

Setup

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import scipy as stats
from scipy.stats import probplot, shapiro
np.random.seed(42)
from numpy.random import default_rng
```

Problem 1

An emergency room (ER) receives patients randomly throughout the day. The number of patients arriving per hour follows a Poisson distribution with parameter 10 (10 patients per hour). Due to hospital capacity limitations, the ER becomes overcrowded when more than 15 patients arrive in a single hour. You will simulate patient arrivals and estimate probabilities related to ER overcrowding.

- i. Use simulation to estimate the probability that the ER experiences overcrowding (more than 15 patients in an hour).

```
def simulate_hours(lambda_param, hours):
    return np.random.poisson(lambda_param, hours)

lambda_param = 10
n = 1*10*7
poisson = simulate_hours(lambda_param, n)
overcrowding_prob = np.mean(poisson > 15)

print(f"Estimated probability of ER overcrowding:
{overcrowding_prob:.4f}")
```

Estimated probability of ER overcrowding: 0.0857

- ii. Simulate 100 days, each consisting of 24 hours of patient arrivals, estimate the proportion of days where the ER is overcrowded for at least 4 hours.

```
simulations = []
lambda_param = 10
```

```

days = 100
hours = 24
for i in range(days):
    poisson = simulate_hours(lambda_param, hours)
    simulations.append(poisson)

crowded_days = 0
for day in simulations:
    crowded_hours = 0
    for hour in day:
        if hour > 15:
            crowded_hours += 1
    if crowded_hours >= 4:
        crowded_days += 1

print(crowded_days/days)

```

0.01

iii. Using the data simulated in the previous question, compute the mean of 24 hourly arrivals in each day, then you will have 100 average arrivals. Compare it to a normal distribution using QQ plot.

```

daily_means = np.array([np.mean(day) for day in simulations])

plt.figure(figsize=(10, 6))

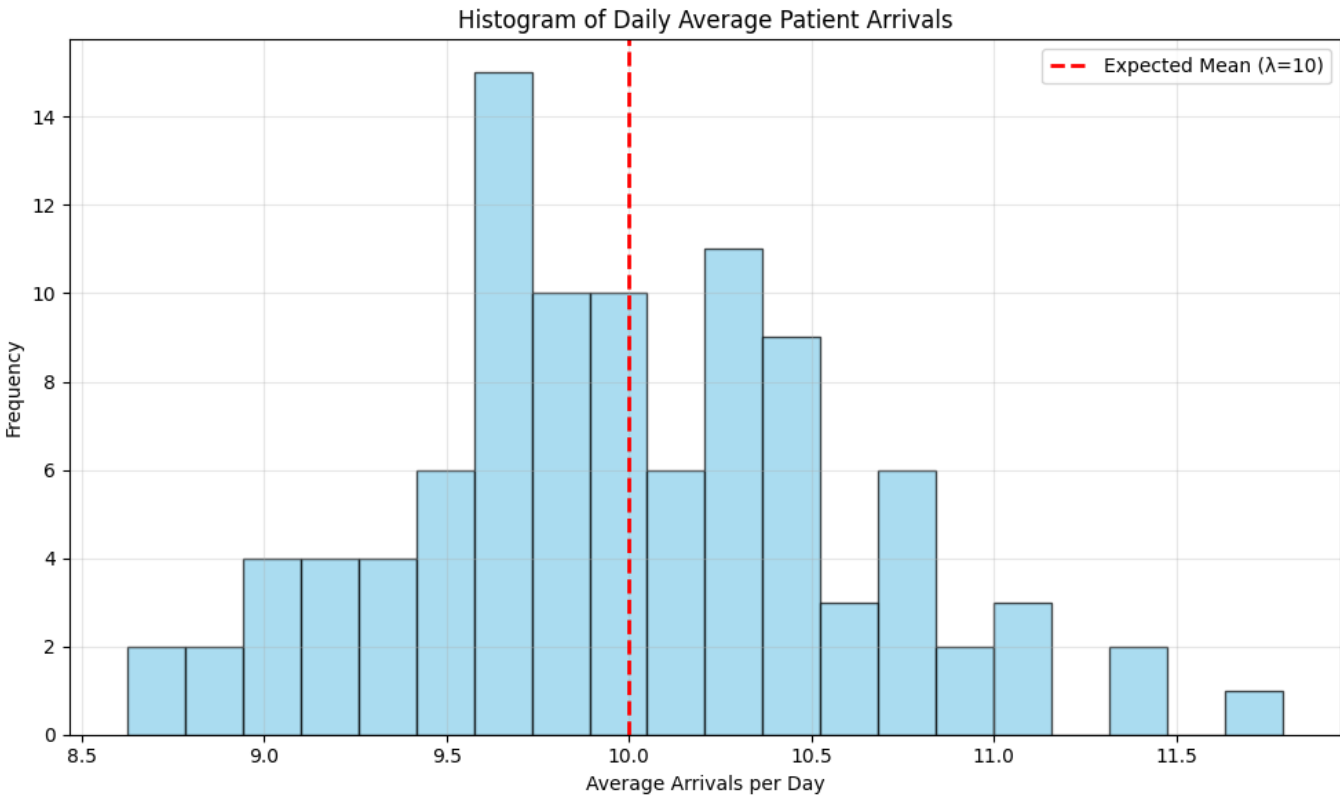
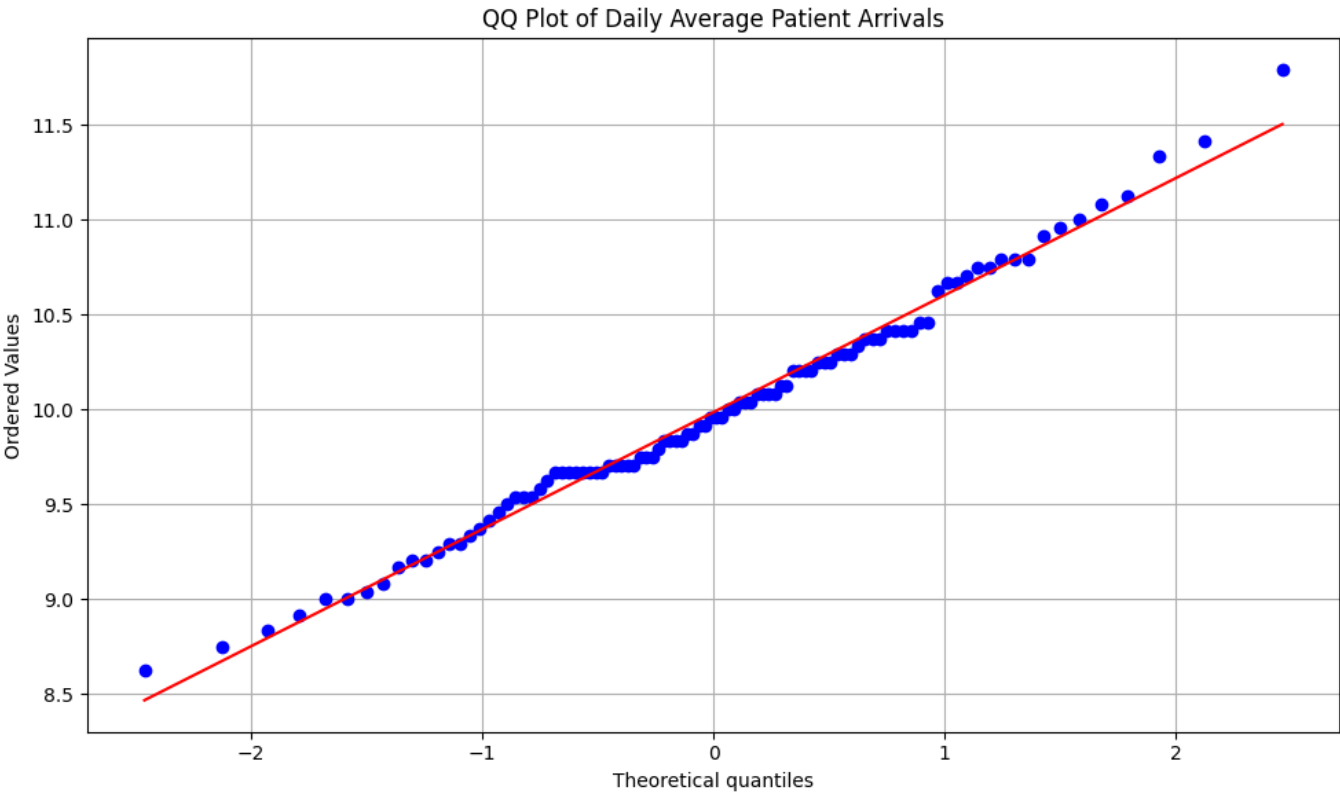
probplot(daily_means, dist="norm", plot=plt)

plt.title('QQ Plot of Daily Average Patient Arrivals')
plt.grid(True)
plt.tight_layout()

plt.figure(figsize=(10, 6))
plt.hist(daily_means, bins=20, alpha=0.7, color='skyblue',
edgecolor='black')
plt.axvline(lambda_param, color='red', linestyle='dashed', linewidth=2,
label=f'Expected Mean ( $\lambda$ = $\{\lambda_{\text{lambda\_param}}\}$ )')
plt.title('Histogram of Daily Average Patient Arrivals')
plt.xlabel('Average Arrivals per Day')
plt.ylabel('Frequency')
plt.grid(True, alpha=0.3)
plt.legend()
plt.tight_layout()

plt.show()

```



Problem 2

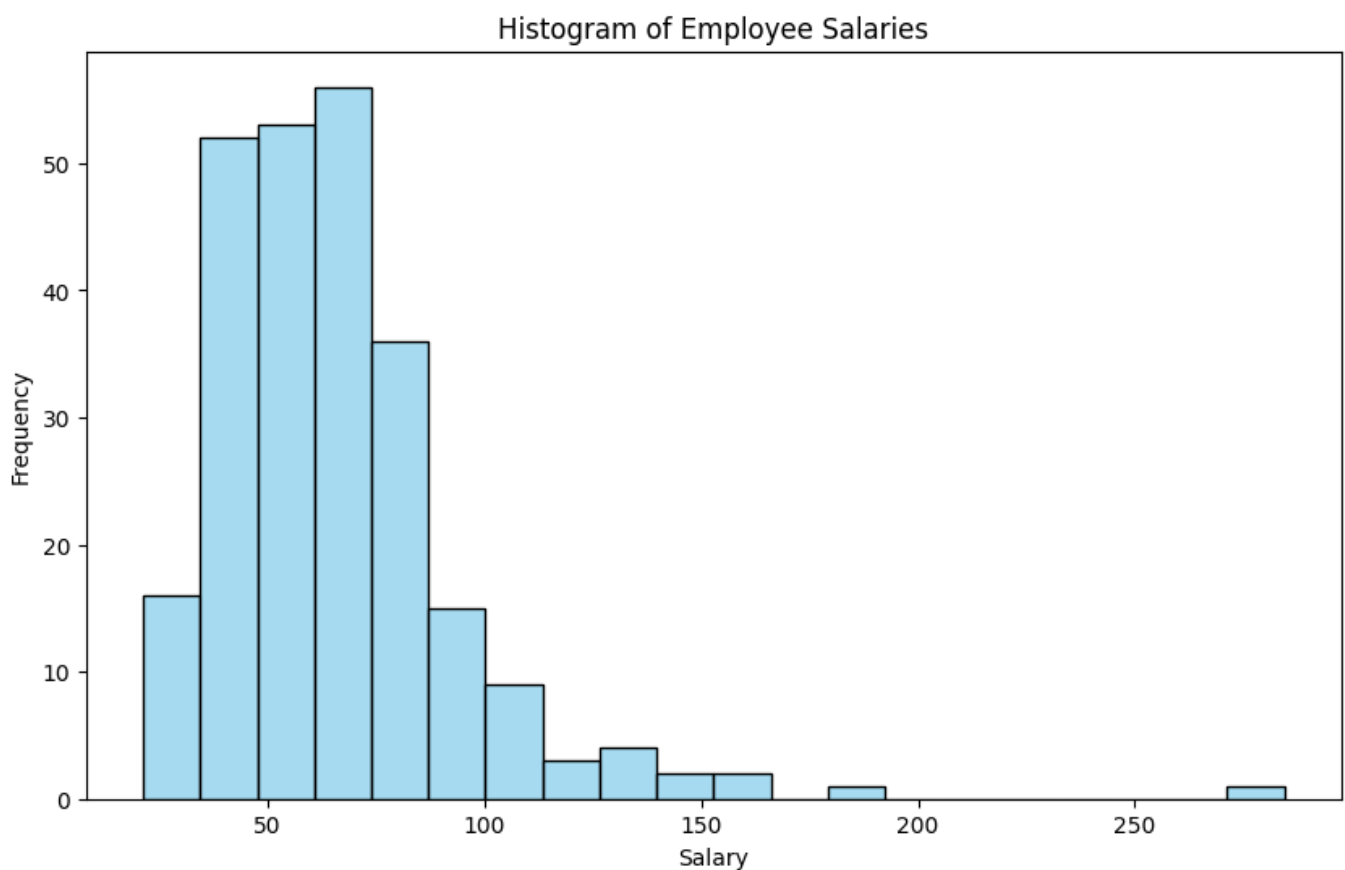
You are given a dataset containing the annual salaries (in \$1000s) of 250 employees in a company. The dataset is stored as a csv file(employee_salaries.csv).

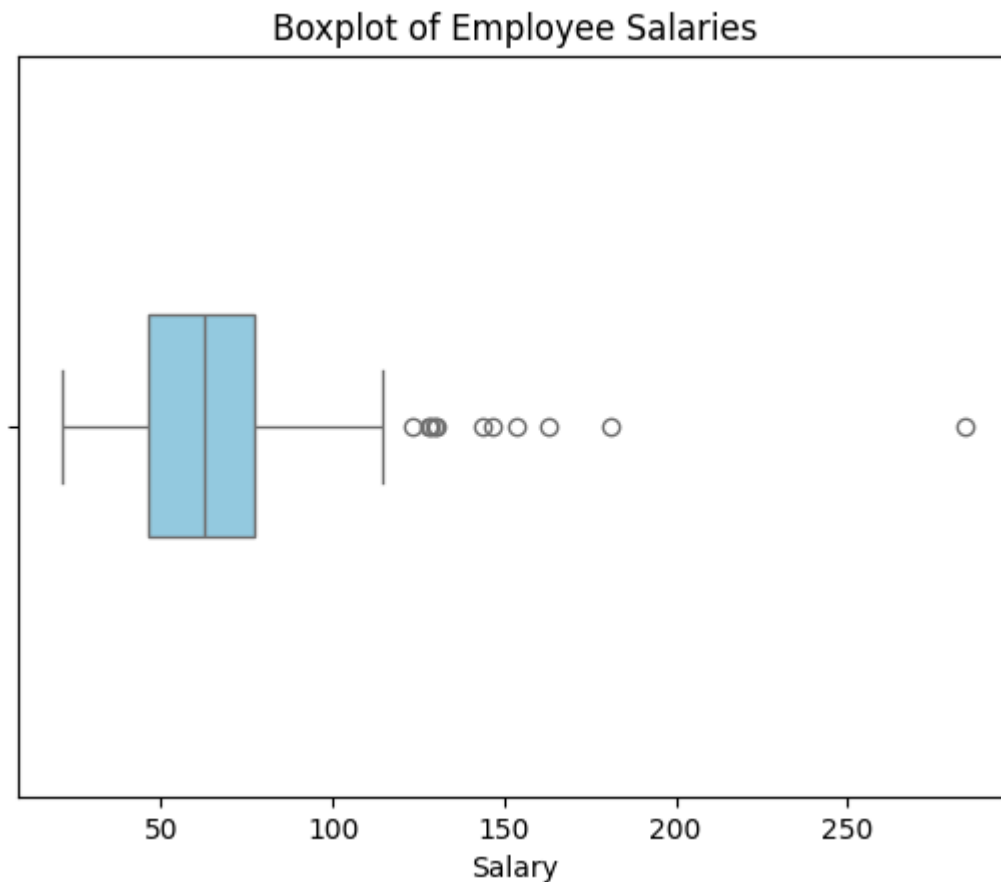
- i. Load the dataset and visualize the distribution of salaries using a histogram and a boxplot.

```
df = pd.read_csv('employee_salaries.csv')
plt.figure(figsize=(10, 6))

sns.histplot(df['Salary (in $1000s)'], bins=20, color='skyblue',
edgecolor='black')
plt.title('Histogram of Employee Salaries')
plt.xlabel('Salary')
plt.ylabel('Frequency')
plt.show()

sns.boxplot(x='Salary (in $1000s)', data=df, color='skyblue', width=0.3)
plt.title('Boxplot of Employee Salaries')
plt.xlabel('Salary')
plt.show()
```





ii. Compute the observed sample mean, median, and standard deviation of salaries.

```
mean = df['Salary (in $1000s)'].mean()
median = df['Salary (in $1000s)'].median()
std = df['Salary (in $1000s)'].std()

print(f"Mean: {mean:.2f}")
print(f"Median: {median:.2f}")
print(f"Standard Deviation: {std:.2f}")
```

```
Mean: 65.79
Median: 62.37
Standard Deviation: 29.12
```

iii. Bootstrap estimation of the bias for the median: a. Perform 1,000 bootstrap resamples (sampling with replacement from the dataset) b. Compute the bootstrap median for each resample

```
n_bootstraps = 1000
bootstrap_medians = np.zeros(n_bootstraps)
salaries = df['Salary (in $1000s)'].values
rng = default_rng(42)
```

```

for i in range(n_bootstraps):
    # Generate bootstrap sample (with replacement)
    bootstrap_sample = rng.choice(salaries, size=len(salaries),
    replace=True)

    # Calculate the median of the bootstrap sample
    bootstrap_medians[i] = np.median(bootstrap_sample)
print(bootstrap_medians)

```

```

[62.59317019 60.42718467 62.34736438 62.59317019 62.19018674 62.44837942
61.63154753 60.58449623 61.13363628 57.10682449 62.94035728 63.35690052
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62.34736438 59.58556526 65.04681184 61.13363628]

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c. Estimate the bias of the sample median

```

bootstrap_median_mean = np.mean(bootstrap_medians)

original_median = np.median(salaries)
bootstrap_bias = bootstrap_median_mean - original_median
print(bootstrap_bias)

```

```

-0.580979963667076

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