```
In [1]: # import required libraries
  import pandas as pd
  import statsmodels.api as sm
  from statsmodels.formula.api import ols
  from scipy import stats
```

```
In [2]: import pandas as pd
        import numpy as np
        np.random.seed(123)
        n = 150
        gender = np.random.choice(['Male', 'Female'], size=n)
        age_group = np.random.choice(['Young', 'Middle-aged', 'Old'], size=n)
        beauty = np.random.normal(
            loc=[6.5 if a == 'Young' else 5.5 if a == 'Middle-aged' else 4.5 for a in ag
            scale=1.2,
            size=n
        beauty = np.clip(beauty, 1, 10)
        eval_score = (
            0.6 * beauty +
            np.where(gender == 'Female', 0.4, -0.2) +
            np.random.normal(0, 1.2, size=n)
        eval_score = np.clip(eval_score, 1, 10)
        data = pd.DataFrame({
            'gender': gender,
            'age_group': age_group,
            'beauty': beauty.round(2),
            'eval': eval_score.round(2)
        })
        data.to_csv("teacher_evaluations.csv", index=False)
        data
```

```
gender
                       age_group beauty eval
Out[2]:
                                    5.90 2.60
               Male
                          Young
           1 Female Middle-aged
                                    5.78 1.90
           2
               Male
                          Young
                                    6.27 5.29
           3
               Male Middle-aged
                                    6.77 3.33
                             Old
           4
               Male
                                    1.53 1.00
         145
               Male Middle-aged
                                    4.35 1.08
         146 Female
                          Young
                                    4.89 4.27
         147 Female Middle-aged
                                    6.68 3.46
         148 Female
                          Young
                                    6.44 3.21
         149 Female Middle-aged
                                    6.68 6.92
```

150 rows × 4 columns

```
In [3]: # Q1. Regression with T-test: Does gender affect teaching evaluation rates?

# assuming columns are named 'eval' (teaching evaluation) and 'gender' (male/fem model_ttest = ols('eval ~ gender', data=data).fit()
    ttest_result = model_ttest.t_test([0, 1]) # tests coefficient of gender variabl
    print("\nQ1. Regression with T-test: Does gender affect teaching evaluation rate print(model_ttest.summary())
```

Q1. Regression with T-test: Does gender affect teaching evaluation rates?

OLS Regression Results

\_\_\_\_\_\_ Dep. Variable: eval R-squared: 0.076 Model: OLS Adj. R-squared: 0.070 Least Squares F-statistic: Method: 12.18 Tue, 28 Oct 2025 Prob (F-statistic): Date: 0.000637 Time: 12:08:36 Log-Likelihood: -260.25 No. Observations: 524.5 150 AIC: Df Residuals: 148 BIC: 530.5 Df Model: 1 Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
Intercept	3.8953	0.158	24.590	0.000	3.582	4.208
<pre>gender[T.Male]</pre>	-0.7870	0.226	-3.490	0.001	-1.233	-0.341
===========	========	========	========	========	========	=====
Omnibus:		1.670	Durbin-Watson:		1.871	
Prob(Omnibus):		0.434	Jarque-Bera (JB):		1.467	
Skew:		0.102	Prob(JB):		0.480	
Kurtosis:		2.561	Cond. No.		2.60	

## Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correctly spec

```
In [4]: # Q2. Regression with ANOVA: Does beauty score differ by age group?
model_anova = ols('beauty ~ C(age_group)', data=data).fit()
```

```
print("\nQ2. Regression with ANOVA: Does beauty score for instructors differ by
     print(anova_table)
    Q2. Regression with ANOVA: Does beauty score for instructors differ by age?
                 sum_sq df F PR(>F)
    C(age group) 90.781678 2.0 27.900131 5.349196e-11
    Residual 239.154911 147.0 NaN
                                   NaN
In [5]: # Q3 . Correlation: Using the teachers' rating dataset, Is teaching evaluation s
     import statsmodels.api as sm
     from statsmodels.formula.api import ols
     # perform OLS regression: eval vs beauty
     model = ols('eval ~ beauty', data=data).fit()
     # display summary
     print(model.summary())
                       OLS Regression Results
     ______
    Dep. Variable:
                           eval R-squared:
    Model:
                            OLS Adj. R-squared:
                                                       0.235
                   Least Squares F-statistic:
    Method:
                                                       46.86
                 Tue, 28 Oct 2025 Prob (F-statistic):
    Date:
                                                  1.89e-10
                        12:08:55 Log-Likelihood:
    Time:
                                                     -245.55
    No. Observations:
                            150 AIC:
                                                       495.1
    Df Residuals:
                            148 BIC:
                                                       501.1
                             1
    Df Model:
    Covariance Type:
                       nonrobust
    ______
              coef std err t P>|t| [0.025 0.975]
     -----
              0.9022 0.394 2.290 0.023 0.124
    Intercept
              0.4719
                      0.069
                              6.846
                                     0.000
                                              0.336
     ______
    Omnibus:
                          0.729 Durbin-Watson:
                                                       1.938
                          0.695 Jarque-Bera (JB):
    Prob(Omnibus):
                                                      0.854
                         -0.142 Prob(JB):
    Skew:
                                                      0.652
                          2.764 Cond. No.
    Kurtosis:
                                                       22.7
     ______
    Notes:
     [1] Standard Errors assume that the covariance matrix of the errors is correctly spec
```

In [ ]:

anova table = sm.stats.anova lm(model anova, typ=2)