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Exp 4 : Statistical Hypothesis Testing Using SciPy and Scikit-Learn

Aim: Implementation of Statistical Hypothesis Test using Scipy and Sci-kit learn.

Problem Statement: Perform the following Tests:Correlation Tests:
a) Pearson's Correlation Coefficient
b) Spearman's Rank Correlation
c) Kendall's Rank Correlation
d) Chi-Squared Test

Introduction to Hypothesis Testing

Hypothesis testing is a statistical method used to make inferences about a population based on sample data. It helps in determining whether the observed results are due to chance or if there is a statistically significant relationship between variables.

In this experiment, we will conduct **correlation tests and a chi-squared test** using Python's `scipy.stats` library.

Theory and Output:

1.Loading dataset:

Data loading is the first step in data analysis. The dataset is stored in a CSV file and read using `pandas.read_csv()`.
The first few rows are displayed to understand the dataset structure

```
import pandas as pd

# Load the dataset
df = pd.read_csv('sc.csv') # suoermarket sales data

# Display first few rows
print(df.head())

# Display column names and data types
print(df.info())

# Summary statistics
print(df.describe())
```

```
     Invoice ID Branch       City Customer type  Gender  \
0  750-67-8428      A     Yangon         Member  Female
1  226-31-3081      C  Naypyitaw         Normal  Female
2  631-41-3108      A     Yangon         Normal    Male
3  123-19-1176      A     Yangon         Member    Male
4  373-73-7910      A     Yangon         Normal    Male

             Product line  Unit price  Quantity    Tax 5%      Total       Date  \
0       Health and beauty       74.69         7  26.1415  548.9715   1/5/2019
1  Electronic accessories       15.28         5   3.8200   80.2200   3/8/2019
2      Home and lifestyle       46.33         7  16.2155  340.5255   3/3/2019
3       Health and beauty       58.22         8  23.2880  489.0480  1/27/2019
4       Sports and travel       86.31         7  30.2085  634.3785   2/8/2019

    Time      Payment    cogs  gross margin percentage  gross income  Rating
0  13:08      Ewallet  522.83                 4.761905       26.1415     9.1
1  10:29         Cash   76.40                 4.761905        3.8200     9.6
2  13:23  Credit card  324.31                 4.761905       16.2155     7.4
3  20:33      Ewallet  465.76                 4.761905       23.2880     8.4
4  10:37      Ewallet  604.17                 4.761905       30.2085     5.3
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 17 columns):
 #   Column                   Non-Null Count  Dtype
---  ------                   --------------  -----
 0   Invoice ID               1000 non-null   object
 1   Branch                   1000 non-null   object
 2   City                     1000 non-null   object
```

2.Pearson's Correlation Coefficient:

Pearson's Correlation Coefficient (denoted as **r**) measures the **linear** relationship between two continuous variables.
Values range from **-1 to +1**:

- **+1**: Perfect positive correlation
- **0**: No correlation
- **-1**: Perfect negative correlation

The formula for Pearson's Correlation Coefficient is:

$$r = \frac{\sum(X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\sum(X_i - \bar{X})^2 \sum(Y_i - \bar{Y})^2}}$$

```
#pearson coorelation between quantity and total
from scipy.stats import pearsonr

corr, p_value = pearsonr(df['Total'], df['Quantity'])
print(f"Pearson Correlation Coefficient: {corr:.4f}")
print(f"P-value: {p_value:.4f}")
# this coorelation is significant..as p <0.005.....for perarson coorelation strong +ve = 1...strong -ve = -1.....no coorelation = 0
```

```
Pearson Correlation Coefficient: 0.7055
P-value: 0.0000
```

3.Spearman's Rank Correlation

- Spearman's Rank Correlation (denoted as ρ, rho) measures the monotonic relationship between two variables.
- It does not require normally distributed data.
- If ranks of two variables are related, it indicates correlation.
- The formula is:

$$\rho = 1 - \frac{6 \sum d_i^2}{n(n^2 - 1)}$$

```
from scipy.stats import spearmanr

corr, p_value = spearmanr(df['Customer type'], df['Rating'])
print(f"Spearman Correlation Coefficient: {corr:.4f}")
print(f"P-value: {p_value:.4f}")
```

```
Spearman Correlation Coefficient: 0.0187
P-value: 0.5552
```

4.Kendall's Rank Correlation

Theory:

- Kendall's Tau (τ) measures the **ordinal association** between two variables.
- It counts **concordant** and **discordant** pairs:
    - **Concordant pairs**: If one variable increases, the other also increases.
    - **Discordant pairs**: One increases while the other decreases.
- The formula is:

$$\tau = \frac{(C - D)}{\frac{1}{2}n(n-1)}$$

```python
from scipy.stats import kendalltau

corr, p_value = kendalltau(df['Gender'], df['Payment'])
print(f"Kendall's Rank Correlation Coefficient: {corr:.4f}")
print(f"P-value: {p_value:.4f}")
```

```
Kendall's Rank Correlation Coefficient: 0.0420
P-value: 0.1587
```

5. Chi-Squared Test

- The **Chi-Squared Test** is used for **categorical data** to check if two variables are independent.
- It compares **observed** and **expected** frequencies.
- The formula is:

$$\chi^2 = \sum \frac{(O_i - E_i)^2}{E_i}$$

```python
from scipy.stats import chi2_contingency

# Create a contingency table
contingency_table = pd.crosstab(df['Gender'], df['Product line'])

# Perform Chi-Squared test
chi2_stat, p_value, dof, expected = chi2_contingency(contingency_table)
print(f"Chi-Squared Statistic: {chi2_stat:.4f}")
print(f"P-value: {p_value:.4f}")
print(f"Degrees of Freedom: {dof}")

#p-value ≥ 0.05 → No significant relationship.
```

```
Chi-Squared Statistic: 5.7445
P-value: 0.3319
Degrees of Freedom: 5
```

Conclusion

1. **Pearson's Correlation**: Measures **linear relationship** between numerical variables. If **p < 0.05**, the correlation is significant.
2. **Spearman's Correlation**: Checks for **monotonic relationship**. If **p < 0.05**, variables move together in a ranked order.
3. **Kendall's Correlation**: Identifies **ordinal association**. A small **p-value** means a strong relationship.
4. **Chi-Square Test**: Determines **independence of categorical variables**. If **p < 0.05**, variables are dependent; otherwise, they are independent.

Final Summary:

- If **p < 0.05**, the test indicates a significant relationship.
- If **p > 0.05**, no strong relationship exists.

These tests help understand **associations** in the dataset for data-driven decisions.