

Samir Khadka (19701)

Question 1:

Answer:

Given:

Mean( $\mu$ )=10.3 cm

Standard Deviation( $\sigma$ ) = 0.65cm

a. Less than 9 cm

$$Z\text{-score}(Z) = \frac{X-\mu}{\sigma} = \frac{9-10.3}{0.65} = \frac{-1.3}{0.65} = -2$$

b. Between 9.5 cm to 10.6 cm

$$Z_{\text{lower}} = \frac{X-\mu}{\sigma} = \frac{9.5-10.3}{0.65} = \frac{-0.8}{0.65} \approx -1.23$$

$$Z_{\text{upper}} = \frac{X-\mu}{\sigma} = \frac{10.6-10.3}{0.65} = \frac{0.3}{0.65} \approx 0.461$$

$$P(9.5 < X < 10.6) = P(Z_{\text{upper}}) - P(Z_{\text{lower}})$$

c.

$$X = \mu + Z_{\text{top 20\%}} + \sigma$$

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0s
import scipy.stats as stats

mean = 10.3
sd = 0.65

z_score_a = (9 - mean) / sd
probability_a = stats.norm.cdf(z_score_a) * 100

z_score_b_lower = (9.5 - mean) / sd
z_score_b_upper = (10.6 - mean) / sd
probability_b = (stats.norm.cdf(z_score_b_upper) - stats.norm.cdf(z_score_b_lower)) * 100

z_score_c_top_20 = stats.norm.ppf(0.8)
length_c_top_20 = mean + z_score_c_top_20 * sd

print("a. Probability of anchovies length less than 9 cm:", round(probability_a, 2), "%")
print("b. Probability of anchovies length between 9.5 cm and 10.6 cm:", round(probability_b, 2), "%")
print("c. Minimum length for the top 20%:", round(length_c_top_20, 2), "cm")

a. Probability of anchovies length less than 9 cm: 2.28 %
b. Probability of anchovies length between 9.5 cm and 10.6 cm: 56.86 %
c. Minimum length for the top 20%: 10.85 cm
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<https://colab.research.google.com/drive/192Mj67Cz10InRFsDNrdwX0-uaoWvICcq?authuser=0>

Question 2:

Answer:

Given that:

Mean of  $X(\mu_X)$ =10

Mean of  $Y(\mu_Y)$ =15

Standard Deviation of X( $\sigma_X$ ) = 3

Standard Deviation of Y( $\sigma_Y$ ) = 8

1. X+Y:

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

$$\text{Variance}(\sigma_{X+Y}^2) = \sigma_X^2 + \sigma_Y^2 = 3^2 + 8^2 = 9 + 64 = 73$$

$$\text{Standard Deviation}(\sigma_{X+Y}) = \sqrt{\sigma_{X+Y}^2} = \sqrt{73} \approx 8.55$$

2. X-Y:

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

$$\text{Variance}(\sigma_{X-Y}^2) = \sigma_X^2 + \sigma_Y^2 = 3^2 + 8^2 = 9 + 64 = 73$$

$$\text{Standard Deviation}(\sigma_{X-Y}) = \sqrt{\sigma_{X-Y}^2} = \sqrt{73} \approx 8.55$$

3. 3X:

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

$$\text{Variance}(\sigma_{3X}^2) = 3^2 \cdot \sigma_X^2 = 3^2 \cdot 3^2 = 81$$

$$\text{Standard Deviation}(\sigma_{3X}) = 3 \cdot \sigma_X = 9$$

4. 4X+5Y:

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

$$\text{Variance}(\sigma_{4X+5Y}^2) = 4^2 \cdot \sigma_X^2 + 5^2 \cdot \sigma_Y^2 = 16 \cdot 3^2 + 25 \cdot 8^2 = 144 + 1600 = 1744$$

$$\text{Standard Deviation}(\sigma_{4X+5Y}) = \sqrt{\sigma_{4X+5Y}^2} = \sqrt{1744} \approx 41.76$$

Question 3:

Answer:

<https://colab.research.google.com/drive/1qeFz3Z9xBTlydqKj4DogHPKomOnIxoqo?authuser=0#scrollTo=9meychACSSle>

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#question 3
import numpy as np
import matplotlib.pyplot as plt

np.random.seed(42)

n = 100
p = 0.05
q = 1 - p

binomial_data = np.random.binomial(n, p, 10000)

mean_cal = np.mean(binomial_data)
std_dev_cal = np.std(binomial_data)

mean_theoretical = n * p
std_dev_theoretical = np.sqrt(n * p * q)

print("Calculated Mean:", mean_cal)
print("Theoretical Mean:", mean_theoretical)
print("Calculated Standard Deviation:", std_dev_cal)
print("Theoretical Standard Deviation:", std_dev_theoretical)

plt.hist(binomial_data, bins=np.arange(-0.5, n + 1.5, 1), density=True, alpha=0.75, label='Binomial Distribution')

x = np.linspace(0, n, 100)
normal_approximation = (1 / (std_dev_theoretical * np.sqrt(2 * np.pi))) * np.exp(-(x - mean_theoretical)**2 / (2 * std_dev_theoretical**2))
plt.plot(x, normal_approximation, 'r', label='Normal Approximation')

plt.title('Binomial Distribution and Normal Approximation')
plt.xlabel('Number of Successes')
plt.ylabel('Probability')
plt.legend()

plt.show()

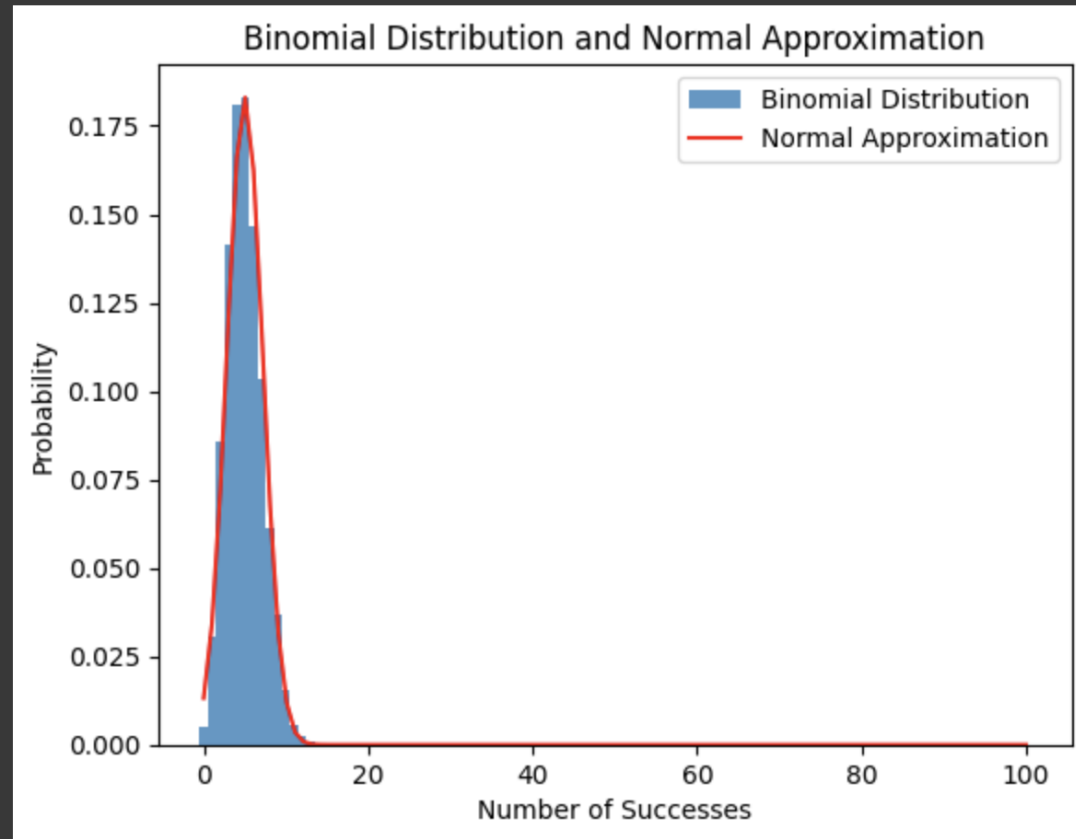
```

Calculated Mean: 4.9557

Theoretical Mean: 5.0

Calculated Standard Deviation: 2.1521936506736563

Theoretical Standard Deviation: 2.179449471770337



Question 4:

Answer:

<https://colab.research.google.com/drive/1tPYvsgSjUrpXlccSOlsGwSfORLCrqHgg?authuser=0>

Individual Assignment: HW A x Course Modules: CS483(B) x Untitled2.ipynb - Colaboratory x

colab.research.google.com/drive/1tPYvsgSjUrpXlccSOIsGwSfORLCrQHgg?authuser=0

Untitled2.ipynb

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#Question 4

import numpy as np

import matplotlib.pyplot as plt

from scipy.stats import binom, norm

np.random.seed(42)

n = 20

p = 0.3

binomial\_data = np.random.binomial(n, p, 10000)

plt.hist(binomial\_data, bins=np.arange(-0.5, n + 1.5, 1), density=True, alpha=0.75, label='Binomial Dis

mu = n \* p

sigma = np.sqrt(n \* p \* (1 - p))

x = np.linspace(0, n, 100)

plt.plot(x, norm.pdf(x, mu, sigma), 'r', label='Normal Approximation')

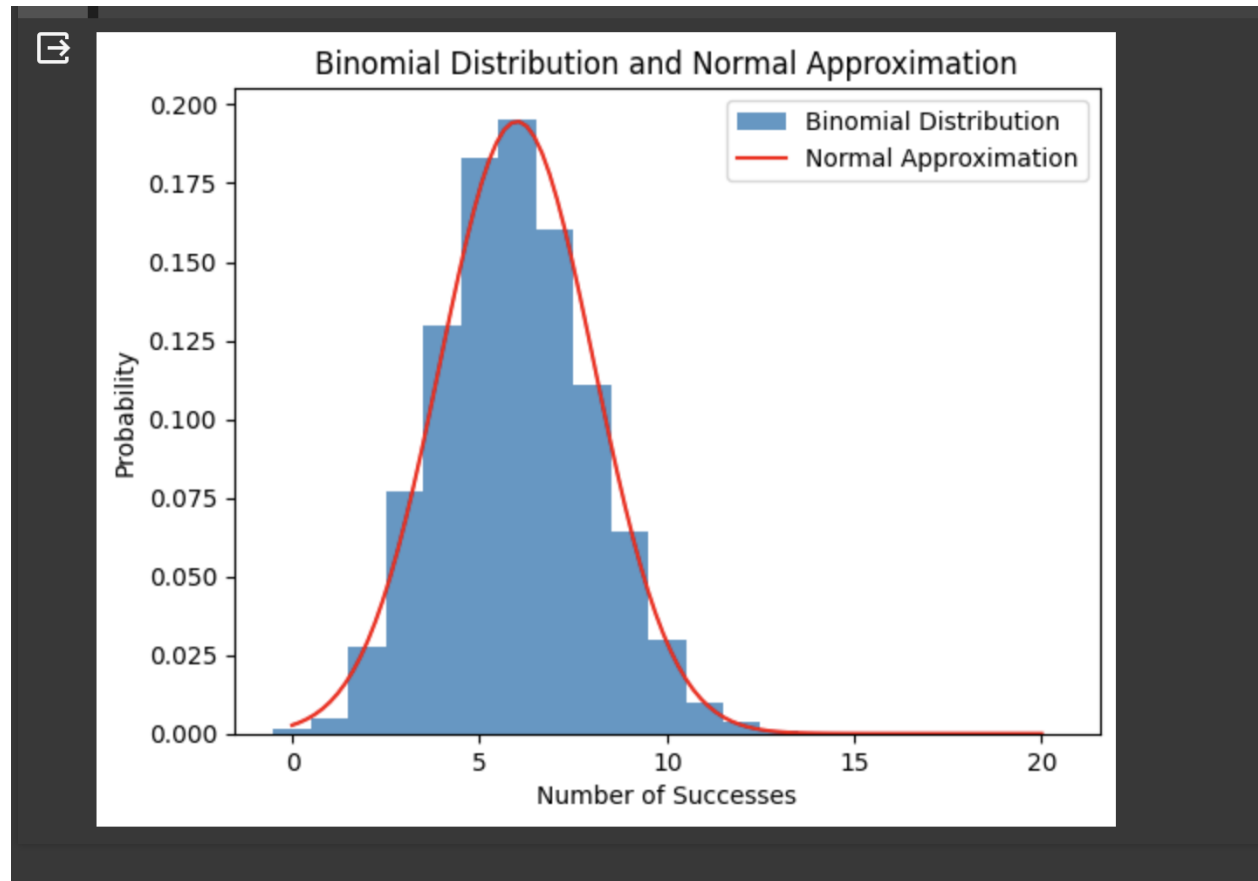
plt.title('Binomial Distribution and Normal Approximation')

plt.xlabel('Number of Successes')

plt.ylabel('Probability')

plt.legend()

plt.show()



Question number:5

Answer:

Given that:

Let,

n be number of trials i.e. coin tosses

p be probability of success i.e getting heads in a single toss

x be number of heads to get probability

We have,

n=12, p= 1/2=0.5, x=6

Now,

Mean ( $\mu$ )=n x p= 12 x 0.5 =6

Standard Deviation ( $\sigma$ )= $\sqrt{n \cdot p \cdot (1 - p)} = \sqrt{12 \cdot 0.5 \cdot 0.5} = \sqrt{3} = 1.73205080757$

Therefore, the required mean is 6 and standard deviation is 1.73205080757

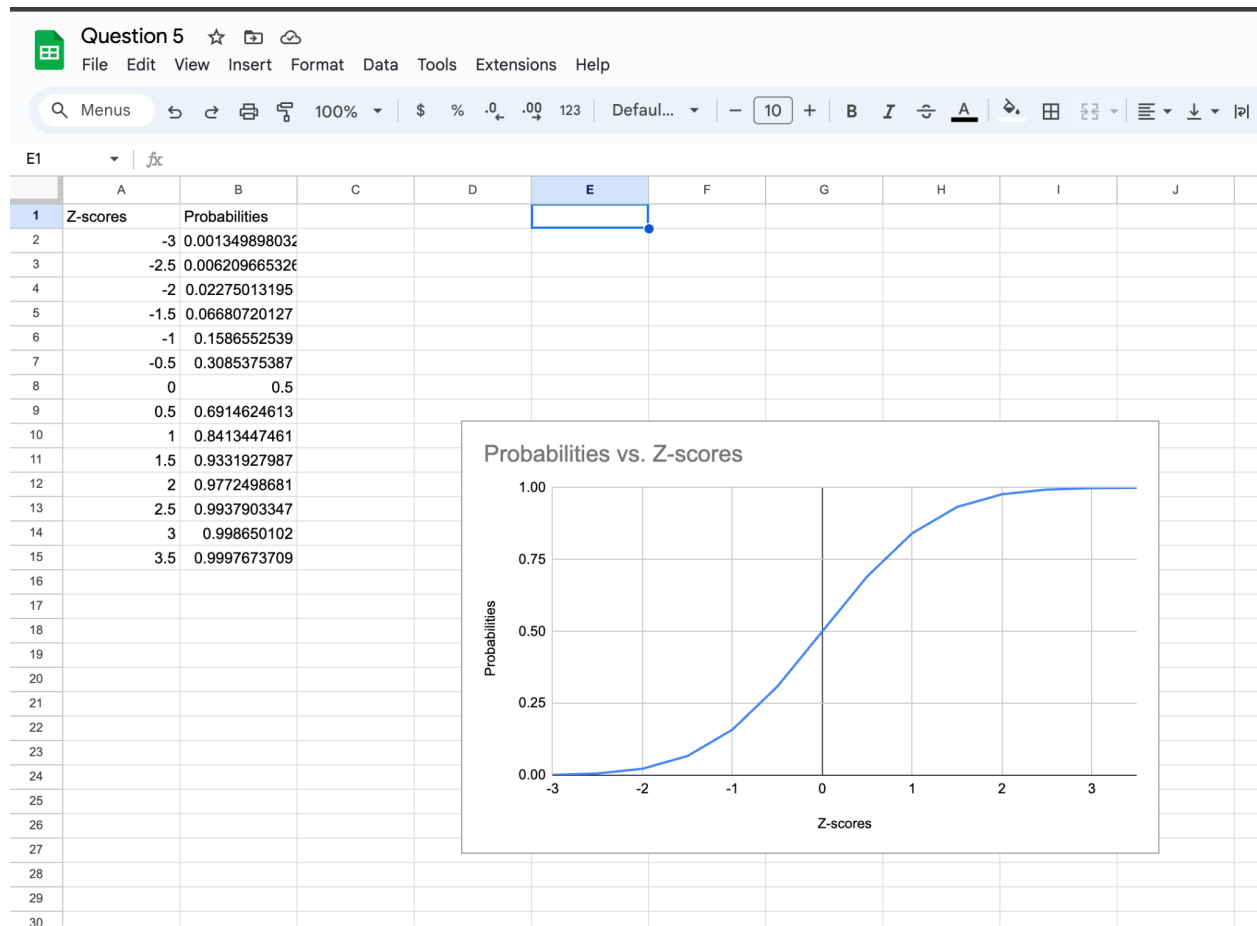
Z-score ( $Z$ )= $\frac{x-\mu}{\sigma}$

For  $x = 6$ :

$$Z = \frac{x-\mu}{\sigma} = \frac{6-6}{\sqrt{3}} = \frac{0}{\sqrt{3}} = 0$$

Therefore, the required Z-score is 0.

Now For normal distribution,



Question 6:

Answer:

Given that:

Let,

n be number of trials i.e. sample size

p be probability of success i.e defective rate

x be number of defective batteries to get probability

We have,

n=150, p=0.06, x=12

Now,

Mean ( $\mu$ )=n x p= 150 x 0.06 =9

Standard Deviation ( $\sigma$ )= $\sqrt{n \cdot p \cdot (1 - p)} = \sqrt{150 \cdot 0.06 \cdot 0.94} = \sqrt{8.46} \approx 2.908$

Therefore, the required mean is 6 and standard deviation is 1.73205080757

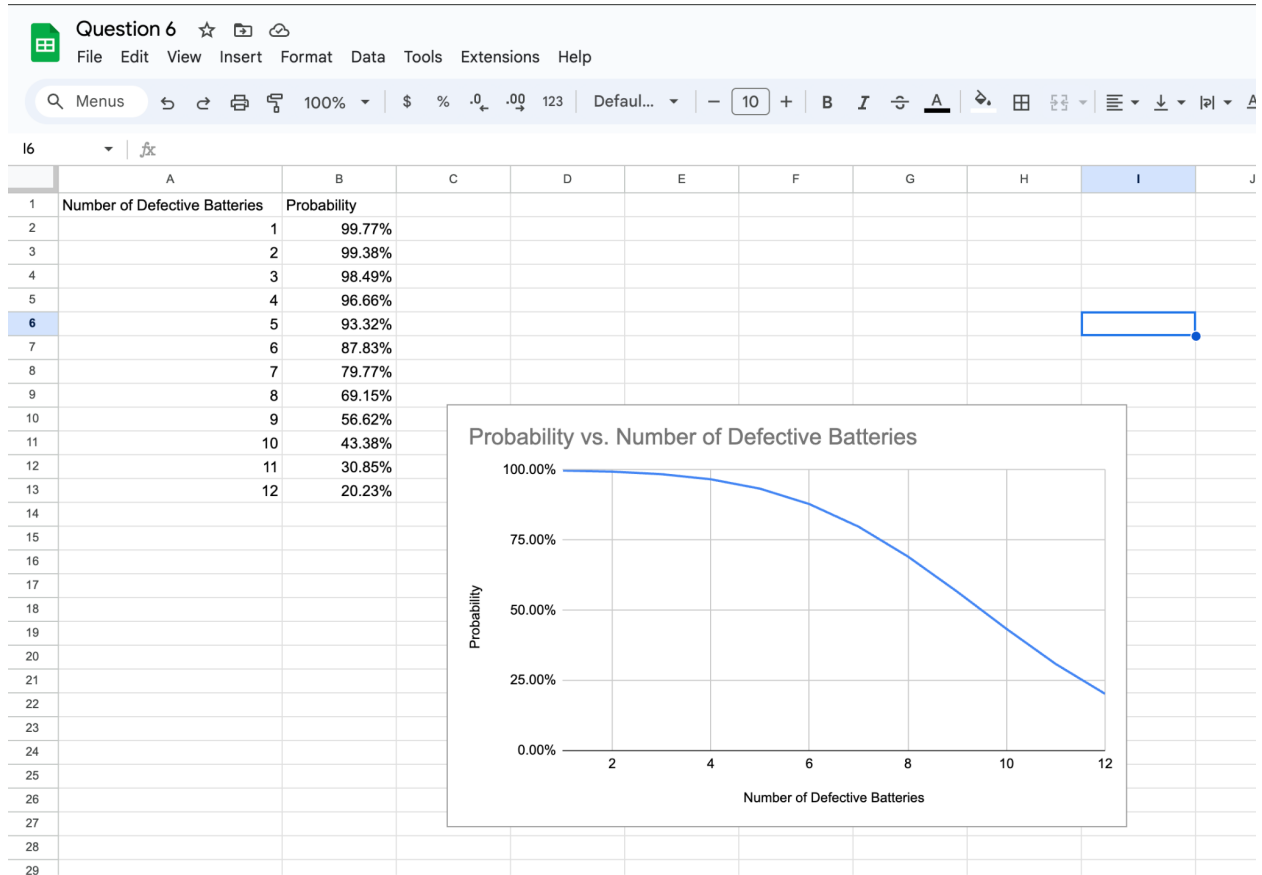
Z-score (Z)= $\frac{x-\mu}{\sigma}$

For  $x = 6$ :

$$Z = \frac{x - \mu}{\sigma} = \frac{12 - 9}{2.908} = \frac{3}{2.908} \approx 1.031$$

Therefore, the required Z-score is 1.031.

Now, for normal distribution:



Question 7:

Answer:

<https://colab.research.google.com/drive/192Mj67Cz10InRFsDNrdwX0-uaoWvICcq?authuser=0#scrollTo=z3C342BhedMK>





```
#Question 7
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import t

def gen_t_dist_samp(size, df=10):
    return t.rvs(df, size=size)

p = 15
n = 30
total_random_numbers = 100
df = 10
rnd_num = gen_t_dist_samp(total_random_numbers, df)

sampling_groups = [gen_t_dist_samp(n, df) for _ in range(p)]

means = [np.mean(group) for group in sampling_groups]

avg_mean = np.mean(means)
means_std = np.std(means, ddof=1)
print(f"Mean of the 100 random numbers: {np.mean(rnd_num)}")
print(f"Mean of the means from sampling groups: {avg_mean}")
print(f"Standard deviation of the means from sampling groups: {means_std}")

plt.hist(means, bins=15, edgecolor='black')
plt.title('Histogram of Means from Sampling Groups')
plt.xlabel('Means')
plt.ylabel('Frequency')
plt.show()
```



Mean of the 100 random numbers:  $-0.04289098622911745$   
Mean of the means from sampling groups:  $-0.04894609562598996$   
Standard deviation of the means from sampling groups:  $0.21082682015163318$

