$

Answer:

$$X: N(0, \sigma_x) \longrightarrow \phi_x(u) = e^{-\frac{u^2 \sigma_x^2}{2}}$$

$$Y: N(0, \sigma_y) \longrightarrow \phi_y(u) = e^{-\frac{u^2 \sigma_y^2}{2}}$$

$$\begin{aligned} Z: \phi_z(u) &= \phi_x(u) \cdot \phi_y(u) = e^{-\frac{u^2 \sigma_x^2}{2}} \cdot e^{-\frac{u^2 \sigma_y^2}{2}} \\ &= e^{-\frac{u^2 (\sigma_x^2 + \sigma_y^2)}{2}} \end{aligned}$$

↓

$$Z: N(0, \sqrt{\sigma_x^2 + \sigma_y^2})$$

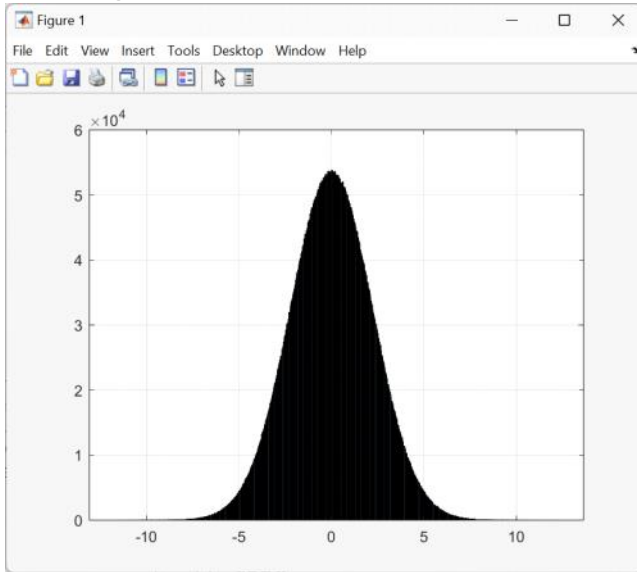
```
>> X = randn(1, 10000000);  
>> Y = 2*randn(1, 10000000);  
>> Z = X+Y;  
>> var(Z)  
ans =  
4.9996
```

```
>> std(Z)  
ans =  
2.2360
```

```
>> var(X) + var(Y)  
ans =  
4.9999
```

```
>> sqrt(var(X) + var(Y))  
ans =  
2.2360
```

>> histogram(Z)



Normality Check:

```
>> S = find(abs(Z - mean(Z)) <= std(Z));
```

```
>> length(S)/length(Z)
```

```
ans =
```

```
0.6826
```

```
>> S = find(abs(Z - mean(Z)) <= 2*std(Z));
```

```
>> length(S)/length(Z)
```

```
ans =
```

```
0.9545
```

```
>> S = find(abs(Z - mean(Z)) <= 3*std(Z));
```

```
>> length(S)/length(Z)
```

```
ans =
```

```
0.9973
```

In general, if $Y = X_1 + X_2 + \dots + X_n$

$$\begin{matrix} \downarrow & \downarrow & \downarrow & \downarrow \\ \phi_Y(u) = \phi_{X_1}(u) \cdot \phi_{X_2}(u) \cdot \dots \cdot \phi_{X_n}(u) \end{matrix}$$

Chapter 4: Elements of Statistics

Sampling Theory: Selecting samples from a collection of data that is too large to be examined individually.

Question: determine how many samples are required for a given degree of confidence in the result.

Population (N): Collection of data being studied.

Sample Size (n), where sample is part of the population selected at random

- Sample Mean

X_1, X_2, \dots, X_n random variables (all having the same \bar{X})

Define Sample Mean as:

$$\hat{\bar{X}} = \frac{1}{n} \sum_{i=1}^n X_i$$

↑
True Mean

Define sample mean as:

$$\hat{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

↑
also a RV

True Mean

$$E[\hat{X}] = E\left[\frac{1}{n} \sum_{i=1}^n X_i\right] = \frac{1}{n} \sum_{i=1}^n E[X_i] = \frac{1}{n} \cdot \sum_{i=1}^n \bar{X} = \frac{1}{n} \cdot n \cdot \bar{X} = \bar{X}$$

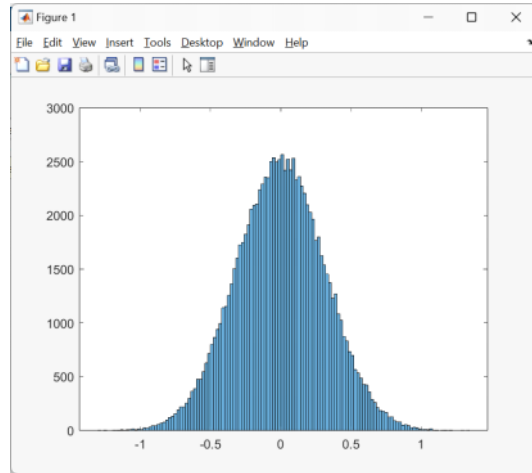
True Mean

Mean of the Sample Mean = True Mean

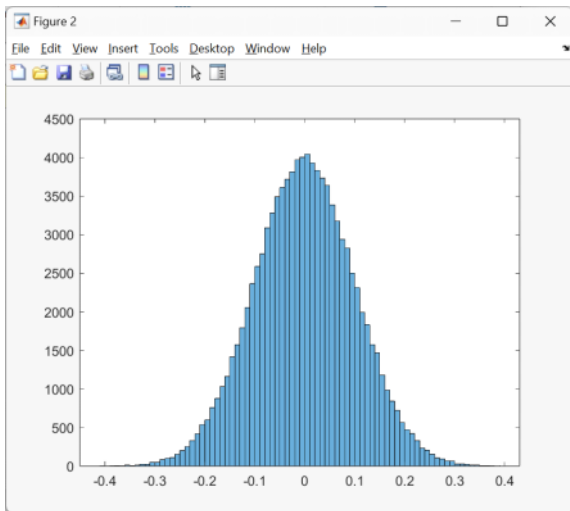
<http://www.ece.uah.edu/~dwpan/course/ee385/code/>

`% sample_mean.m`

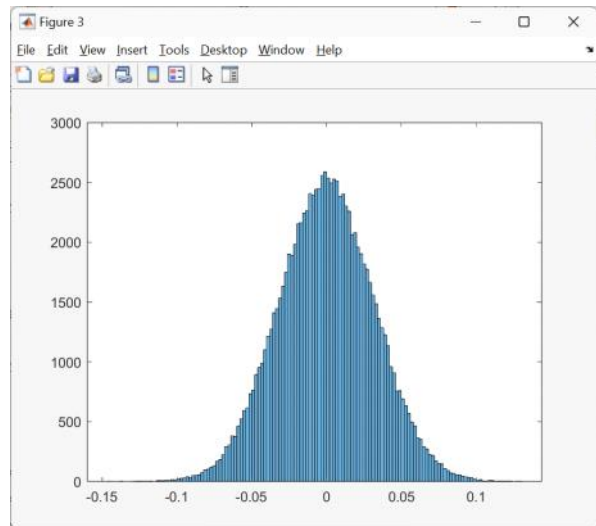
```
sample_size = 10
run = 100000
Xavg = zeros(1, run);
for i = 1: run
    X = randn (1, sample_size);
    Xavg(i) = mean (X);
end
disp('E[X_bar] =');
mean(Xavg)
disp('Var[X_bar] =');
var(Xavg)
```



sample_size = 10, Var[X_bar] = 0.0995
Mean[X_bar] = -0.0015



sample_size = 100, Var[X_bar] = 0.0100



sample_size = 1000, Var[X_bar] = 0.0010

- Variance of the Sample Mean
 - (1) Sample with Replacement: $N \gg n$
 - (2) Sample without Replacement: N is small