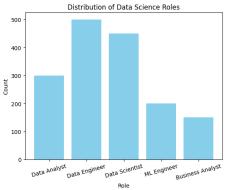


```
[2]: import matplotlib.pyplot as plt

# Data
roles = ['Data Analyst', 'Data Engineer', 'Data Scientist', 'ML Engineer', 'Business Analyst']
counts = [300, 500, 450, 260, 150]

plt.bar(roles, counts, color='skyblue')
plt.title('Distribution of Data Science Roles')
plt.vlabel('Sole')
plt.ylabel('Gount')
plt.vlabel('Count')
plt.xitck(rotation=15)
plt.show()
```



Decrypt data
decrypted_text = cipher_suite.decrypt(cipher_text)

```
# Output

print("Original Data:", plain_text.decode())

print("Tencrypted Data:", cipher_text.decode())

print("Decrypted Data:", decrypted_text.decode())

Original Data: Rajalakshmi Engineering College

Encrypted Data: Rajalakshmi Engineering College

Indian_Score = [38, 55, 90, 129, 165, 200, 239, 270, 310, 350]

Indian_Score = [39, 55, 90, 129, 165, 200, 239, 270, 310, 350]

Srilankan_Score = [25, 78, 90, 120, 160, 170, 195, 220, 255, 279]

cricket.plot(Overs, Indian_Score, color="green", label="INDIA")

cricket.plot(Overs, Srilankan_Score, color="red", label="SRILANKA")

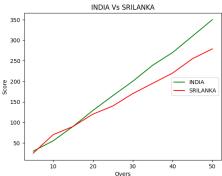
cricket.xlabel("Overs")

cricket.ylabel("Overs")

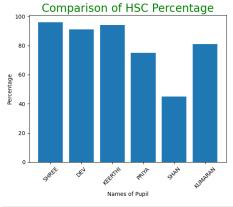
cricket.lagend(lor="center right")

cricket.lagend(lor="center right")

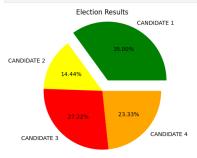
cricket.show()
```



```
import matplotlib.pyplot as hscmark
import numpy as np
Names = ['SHREE', 'DEV', 'KEERTHI', 'PRIYA', 'SHAN', 'KLMARAN']
xaxis = np.arange(len(Names))
Percentage_hsc = [96, 91, 94, 75, 45, 81]
hscmark.tlar(Names, Percentage_hsc)
hscmark.xticks(xaxis, Names, rotation=45)
hscmark.xticks(xaxis, Names, rotation=45)
hscmark.xlabel('Names of Pupil')
hscmark.ylabel('Tenecntage'')
hscmark.xibel('Comparison of HSC Percentage'', fontsize=20, color="green")
hscmark.show()
```



```
[17]: import matplotlib.pyplot as election
    labels = ['CANDIDATE 1', 'CANDIDATE 2', 'CANDIDATE 3', 'CANDIDATE 4']
    Votes = [315, 130, 245, 210]
    colors = ['green', 'yellow', 'red', 'orange']
    explode = (0.2, 0, 0, 0)
    election.pie(Votes, labels-labels, colors-colors, explode=explode, autopct='%0.2f%X')
    election.title('Election Results')
    election.show()
```



```
[18]: import nitk
from nitk.tokenize import word tokenize
from nitk.corpus import gutenberg
nitk.download('gutenberg')
nitk.download('gutenberg')
nitk.download('gutenberg')
sample = gutenberg.rew('austen-emma.txt'')
token = word_tokenize(sample)
wlist = [token[i] for i in range(50)]
wordfreq = [wlist.count(w) for w in wlist]
print('Pdirs\n' * str[list.(zip(wlist, wordfreq))))

[nitk_data] Downloading package gutenberg to
[nitk_data] * Culvers\n'tITMESH\AppBota\noming\nitk_data...
[nitk_data] Package gutenberg is already up-to-date!
[nitk_data] Downloading package nownt to
```

[nltk_data] C:\Users\MITHESH\AppData\Roaming\nltk_data... [nltk_data] Package punkt is already up-to-date! Pairs [('[', 1), ('Emma', 2), ('by', 1), ('Jane', 1), ('Austen', 1), ('1816', 1), (']', 1), ('VOLUME', 1), ('I', 2), ('CHAPTER', 1), ('I', 2), ('Emma', 2), ('Moo dhouse', 1), (',', 5), ('handsome', 1), (',', 5), ('clever', 1), (',', 5), ('and', 3), ('rich', 1), (',', 5), ('with', 2), ('a', 1), ('comfortable', 1), ('home', 1), ('and', 3), ('happy', 1), ('disposition', 1), ('', 5), ('seemed', 1), ('to', 1), ('unite', 1), ('one', 1), ('of', 2), ('kest', 2), ('best', 1), ('best', 2), ('world', 1), ('the', 2), ('world', 1), ('world', import numpy as np array-np.random.randint(1,100,16) def outbetection(array): sorted(array) Q1,Q3-np.percentile(array,[25,75]) IQR-0-0 1r-01-(1.5*IQR) ur-Q3-(1.5*IQR) ur-ur-outbetection(array) print(Ir,ur) import seaborn as sns Mmatploilib inline %matplotlib inline sns.displot(array) -59.375 155.625 [20]: <seaborn.axisgrid.FacetGrid at 0x16b23e6c350> 4.0 3.5 3.0 2.5 S 2.0 1.5 1.0 0.5 0.0 40 80 [24]: import numpy as np import pandas as pd df=pd.read.csv("hotel_Dataset.csv") df.drop_duplicates(inplace=True) index=pa.aray(last.range(a,len(df)))) df.set_index(index,inplace=True) df.drop('age_droup.1')_axis=1,inplace=True) df.eg_group.unique() df.eg_group df.Hotel.unique() df.Hotel.replace(['Ibys'],'Ibis',inplace=True) df.Hotel.replace(['Tbys'], 'Tbis', implace=True) df.FoodPreference.unique df.FoodPreference.unique df.FoodPreference.replace(['Non-Veg'], 'Veg', 'Implace=True) df.FoodPreference.replace(['Non-Veg'], 'Non-Veg', implace=True) df.EstimatedSolary.fillna(round(df.EstimatedSolary.mem()), implace=True) df.ModPpax.fillna(round(df.EstimatedSolary.mem()), implace=True) df.ModPpax.fillna(round(df.EstimatedSolary.median()), implace=True) df.ModPpax.fillna(round(), implace=True) df.ModPpax.fillna(round(), implace=True) df.ModPpax.fillna(round(), implace=True) df.ModPpax.fillna(round(), implace=True) df.ModPpax.fillna(round(), implace=True) df.ModPpax.fillna(round(), implace=True) df.Mo CustomerID Age_Group Rating(1-5) Hotel FoodPreference Bill NoOfPax EstimatedSalary Ibis 2 40000 20-25 Veg 1300 1 2 30-35 5 LemonTree Non-Veg 2000 3 59000 3 25-30 6 RedFox Veg 1322 30000 **3** 4 20-25 -1 LemonTree Veg 1234 2 120000 35+ Veg 989 45000 5 6 35+ 3 lbis Non-Veg 1909 2 122220 6 7 35+ 4 RedFox Veg 1000 -1 21122 7 8 20-25 7 LemonTree Veg 2999 -10 345673 -1 2 Ibis 25-30 Non-Veg 3456 3 -99999 8 **9** 10 30-35 5 RedFox Non-Veg -6755 4 87777 import matplotlib.pyplot as plt # Generate a population population = np.random.normal(50, 10, 100000) Dit.figure(figsize-(12, 8)) for i, size in enumerate(sample_sizes): sample_means = [np.mean(np.rendom.choice(population, size-size, replace-False)) for _ in range(num_samples)] plt.wibplot(len(sample_gizes), i, ii) plt.hist(sample_means, bins-80, alpha-0.7, label-f'Sample Size (size)') plt.woline(np.mean(population), color-'red', linestyle-'dashed', linewidth-1.5, label-'Population Mean') plt.viabe('Sample Mean') plt.vlabe('Sample Mean') plt.vlabel('Frequency') plt.vlabel('Frequency') plt.legend() plt.tight_layout() Sampling Distribution (Sample Size 30) Sample Size 30 100 --- Population Mean 80 60 40 20 0 -50 52 Sample Mean Sampling Distribution (Sample Size 50) Sample Size 50 80 --- Population Mean 60

```
Sample Mean
                                                                                                                     Sampling Distribution (Sample Size 100)
                                                                                                                                                                                                                                                      Sample Size 100
                    80
                                                                                                                                                                                                                                                              Population Mean
                60 کا
               Freque
04
                    20
                                                                                                                             49
                                                                                                                                                                                                   51
                                                                                                                                                                                                                                                                        53
model = LinearRegression()
model.fit(x, train, y, train)
train, score = model.score(x, train, y, train)
test_score = model.score(x, test, y, test)
coef = model.coef_
intercept = model.intercept_
print(f"Testing Score: (train_score:.2f)")
print(f"Testing Score: (test_score:.2f)")
print(f"Testing Score: (test_score:.2f)")
print(f"Griciant: (coef[0][0]:.2f]")
print(f"Intercept: (intercept[0]:.2f)")
              Testing Score: 0.90
Coefficient: 9423.82
Intercept: 25321.58
           import pandas as pd
              final_model.fit(x_train, y_train)
             print("Final Model Training Score:", final_model.score(x_train, y_train))
print("Final Model Testing Score:", final_model.score(x_test, y_test))
             print("\nClassification Report:")
print(classification_report(label, final_model.predict(features)))
             Test: 0.90, Train: 0.84, Random State: 4
Test: 0.91, Train: 0.83, Random State: 47
Test: 0.93, Train: 0.84, Random State: 61
Test: 0.96, Train: 0.82, Random State: 20
Final Model Training Score: 0.8385
Final Model Testing Score: 0.9625
             Classification Report:
                                     precision recall f1-score support
                                        0.85
0.84
                                                            0.92
0.71
                                                                                 0.89
0.77
                                                                                                     257
143
                                                                                0.85
0.83
0.84
                     accuracy
 [29]: import numpy as np
from scipy.stats import f_oneway
from statsmodels.stats.multicomp import pairwise_tukeyhsd
            # Generate data

np.random.seed(42)

n = 25

growth_A = np.random.normal(loc-10, scale=2, size=n)

growth_B = np.random.normal(loc-12, scale=3, size=n)

growth_C = np.random.normal(loc-15, scale=2.5, size=n)
             # ANOVA test
f_stat, p_val = f_oneway(growth_A, growth_B, growth_C)
             w output
print("Means: A=(np.mean(growth_A):.2f), B=(np.mean(growth_B):.2f), C=(np.mean(growth_C):.2f)")
print("F: (f_stat:.4f), P: (p_val:.4f)")
if p_val <.0f.</pre>
                   p_val < 0.05:
print("Reject null: Significant differences between treatments.")
# Tukey's MSO
labels = ['A'] " n + ['B'] " n + ['C'] " n
tukey = pairwise_tukeyhsd(np.concatenate([growth_A, growth_B, growth_C]), labels, alpha=0.05)
print("Antkey's MSO Results:")
print(tukey)
### SO Results:")
                   print("Fail to reject null: No significant differences.")
             Means: A=9.67, B=11.14, C=15.27
F: 36.1214, P: 0.0000
Reject null: Significant differences between treatments.
             Tukey's HSD Results:
Multiple Comparison of Means - Tukey HSD, FWER=0.05
             group1 group2 meandiff p-adj lower upper reject
                                 B 1.4647 0.0877 -0.1683 3.0977 False
C 5.5923 0.0 3.9593 7.2252 True
C 4.1276 0.0 2.4946 5.7605 True
 [30]: import numpy as np
import pandas as pd
from sklearn.preprocessing import StandardScaler
             # Generate a small dataset
```