

Multiple Linear Regression

Comparing Regression Lines for Two Groups

We want to relate course evaluation scores (Y) to the beauty score assigned to the instructor (X_1) and the gender of (female v. male).

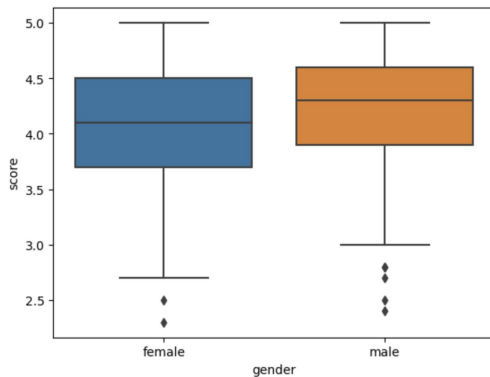
Questions we could ask:

1. Is there a difference in the **mean** *course evaluation score* between *female and male instructors*?
2. Is there a linear relationship between course evaluation scores and beauty scores?
3. Is there a difference in the mean course evaluation score between female and male instructors, after accounting for the beauty score?
4. Is the relationship between beauty score and course evaluation score the same for male and female instructors?

Comparing Regression Lines for Two Groups

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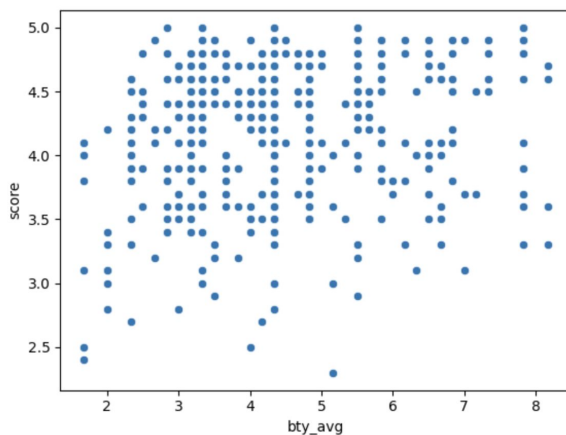
1. Is there a difference in the **mean** course evaluation score between *female* and *male* instructors?



Comparing Regression Lines for Two Groups

We want to relate course evaluation scores (Y) to the beauty score assigned to the instructor (X_1) and the gender of (female v. male).

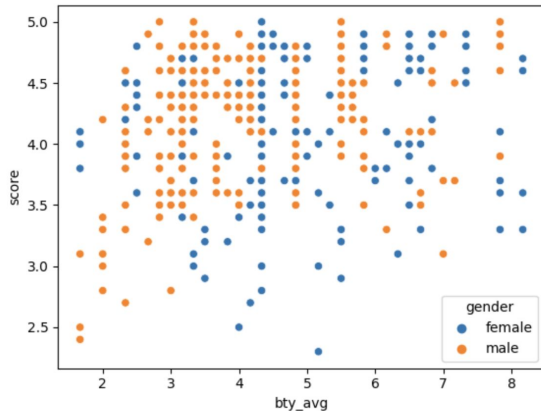
2. Is there a linear relationship between course evaluation scores and beauty scores?



Comparing Regression Lines for Two Groups

We want to relate course evaluation scores (Y) to the beauty score assigned to the instructor (X_1) and the gender of (female v. male).

3. Is there a difference in the mean course evaluation score between female and male instructors, after accounting for the beauty score?



The Regression Model: Common (Parallel) Slope

Let $\text{Gender} = \begin{cases} 1, & \text{Male} \\ 0, & \text{Female} \end{cases}$

Then,

We have two lines:

- Female
- Male

Interpretation:

- β_0 :
- β_1 :
- β_2 :

Interpretation

```
=====
                        OLS Regression Results
=====
Dep. Variable:          score    R-squared:          0.059
Model:                  OLS      Adj. R-squared:       0.055
Method:                 Least Squares    F-statistic:      14.45
Date:                   Tue, 25 Jul 2023    Prob (F-statistic): 8.18e-07
Time:                   16:28:11    Log-Likelihood:    -360.37
No. Observations:      463    AIC:                726.7
Df Residuals:          460    BIC:                739.1
Df Model:              2
Covariance Type:        nonrobust
=====
                        coef    std err          t      P>|t|      [0.025    0.975]
-----
Intercept              3.7473      0.085    44.266     0.000      3.581      3.914
gender[T.male]         0.1724      0.050     3.433     0.001      0.074      0.271
bty_avg                0.0742      0.016     4.563     0.000      0.042      0.106
=====
Omnibus:                30.145    Durbin-Watson:      1.277
Prob(Omnibus):          0.000    Jarque-Bera (JB):    34.960
Skew:                   -0.672    Prob(JB):            2.56e-08
Kurtosis:               2.925    Cond. No.            17.5
=====
```

Notes:

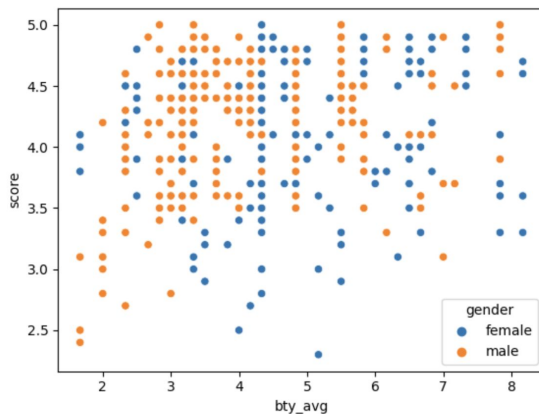
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

	df	sum_sq	mean_sq	F	PR(>F)
gender	1.0	2.260213	2.260213	8.086320	0.004659
bty_avg	1.0	5.819173	5.819173	20.819135	0.000006
Residual	460.0	128.574955	0.279511	NaN	NaN

Comparing Regression Lines for Two Groups

We want to relate course evaluation scores (Y) to the beauty score assigned to the instructor (X_1) and the gender of (female v. male).

4. Is the relationship between beauty score and course evaluation score the same for male and female instructors?



The Regression Model: Different Slopes

Let $\text{Gender} = \begin{cases} 1, & \text{Male} \\ 0, & \text{Female} \end{cases}$

Then,

We have two lines:

- Male
- Female

Interpretation:

- $H_0: \beta_3 = 0 \Rightarrow$
- $H_0: \beta_1 = 0 \Rightarrow$

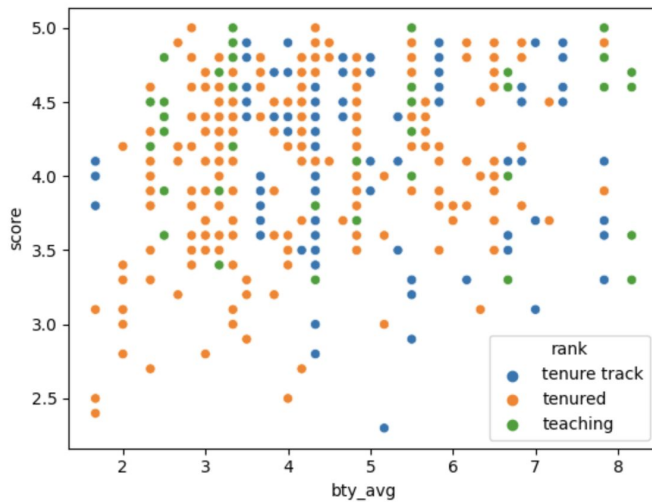
Interpretation

```
=====
                        OLS Regression Results
=====
Dep. Variable:          score      R-squared:                0.071
Model:                  OLS       Adj. R-squared:            0.065
Method:                 Least Squares   F-statistic:           11.74
Date:                   Tue, 25 Jul 2023   Prob (F-statistic):    2.00e-07
Time:                   18:00:28         Log-Likelihood:       -357.35
No. Observations:       463             AIC:                  722.7
Df Residuals:           459             BIC:                  739.3
Df Model:                3
Covariance Type:        nonrobust
=====
                        coef      std err          t      P>|t|      [0.025      0.975]
-----
Intercept              3.9501         0.118     33.475     0.000         3.718         4.182
gender[T.male]         -0.1835         0.153     -1.196     0.232        -0.485         0.118
bty_avg                0.0306         0.024      1.277     0.202        -0.017         0.078
bty_avg:gender[T.male]  0.0796         0.032      2.452     0.015         0.016         0.143
=====
Omnibus:                26.631      Durbin-Watson:           1.282
Prob(Omnibus):           0.000      Jarque-Bera (JB):        30.276
Skew:                    -0.624      Prob(JB):                2.67e-07
Kurtosis:                2.890      Cond. No.                 42.3
=====
```

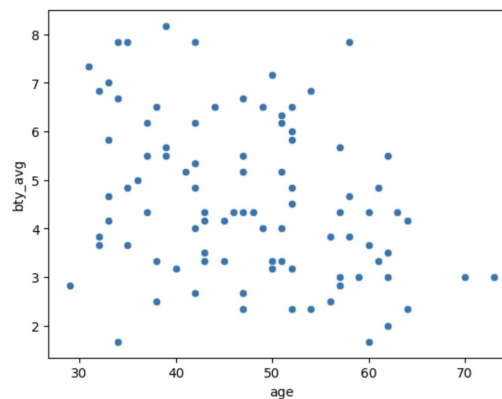
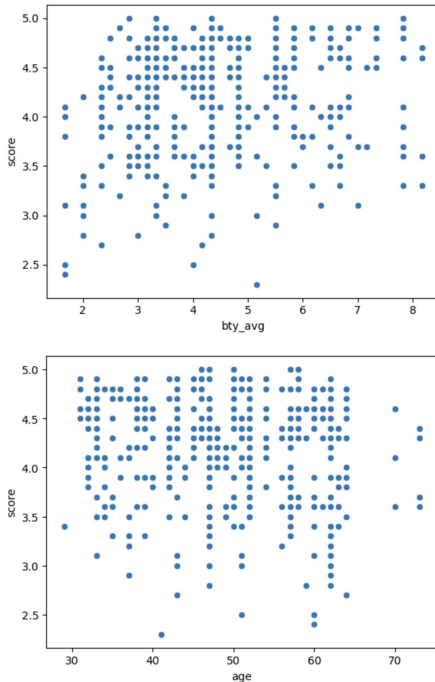
Notes:
[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

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=====
                        df      sum_sq      mean_sq          F      PR(>F)
-----
gender              1.0      2.260213      2.260213      8.174440      0.004442
bty_avg             1.0      5.819173      5.819173     21.046010      0.000006
bty_avg:gender      1.0      1.662530      1.662530      6.012817      0.014574
Residual            459.0     126.912425      0.276498           NaN           NaN
=====
```

What if the categorical variable has more than 2 levels?



What if I want to use more than one quantitative variable?



Effects of Multicollinearity

- If predictors are highly correlated amongst themselves, then the estimated regression coefficients and tests can be:
- The regression tests can be difficult to interpret individually
- One variable alone might work just as well as many
- Explore the potential for multicollinearity by examining scatterplots of the response and the predictors (matrix plot)

Variance Inflation Factor (VIF)

- The variance of the coefficients of correlated predictors is inflated
- The Variance Inflation Factor (VIF) reflects
- For each Predictor X_i , regress X_i onto the other predictors. Record R_i^2 .
Then, for the i^{th} predictor,

$$VIF_i = \frac{1}{1 - R_i^2}$$

- Be suspicious of multicollinearity when