Statistical Analysis Using SciPy and Statsmodels

Statistical analysis involves summarizing data, testing hypotheses, and understanding relationships between variables. Python libraries like SciPy and Statsmodels provide tools for performing such analyses efficiently.

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1. Install and Import Libraries

bash
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```
pip install scipy statsmodels
```

python
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```
import numpy as np
import pandas as pd
from scipy import stats
import statsmodels.api as sm
import statsmodels.formula.api as smf
```

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2. Descriptive Statistics

Basic Statistics

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```
# Example data
data = [12, 15, 14, 10, 13, 14, 16, 11]

# Mean, median, mode
mean = np.mean(data)
median = np.median(data)
mode = stats.mode(data)
print("Mean:", mean)
print("Median:", median)
print("Mode:", mode)
```

Variance and Standard Deviation

python
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```
variance = np.var(data, ddof=1)  # Sample variance
```

```
std_dev = np.std(data, ddof=1)   # Sample standard deviation
print("Variance:", variance)
print("Standard Deviation:", std_dev)
```

Percentiles

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```
percentiles = np.percentile(data, [25, 50, 75])  # Quartiles
print("25th, 50th, 75th Percentiles:", percentiles)
```

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3. Hypothesis Testing

T-tests

One-Sample T-Test: Compare sample mean to a population mean.

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```
t_stat, p_value = stats.ttest_1samp(data, popmean=14)
print("T-Statistic:", t_stat, "P-Value:", p_value)
```

1.

Two-Sample T-Test: Compare means of two independent samples.

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```
data1 = [12, 15, 14, 10, 13, 14]
data2 = [14, 18, 17, 15, 16, 19]
t_stat, p_value = stats.ttest_ind(data1, data2)
print("T-Statistic:", t_stat, "P-Value:", p_value)
```

2.

Paired T-Test: Compare means of two related samples.

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```
t_stat, p_value = stats.ttest_rel(data1, data2)
print("T-Statistic:", t_stat, "P-Value:", p_value)
```

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Chi-Square Test

Test for independence or goodness of fit.

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```
# Contingency table
observed = np.array([[50, 30], [20, 40]])

# Chi-square test
chi2, p, dof, expected = stats.chi2_contingency(observed)
print("Chi-Square:", chi2, "P-Value:", p)
```

ANOVA (Analysis of Variance)

Compare means of three or more groups.

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```
group1 = [14, 15, 13, 16]
group2 = [22, 21, 19, 23]
group3 = [30, 29, 27, 31]

f_stat, p_value = stats.f_oneway(group1, group2, group3)
print("F-Statistic:", f_stat, "P-Value:", p_value)
```

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4. Correlation and Regression

Correlation

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```
# Example data
x = [1, 2, 3, 4, 5]
y = [10, 20, 30, 40, 50]

# Pearson correlation
correlation, p_value = stats.pearsonr(x, y)
print("Pearson Correlation:", correlation)

# Spearman correlation
correlation, p_value = stats.spearmanr(x, y)
print("Spearman Correlation:", correlation)
```

Linear Regression

Using SciPy
python
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```
slope, intercept, r_value, p_value, std_err = stats.linregress(x, y)
print("Slope:", slope, "Intercept:", intercept)
```

1.

Using Statsmodels
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```
# Prepare data
X = sm.add_constant(x)  # Add intercept
model = sm.OLS(y, X).fit()
print(model.summary())
```

2.

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5. Advanced Regression with Statsmodels

Multiple Linear Regression
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```
# Example DataFrame
df = pd.DataFrame({
    'y': [1, 2, 3, 4, 5],
    'x1': [10, 20, 30, 40, 50],
    'x2': [5, 4, 3, 2, 1]
})

# Define model
model = smf.ols('y ~ x1 + x2', data=df).fit()
print(model.summary())
```

Logistic Regression
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```
# Binary target variable
df['target'] = [0, 1, 0, 1, 0]

# Logistic regression
log_model = smf.logit('target ~ x1 + x2', data=df).fit()
```

```
print(log_model.summary())
```

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6. Time Series Analysis

Autocorrelation

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```
from statsmodels.graphics.tsaplots import plot_acf

# Plot autocorrelation
plot_acf(data, lags=10)
```

ARIMA Model

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```
from statsmodels.tsa.arima.model import ARIMA

# Fit ARIMA model
model = ARIMA(data, order=(1, 1, 1))
model_fit = model.fit()
print(model_fit.summary())
```

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7. Probability Distributions

Generate Random Data

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```
# Normal distribution
data = np.random.normal(loc=0, scale=1, size=1000)

# Uniform distribution
data = np.random.uniform(low=0, high=10, size=1000)
```

Fit and Test Distributions

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```
# Fit normal distribution
params = stats.norm.fit(data)
```

```
# Test goodness of fit
stat, p_value = stats.kstest(data, 'norm', args=params)
print("K-S Statistic:", stat, "P-Value:", p_value)
```

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8. Visualization

Use `matplotlib` or `seaborn` for visualization of statistical results.

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```
import matplotlib.pyplot as plt
import seaborn as sns

# Boxplot
sns.boxplot(data=data)
plt.show()

# Histogram
sns.histplot(data=data, kde=True)
plt.show()
```