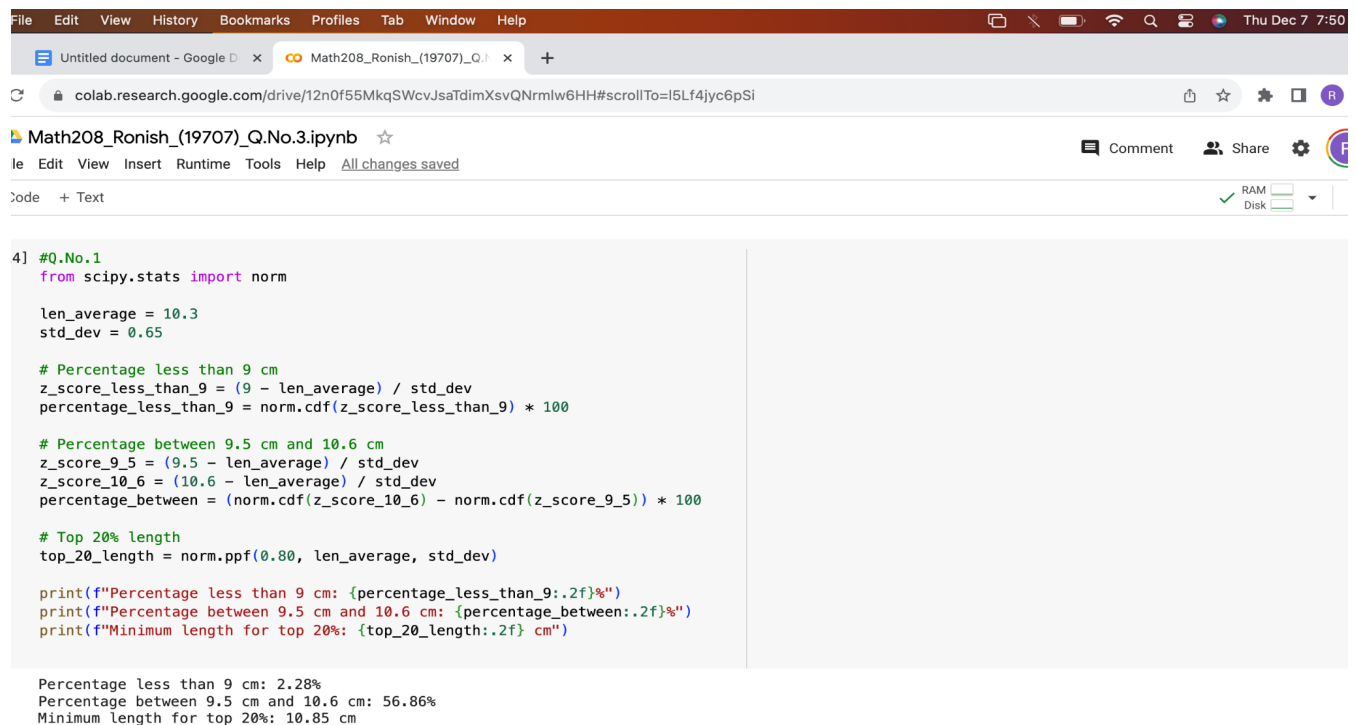


Ronish Shrestha (19707)
Assignment 4

Q.No.1) Assuming that the lengths of American anchovies appease the normal distribution with the mean $\mu = 10.3\text{cm}$ and standard deviation $\sigma = 0.65\text{cm}$, please find the percentages of the lengths in the population of American anchovies. a. Less than 9cm. b. Between 9.5cm and 10.6cm. c. What is the minimum length if a restaurant claimed that the lengths of the sold anchovies are in the top of 20%?

Solution:



```
4] #Q.No.1
from scipy.stats import norm

len_average = 10.3
std_dev = 0.65

# Percentage less than 9 cm
z_score_less_than_9 = (9 - len_average) / std_dev
percentage_less_than_9 = norm.cdf(z_score_less_than_9) * 100

# Percentage between 9.5 cm and 10.6 cm
z_score_9_5 = (9.5 - len_average) / std_dev
z_score_10_6 = (10.6 - len_average) / std_dev
percentage_between = (norm.cdf(z_score_10_6) - norm.cdf(z_score_9_5)) * 100

# Top 20% length
top_20_length = norm.ppf(0.80, len_average, std_dev)

print(f"Percentage less than 9 cm: {percentage_less_than_9:.2f}%")
print(f"Percentage between 9.5 cm and 10.6 cm: {percentage_between:.2f}%")
print(f"Minimum length for top 20%: {top_20_length:.2f} cm")

Percentage less than 9 cm: 2.28%
Percentage between 9.5 cm and 10.6 cm: 56.86%
Minimum length for top 20%: 10.85 cm
```

- a. Less than 9 cm = 2.28%
- b. Between 9.5 and 10.6 cm = 56.86%
- c. Min length for top 20% = 10.85cm

Q.No.2) If the random variables X and Y are normal distributions with $\mu = 10$ & $\sigma = 3$ and $\mu = 15$ & $\sigma = 8$, namely, $X \sim N(10, 3)$ and $Y \sim N(15, 8)$, and they are independent, what is the probability distribution and statistical parameters of (1) $X + Y$ (2) $X - Y$ (3) $3X$ (4) $4X + 5Y$

Solution:

If X and Y are independent normal distributions with means (μ) and standard deviations (σ) as described, we can calculate the probability distribution and statistical parameters for the following combinations:

(1) $X + Y$:

Explanation:

The sum of two independent normal distributions is also a normal distribution. The mean of the sum is the sum of the means, and the variance of the sum is the sum of the variances.

$$\text{Mean } (\mu_{X+Y}) = \mu_X + \mu_Y = 10 + 15 = 25$$

$$\text{Variance } (\sigma^2_{X+Y}) = \sigma^2_X + \sigma^2_Y = 3^2 + 8^2 = 9 + 64 = 73$$

$$\text{SD} = \sqrt{\text{variance}} = \sqrt{73}$$

So, the probability distribution for is $(X + Y) \sim N(25, \sqrt{73})$.

(2) $X - Y$:

Similarly, the difference of two independent normal distributions is a normal distribution.

$$\text{Mean } (\mu_{X-Y}) = \mu_X - \mu_Y = 10 - 15 = -5$$

$$\text{Variance } (\sigma^2_{X-Y}) = \sigma^2_X + \sigma^2_Y = 3^2 + 8^2 = 9 + 64 = 73$$

$$\text{SD} = \sqrt{\text{variance}} = \sqrt{73}$$

So, the probability distribution for is $(X - Y) \sim N(-5, \sqrt{73})$.

(3) $3X$:

Multiplying a normal distribution by a constant scales both the mean and variance.

$$\text{Mean } (\mu_{3X}) = 3\mu_X = 3 * 10 = 30$$

$$\text{Variance } (\sigma^2_{3X}) = (3^2) \sigma^2_X = 81$$

$$\text{SD} = \sqrt{81} = 9$$

So, the probability distribution for $3X$ is $3X \sim N(30, 9)$.

(4) $4X + 5Y$:

The sum of scaled normal distributions is also a normal distribution.

$$\text{Mean } (\mu_{4X+5Y}) = 4\mu_X + 5\mu_Y = 4*10 + 5*15 = 40 + 75 = 115$$

$$\text{Variance } (\sigma^2_{4X+5Y}) = ((4.\sigma_X)^2 + (5.\sigma_Y)^2 = (4.3)^2 + (5.8)^2 = 344$$

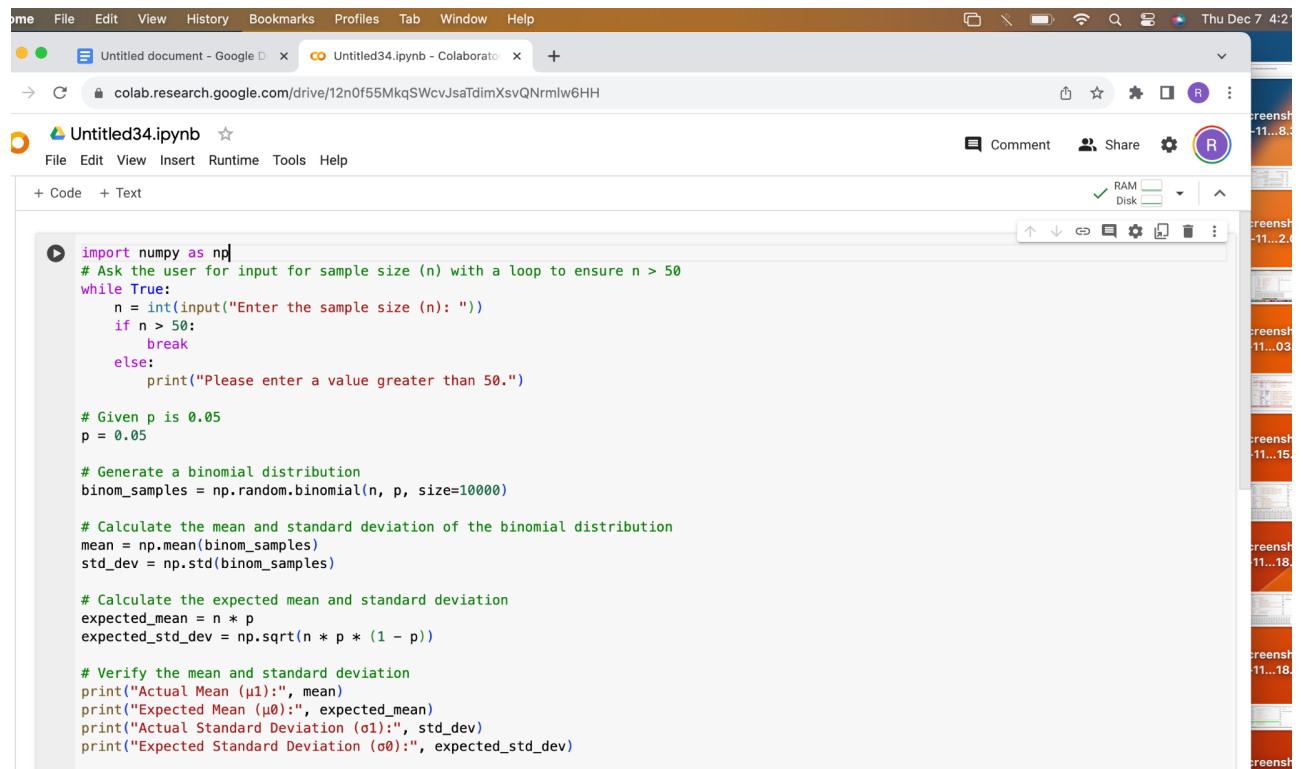
$$\text{SD} = \sqrt{344}$$

So, the probability distribution for $4X + 5Y$ is $4X + 5Y \sim N(115, \sqrt{344})$.

These are the probability distributions and statistical parameters for the given combinations of random variables X and Y .

Q.No.3) For the students in Engineering School, please write Python program to verify the mean $\mu = np$ and standard deviation $\sigma = \sqrt{npq}$ for $p=0.05$ and selecting any n greater than 50 in binomial distribution.

Solution:



The screenshot shows a Google Colab notebook titled 'Untitled34.ipynb'. The code in the first cell is as follows:

```
import numpy as np
# Ask the user for input for sample size (n) with a loop to ensure n > 50
while True:
    n = int(input("Enter the sample size (n): "))
    if n > 50:
        break
    else:
        print("Please enter a value greater than 50.")

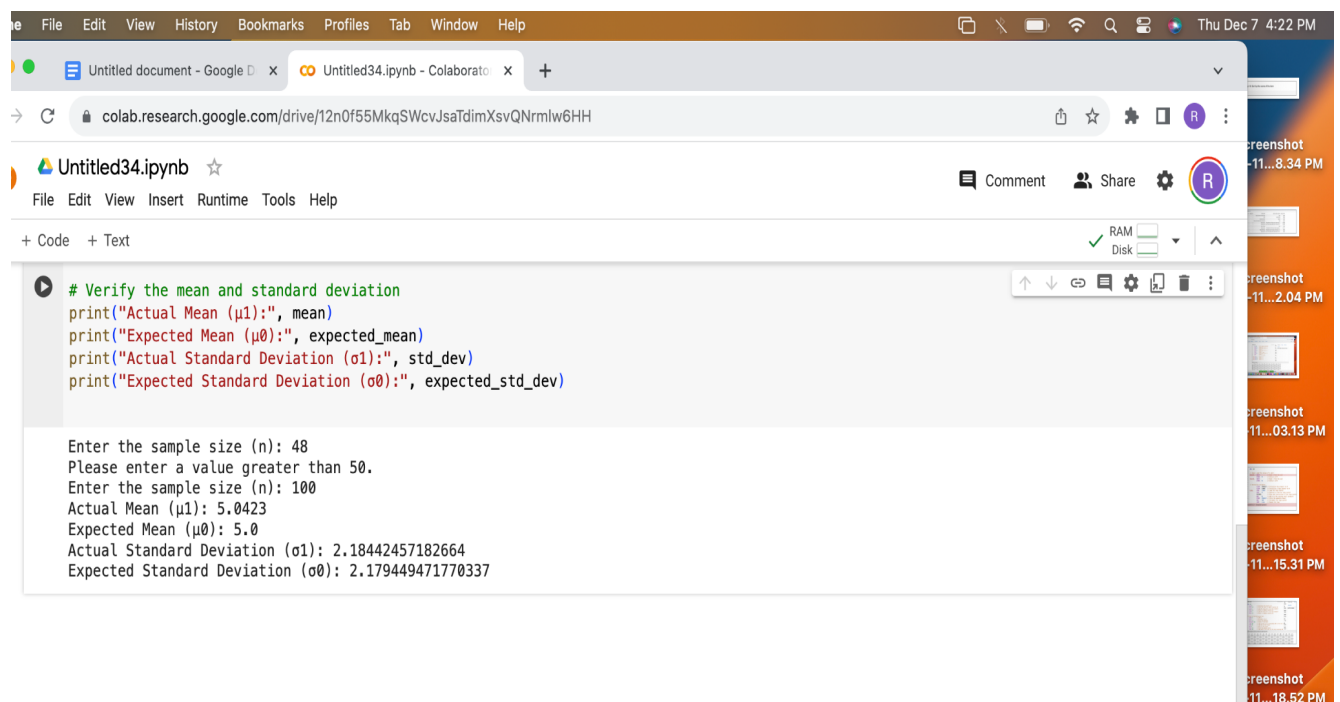
# Given p is 0.05
p = 0.05

# Generate a binomial distribution
binom_samples = np.random.binomial(n, p, size=10000)

# Calculate the mean and standard deviation of the binomial distribution
mean = np.mean(binom_samples)
std_dev = np.std(binom_samples)

# Calculate the expected mean and standard deviation
expected_mean = n * p
expected_std_dev = np.sqrt(n * p * (1 - p))

# Verify the mean and standard deviation
print("Actual Mean ( $\mu_1$ ):", mean)
print("Expected Mean ( $\mu_0$ ):", expected_mean)
print("Actual Standard Deviation ( $\sigma_1$ ):", std_dev)
print("Expected Standard Deviation ( $\sigma_0$ ):", expected_std_dev)
```



The screenshot shows the same Google Colab notebook, but now displaying the output of the code execution. The output is as follows:

```
# Verify the mean and standard deviation
print("Actual Mean ( $\mu_1$ ):", mean)
print("Expected Mean ( $\mu_0$ ):", expected_mean)
print("Actual Standard Deviation ( $\sigma_1$ ):", std_dev)
print("Expected Standard Deviation ( $\sigma_0$ ):", expected_std_dev)

Enter the sample size (n): 48
Please enter a value greater than 50.
Enter the sample size (n): 100
Actual Mean ( $\mu_1$ ): 5.0423
Expected Mean ( $\mu_0$ ): 5.0
Actual Standard Deviation ( $\sigma_1$ ): 2.18442457182664
Expected Standard Deviation ( $\sigma_0$ ): 2.179449471770337
```

Q.No.4) In general, if $np > 5$ and $nq > 5$ in binomial distribution, binomial probabilities can be approximated using the normal distribution. Please select any big enough n and p 's values to verify in Python program or Excel and plot the histogram.

Solution:

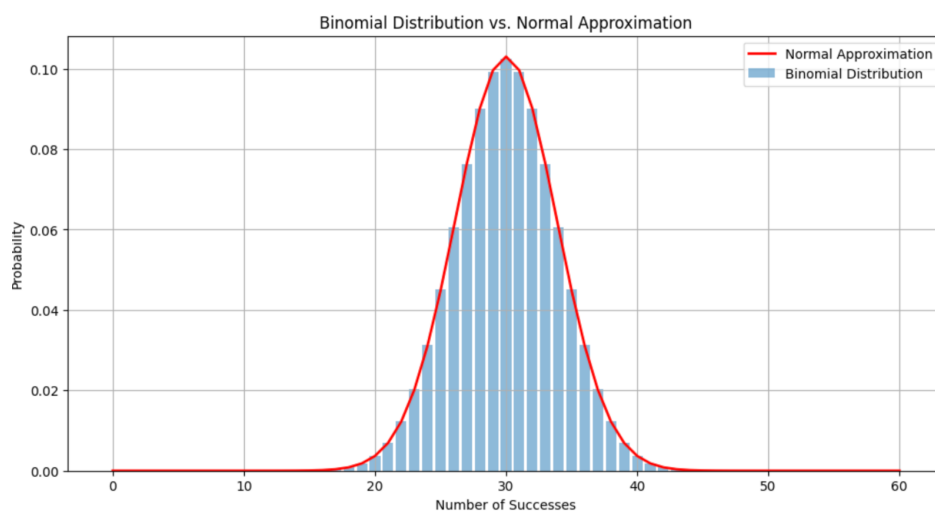
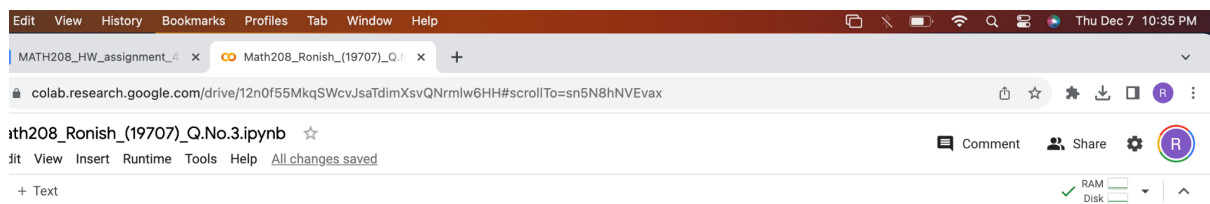
```
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28] #Q.No.4
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import binom, norm

# Define the parameters
n = 60 # number of trials
p = 0.5 # probability of success

# Generate binomial distribution
binomial_dist = binom(n, p)
x = np.arange(0, n+1)
binomial_probs = binomial_dist.pmf(x)

# Approximate with normal distribution
mean, var = binom.stats(n, p, moments='mv')
normal_dist = norm(mean, np.sqrt(var))
normal_probs = normal_dist.pdf(x)

# Plotting
plt.figure(figsize=(12, 6))
plt.bar(x, binomial_probs, alpha=0.5, label='Binomial Distribution')
plt.plot(x, normal_probs, color='red', linewidth=2, label='Normal Approximation')
plt.xlabel('Number of Successes')
plt.ylabel('Probability')
plt.title('Binomial Distribution vs. Normal Approximation')
plt.legend()
plt.grid(True)
plt.show()
```



Q.No.5) In coin tossing experiments, please find the probability of the exact 6 heads from 12 tossing by ONLY using the normal distribution method.

Solution:

We have,

$n = 12$ (no. of trials or tosses)

$p = 0.5$ (probability of getting head)

$q = 0.5$ (probability of failure(no head))

For a binomial distribution, the mean (μ) and standard deviation (σ) are given by:

$$\text{Mean}(\mu) = n * p = 12 * 0.5 = 6$$

$$\text{S.D}(\sigma) = \sqrt{n * p * q} = \sqrt{12 * 0.5 * 0.5} = \sqrt{3}$$

The probability of getting exactly 6 heads can be found by finding the area under the normal distribution curve between 5.5 and 6.5 (since we want exactly 6 heads).

To find the probability of getting exactly 6 heads, we need to calculate the area under the normal curve between 5.5 and 6.5.

For exactly 6 heads, we consider the range from 5.5 to 6.5 (to include 6 heads but not 7).

$$\text{Z-Score for 5.5 (Z1)} = \frac{(x - \mu)}{\sigma} = \frac{(5.5 - 6)}{\sqrt{3}} = -0.289$$

$$\text{Z-Score for 6.5 (Z2)} = \frac{(x - \mu)}{\sigma} = \frac{(6.5 - 6)}{\sqrt{3}} = 0.289$$

Using Python function , The probabilities $P(Z_1)$ and $P(Z_2)$ are found using the cumulative distribution function (CDF) for a standard normal distribution.

```
from scipy.stats import norm

# Given values for the problem
mean = 6 # Mean (μ) from np = 12 * 0.5
standard_deviation = 3**(0.5) # Standard Deviation (σ) from √(np(1-p))

# Calculate Z-Scores for 5.5 and 6.5
z1 = (5.5 - mean) / standard_deviation
z2 = (6.5 - mean) / standard_deviation

# Use the Z-Scores to find the probability for each
p_z1 = norm.cdf(z1)
p_z2 = norm.cdf(z2)

# The probability of exactly 6 heads is the difference between P(Z_2) and P(Z_1)
probability_exactly_6_heads = p_z2 - p_z1

p_z1, p_z2, probability_exactly_6_heads
```

(0.38641499634222376, 0.6135850036577762, 0.22717000731555248)

Here,

$P(Z_1) \sim 0.386$

$P(Z_2) \sim 0.614$ [Using CDF in python]

Now,

$$\begin{aligned}\text{Probability of exactly six heads} &= P(Z2) - P(Z1) \\ &= 0.614 - 0.386 \\ &= 0.227\end{aligned}$$

Therefore, the probability of getting exactly 6 heads in 12 coin tosses is 22.7%

Q.No.6) Given that the defective rate of a product of the batteries in a manufacturing company is 6%, 150 batteries are randomly selected from the population. Please find the probability of 12 or more defective ones in them by ONLY using the normal distribution method.

Solution

Given:

$$n = 150$$

Probability of a defective battery (p) = 0.06.

$$\text{Mean } (\mu) = np = 150 * 0.06 = 9$$

$$\text{Standard Deviation } (\sigma) = \sqrt{n * p * q} = \sqrt{150 * 0.06 * 0.94} = 2.91$$

The probability of 12 or more defective batteries can be calculated using the normal approximation to the binomial distribution.

$$Z = \frac{(x - \mu)}{\sigma} = \frac{(12 - 9)}{2.91} \approx 1.031$$

Now, the probability using the standard normal distribution table for $z \approx 1.031$

$$P(X \geq 12) \approx P(X > 11.5)$$

$$P(X > 11.5) \approx 1 - P(Z \leq 1.06)$$


Using Python function for the standard normal distribution (Z), we get,

```
import scipy.stats as stats

# Define the value of Z
z = 1.031

# Calculate the probability P(Z ≤ z)
probability = stats.norm.cdf(z)

print(f"P(Z ≤ {z}) = {probability:.4f}")
```

 $P(Z \leq 1.031) = 0.8487$

Using standard normal distribution:

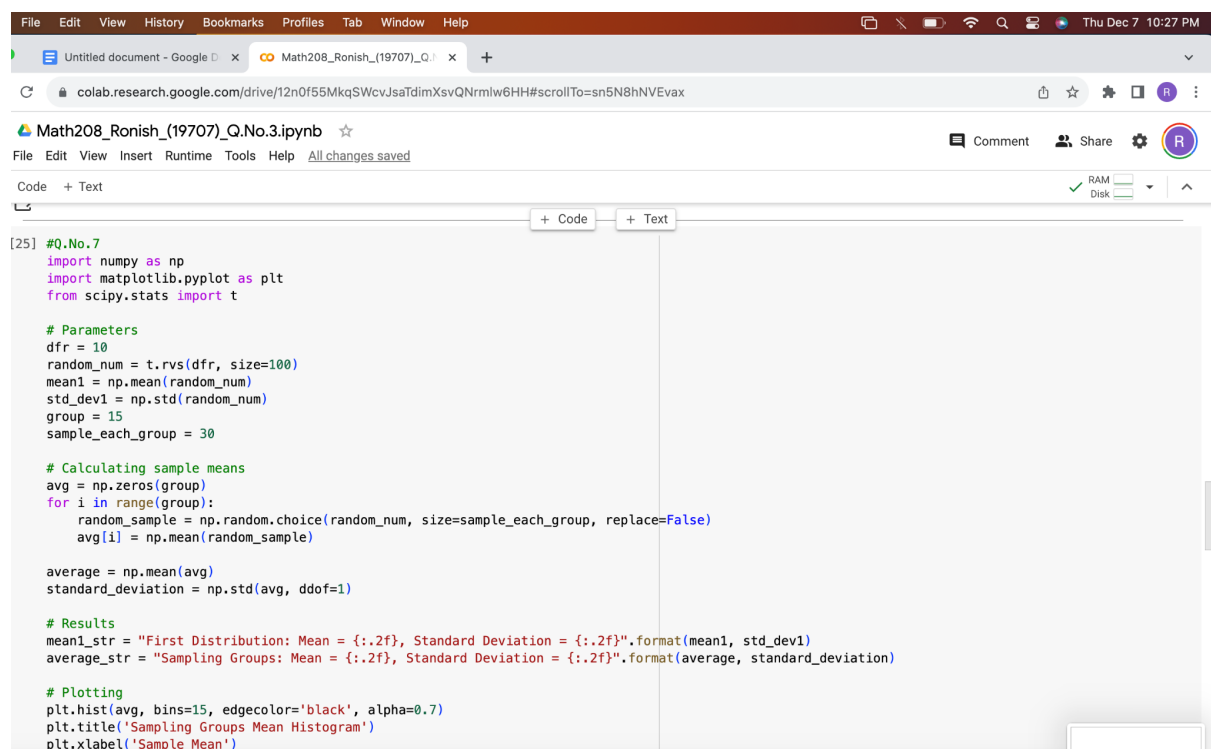
$$P(X \geq 12) = 1 - P(Z \leq 1.031)$$

$$= 1 - 0.8487$$

$$\sim 0.1513$$

Therefore, the probability of 12 or more defective batteries is 0.1513.[i.e. (15.13%)]

Q.No.7) Write a Python program by calling functions in the following link to create 100 random numbers in T distribution with $df=10$ (degree of freedom) and calculate the mean μ and standard deviation σ . After that, the 30 samples will be randomly selected from these random numbers in each sampling group. A total of 15 sampling groups should be created. Based on Central Limit Theorem (CLT), the mean value \bar{x} in total 15 sampling group is roughly the mean μ of 100 random numbers and $\sigma_{\bar{x}} = \sigma/\sqrt{n}$. Please verify it and plot the histogram, which should be normal distribution.



```
[25] #Q.No.7
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import t

# Parameters
dfr = 10
random_num = t.rvs(dfr, size=100)
mean1 = np.mean(random_num)
std_dev1 = np.std(random_num)
group = 15
sample_each_group = 30

# Calculating sample means
avg = np.zeros(group)
for i in range(group):
    random_sample = np.random.choice(random_num, size=sample_each_group, replace=False)
    avg[i] = np.mean(random_sample)

average = np.mean(avg)
standard_deviation = np.std(avg, ddof=1)

# Results
mean1_str = "First Distribution: Mean = {:.2f}, Standard Deviation = {:.2f}".format(mean1, std_dev1)
average_str = "Sampling Groups: Mean = {:.2f}, Standard Deviation = {:.2f}".format(average, standard_deviation)

# Plotting
plt.hist(avg, bins=15, edgecolor='black', alpha=0.7)
plt.title('Sampling Groups Mean Histogram')
plt.xlabel('Sample Mean')
```

```
plt.show()  
  
mean1_str, average_str
```

