

EX NO: 10

DATE:

STATISTICAL HYPOTHESIS TESTING

Z-TEST, T-TEST, F-TEST

AIM:

ALGORITHM:

PROGRAM:

#SAMPLE DATASET

```
class_A_scores = [78, 85, 90, 87, 76, 82, 89, 84, 91, 77] # Sample A
```

```
class_B_scores = [72, 70, 75, 78, 80, 74, 77, 79, 73, 76] # Sample B
```

```
import numpy as np
```

```
from scipy.stats import ttest_1samp, ttest_ind, ttest_rel, f
```

```
from statsmodels.stats.weightstats import ztest
```

```
# Sample dataset
```

```
class_A_scores = [78, 85, 90, 87, 76, 82, 89, 84, 91, 77]
```

```
class_B_scores = [72, 70, 75, 78, 80, 74, 77, 79, 73, 76]
```

```
population_mean = 80
```

```
print("=== Z-Test ===")
```

```
# One-sample Z-test
```

```
z_stat, p_val = ztest(class_A_scores, value=population_mean)
```

```
print("One-sample Z-test (Class A vs Population Mean):")
```

```
print("Z-statistic:", z_stat, "| P-value:", p_val)
```

```
# Two-sample Z-test
```

```
z_stat, p_val = ztest(class_A_scores, class_B_scores)
```

```
print("\nTwo-sample Z-test (Class A vs Class B):")
```

```
print("Z-statistic:", z_stat, "| P-value:", p_val)
```

```
print("\n=== T-Test ===")
```

```
# One-sample T-test
```

```
t_stat, p_val = ttest_1samp(class_A_scores, population_mean)
```

```
print("One-sample T-test (Class A vs Population Mean):")
```

```
print("T-statistic:", t_stat, "| P-value:", p_val)
```

```
# Independent two-sample T-test

t_stat, p_val = ttest_ind(class_A_scores, class_B_scores)

print("\nIndependent Two-sample T-test (Class A vs Class B):")

print("T-statistic:", t_stat, "| P-value:", p_val)
```

```
# Paired T-test (assume scores before and after training)

before_training = [60, 65, 68, 70, 75]

after_training = [70, 75, 78, 80, 85]

t_stat, p_val = ttest_rel(before_training, after_training)

print("\nPaired T-test (Before vs After Training):")

print("T-statistic:", t_stat, "| P-value:", p_val)

print("\n=== F-Test ===")
```

```
# F-test to compare variance between Class A and Class B

var1 = np.var(class_A_scores, ddof=1)

var2 = np.var(class_B_scores, ddof=1)

f_stat = var1 / var2

df1 = len(class_A_scores) - 1

df2 = len(class_B_scores) - 1

p_val = 1 - f.cdf(f_stat, df1, df2)

print("F-statistic (Class A vs Class B variances):", f_stat)

print("P-value (one-tailed):", p_val)
```

OUTPUT:

=== Z-Test ===

One-sample Z-test (Class A vs Population Mean):

Z-statistic: 2.2396703154756357 | P-value: 0.02511233407102513

Two-sample Z-test (Class A vs Class B):

Z-statistic: 4.219056529150508 | P-value: 2.453267629329014e-05

=== T-Test ===

One-sample T-test (Class A vs Population Mean):

T-statistic: 2.2396703154756357 | P-value: 0.05187118134045702

Independent Two-sample T-test (Class A vs Class B):

T-statistic: 4.219056529150508 | P-value: 0.0005159136080406046

Paired T-test (Before vs After Training):

T-statistic: -inf | P-value: 0.0

=== F-Test ===

F-statistic (Class A vs Class B variances): 2.953463203463203

P-value (one-tailed): 0.06119106487728909

RESULT:

| | |
|------------------|---------------------------------|
| EX NO: 11 | LONGITUDINAL DEVELOPEMNT |
| DATE: | |

AIM:

ALGORITHM:

PROGRAM:

```
import pandas as pd
import numpy as np
import statsmodels.formula.api as smf

# Simulate longitudinal data
np.random.seed(1)
n_participants = 100
time_points = [0, 3, 6, 9, 12]

data = []

for pid in range(1, n_participants + 1):
    base_sleep = np.random.normal(8, 1) # baseline sleep duration (in hours)
    for t in time_points:
        usage = np.random.normal(3 + 0.2 * t, 1) # social media use slightly increases over time
        sleep_quality = base_sleep - 0.3 * usage + np.random.normal(0, 0.5) # effect of usage on sleep
        data.append([pid, t, usage, sleep_quality])

df = pd.DataFrame(data, columns=['ParticipantID', 'Month', 'SocialMediaUsage', 'SleepHours'])

# Fit linear mixed model
model = smf.mixedlm("SleepHours ~ SocialMediaUsage + Month", df, groups=df["ParticipantID"])
result = model.fit()

print(result.summary())
```

OUTPUT:

| Mixed Linear Model Regression Results | | | | | | |
|---------------------------------------|---------|---------------------|---------|------------|---------------|--------|
| ===== | | | | | | |
| Model: | MixedLM | Dependent Variable: | | SleepHours | | |
| No. Observations: | 500 | Method: | | REML | | |
| No. Groups: | 100 | Scale: | | 0.2572 | | |
| Min. group size: | 5 | Log-Likelihood: | | -522.1425 | | |
| Max. group size: | 5 | Converged: | | Yes | | |
| Mean group size: | 5.0 | | | | | |
| ----- | | | | | | |
| | Coef. | Std.Err. | z | P> z | [0.025 0.975] | |
| ----- | | | | | | |
| Intercept | 8.230 | 0.132 | 62.253 | 0.000 | 7.971 | 8.489 |
| SocialMediaUsage | -0.312 | 0.027 | -11.617 | 0.000 | -0.365 | -0.259 |
| Month | -0.016 | 0.007 | -2.229 | 0.026 | -0.030 | -0.002 |
| Group Var | 0.887 | 0.294 | | | | |
| ===== | | | | | | |

RESULT:

| | |
|-----------|--|
| EX NO: 12 | DIMENSIONALITY REDUCTION & FEATURE INSIGHTS |
| DATE: | |

AIM:

ALGORITHM:

PROGRAM:

```
# Import necessary libraries

import pandas as pd

import seaborn as sns

import matplotlib.pyplot as plt

from sklearn.preprocessing import StandardScaler

from sklearn.decomposition import PCA


# Load the dataset

df = sns.load_dataset("iris")

print("Original Dataset Shape:", df.shape)


# Drop the categorical target column for PCA

X = df.drop("species", axis=1)


# Step 1: Standardize the features

scaler = StandardScaler()

X_scaled = scaler.fit_transform(X)


# Step 2: Apply PCA

pca = PCA(n_components=2) # Reduce to 2 dimensions for visualization

X_pca = pca.fit_transform(X_scaled)


# Step 3: Create a new DataFrame with principal components

pca_df = pd.DataFrame(data=X_pca, columns=["PC1", "PC2"])

pca_df["species"] = df["species"]


# Step 4: Visualize the reduced dimensions

plt.figure(figsize=(8,6))

sns.scatterplot(x="PC1", y="PC2", hue="species", data=pca_df, palette="Set2")

plt.title("PCA - Iris Dataset")
```

```
plt.xlabel("Principal Component 1")
plt.ylabel("Principal Component 2")
plt.grid(True)
plt.show()
```

```
# Step 5: Feature Contribution (Loadings)
```

```
loadings = pd.DataFrame(pca.components_.T,
                        columns=["PC1", "PC2"],
                        index=X.columns)
```

```
print("\n Feature Contribution to Principal Components:\n")
print(loadings)
```

```
# Step 6: Plot feature contributions
```

```
plt.figure(figsize=(8,5))
loadings.plot(kind='bar')
plt.title("Feature Contribution to PC1 and PC2")
plt.ylabel("Loading Score")
plt.grid(True)
plt.xticks(rotation=0)
plt.tight_layout()
plt.show()
```

OUTPUT:

```
Original Dataset Shape: (150, 5)
```

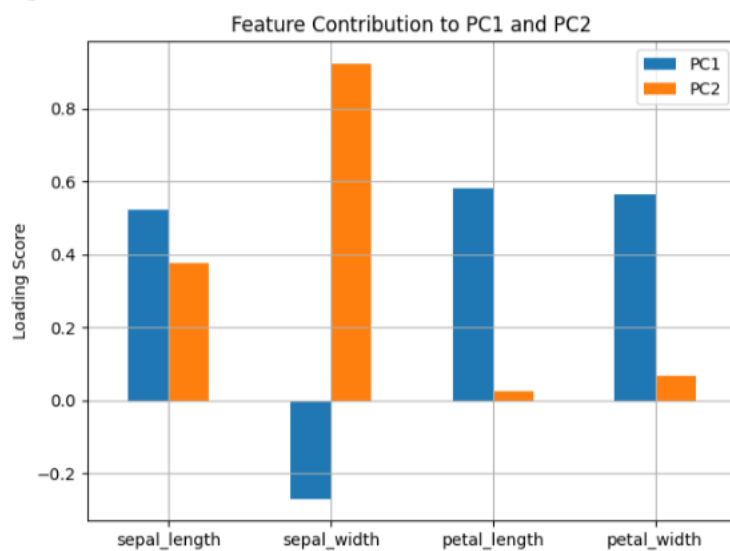


Feature Contribution to Principal Components:

```

          PC1      PC2
sepal_length 0.521066 0.377418
sepal_width  -0.269347 0.923296
petal_length 0.580413 0.024492
petal_width  0.564857 0.066942
<Figure size 800x500 with 0 Axes>

```



RESULT: