Experiment 1,2 - advertising.csv

Experiment 3,4,5 - smarket.csv (exp3 is summery)

Experiment 6,7,8,9,10 - collegedata.csv (exp7 is summery)

Experiment 1:

Least squares coefficient for TV: 0.05546477046955886

Experiment 2:

```
In [6]: 1 import statsmodels.api as sm
2 x = data['TV'].values.reshape(-1, 1)
3 y =data['Sales'].values
4 x = sm.add_constant(x)
5 model = sm.OLS(y, x).fit()
6 t_stat = model.tvalues[1] #T - Statistics
7 rse = model.mse_resid ** 0.5 #Residual Standard Error
8 f_stat = model.fvalue #F - Statistic
9 rss = model.ssr #Residual Sum of Squares Error
10 print("T-Statistic:", t_stat)
11 print("Residual Standard Error:", rse)
12 print("F-Statistic:", f_stat)
13 print("Residual Sum of Squares (RSS):", rss)
```

T-Statistic: 29.260497480686528

Residual Standard Error: 2.2957457136214456

F-Statistic: 856.1767128172628

Residual Sum of Squares (RSS): 1043.5487795590257

Experiment 3:

Write about the statistically significant - hypothesis

Experiment 4:

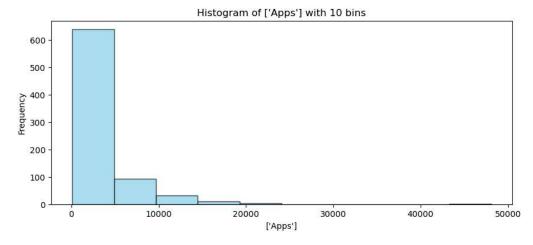
Experiment 5:

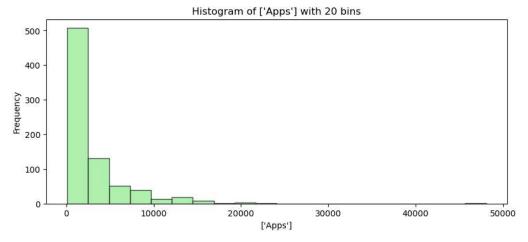
```
In [51]:
           1 from sklearn.neighbors import KNeighborsClassifier
           2 from sklearn.model selection import train test split
           3 from sklearn.preprocessing import StandardScaler
           4 | from sklearn.metrics import r2_score
           6 data = pd.read_csv('smarket.csv')
           7 | data['Direction'] = data['Direction'].map({'Up': 1, 'Down': 0})
           8 | X = data[['Lag1', 'Lag2', 'Lag3', 'Lag4', 'Lag5', 'Volume']]
           9 y = data['Direction']
          10 X_train, X_test, y_train, y_test = train_test_split(X, y, test_
          11
          12 knn = KNeighborsRegressor(n_neighbors=3)
          13 knn.fit(X_train, y_train)
          14 y_pred = knn.predict(X_test)
          15 # Compute R-squared, Mallow's Cp, AIC and BIC:
          16 adj_r_squared = 1 - (1 - r2_score(y_test, y_pred)) * ((len(y_te
          17 mse = np.mean((y_test - y_pred) ** 2)
          18 | cp = mse + 2 * X_test.shape[1] * (mse / len(y_test))
          19 aic = len(y_test) * np.log(mse) + 2 * X_test.shape[1]
          20 bic = len(y_test) * np.log(mse) + X_test.shape[1] * np.log(len(
          21
          22 | print("Adjusted R-squared:", adj_r_squared)
          23 print("Mallow's Cp:", cp)
          24 print("AIC:", aic)
          25 print("BIC:", bic)
```

Adjusted R-squared: -0.3285229007660486 Mallow's Cp: 0.338620444444445 AIC: -270.43975441623655 BIC: -249.31098890906307

Experiment 6:

```
In [76]:
              import pandas as pd
             import matplotlib.pyplot as plt
           2
             college_data = pd.read_csv('collegedata.csv')
             var = ['Apps'] # you can add some more variables and using a fo
             plt.figure(figsize=(10, 4))
           6
           7
             plt.hist(college_data[var], bins=10, color='skyblue', edgecolor
             plt.title(f'Histogram of {var} with 10 bins')
           9
             plt.xlabel(var)
          10 plt.ylabel('Frequency')
          11 plt.show()
          12 plt.figure(figsize=(10, 4))
          13 plt.hist(college_data[var], bins=20, color='lightgreen', edgeco
          14 plt.title(f'Histogram of {var} with 20 bins')
          15 plt.xlabel(var)
          16 plt.ylabel('Frequency')
             plt.show()
          17
```





Experiment 7:Write about the histograms that you have observed

Experiment 8:

```
In [87]:
             from scipy.stats import pearsonr, spearmanr, kendalltau
             selected_variables = ['Apps', 'Accept', 'Enroll']
           2
             # Calculate Pearson correlation, Spearman correlation, Kendall
           4
             pearson corr = college data[selected variables].corr(method='pe
             spearman_corr = college_data[selected_variables].corr(method='s
           7
             kendall corr = college data[selected variables].corr(method='ke
           9
             print("Pearson Correlation:\n",pearson_corr)
          10 | print("\nSpearman Correlation:\n", spearman corr)
             print("\nKendall Correlation:\n",kendall corr)
         Pearson Correlation:
                                        Enroll
                      Apps
                              Accept
                 1.000000 0.943451 0.846822
         Apps
         Accept 0.943451 1.000000
                                     0.911637
         Enroll 0.846822 0.911637 1.000000
         Spearman Correlation:
                                       Enroll
                      Apps
                            Accept
         Apps
                 1.000000 0.97939 0.926169
         Accept 0.979390 1.00000 0.946400
         Enroll 0.926169
                           0.94640 1.000000
         Kendall Correlation:
                      Apps
                                        Enroll
                              Accept
         Apps
                 1.000000 0.886006
                                     0.763762
         Accept 0.886006 1.000000
                                     0.801569
         Enroll 0.763762
                           0.801569 1.000000
```

Experiment 10:

```
In [85]:
             import numpy as np
           2 array = np.array([[1, 2],
           3
             [2, 1]
           4 | # Compute eigenvalues and eigenvectors
           5 eigenvalues, eigenvectors = np.linalg.eig(array)
             # Print eigenvalues and eigenvectors
             print("Eigenvalues:")
           7
           8 print(eigenvalues)
           9 print("Eigenvectors:")
          10 print(eigenvectors)
         Eigenvalues:
         [ 3. -1.]
         Eigenvectors:
         [[ 0.70710678 -0.70710678]
          [ 0.70710678  0.70710678]]
```

Experiment 9:

```
In [130]:
              import pandas as pd
            2 from scipy.stats import ttest_1samp, ttest_rel, ttest_ind, mann
            3 college data = pd.read csv('collegedata.csv')
            5 t statistic, p value = ttest 1samp(college data['Apps'], 1000)
            6 print("Simple Hypothesis T-test statistic: ",t_statistic)
            7
            8 group1 = college data[college data['Private'] == 'Yes']['Accept
              group2 = college data[college data['Private'] == 'No']['Accept'
            9
           10 | t statistic, p value = ttest ind(group1, group2)
              print("\nStudent T-test statistic:",t statistic)
           12
           13 t_statistic, p_value = ttest_rel(college_data['Apps'], college_
              print("\nPaired t-test between Apps and Accept: ",t statistic)
           14
           15
           16 | group1 = college_data[college_data['Private'] == 'Yes']['Accept
              group2 = college data[college data['Private'] == 'No']['Accept'
           18 | u statistic, p value = mannwhitneyu(group1, group2)
           19
              print("\nU test statistic:",u statistic)
           20
           21 | correlation = college_data[['Apps', 'Accept', 'Enroll']].corr(m
           22
              print("\nCorrelation:\n",correlation)
           23
           24 | covariance = college data['Apps'].cov(college data['Accept'])
           25 print("\nCovariance", covariance)
           26
           27 | contingency_table = pd.crosstab(college_data['Apps'], college_d
              chi2_statistic, p_value, _, _ = chi2_contingency(contingency_ta
              print("\nChi-square Test statistic:",chi2_statistic)
          Simple Hypothesis T-test statistic: 14.416590015808657
          Student T-test statistic: -15.037175252579376
          Paired t-test between Apps and Accept: 15.593495811336158
          U test statistic: 21811.0
          Correlation:
                       Apps
                               Accept
                                         Enroll
          Apps
                  1.000000 0.943451 0.846822
          Accept 0.943451 1.000000 0.911637
          Enroll 0.846822 0.911637 1.000000
          Covariance 8949859.811893819
          Chi-square Test statistic: 494905.8333333333
```