

# Inference

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COGS 108 Fall 2025  
Jason Chen  
Week 6

xic007@ucsd.edu  
OH: Tue 3-5 pm

# Due dates

- A2: Wednesday (11/05)
- D5: Friday (11/07)

# Project Proposal

- Feedback will be out by the end of today
- Make revisions in next checkpoint

## Part II : quadratic transformation

Choose quad\_b so that

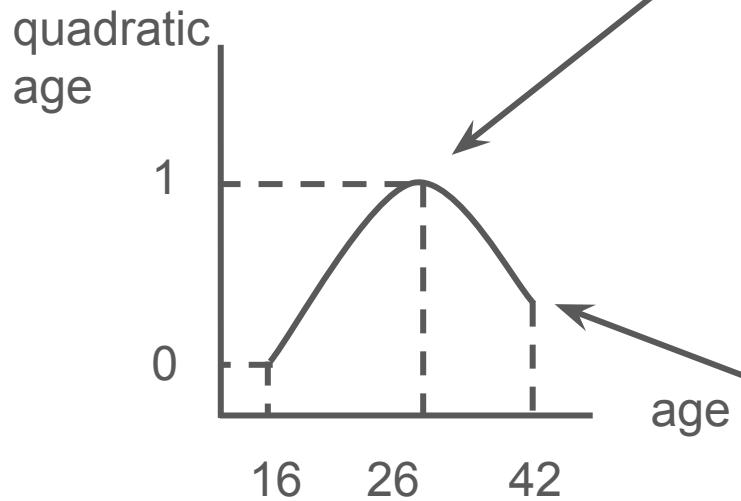
$$\max(\text{quad\_a} * (\text{age} - 26)^2 + \text{quad\_b}) = 1$$

$$\max(\text{quad\_a} * (\text{age} - 26)^2 + \text{quad\_b})$$

$$= \text{quad\_a} * (26 - 26)^2 + \text{quad\_b}$$

$$= \text{quad\_b}$$

$$\text{quad\_b} = ?$$



Choose quad\_a so that for the max value of  $(\text{age} - 26)^2$ ,

$$\text{quad\_a} * \max((\text{age} - 26)^2) = 1$$

$$\max((\text{age} - 26)^2) = (42 - 26)^2 = 256$$

$$\text{quad\_a} = 1 / \max((\text{age} - 26)^2) = ?$$

# Part III : statsmodels.regression.linear\_model.OLS

What is **OLS** (Ordinary Least Squares)?

A fundamental method for estimating relationships between variables in **regression** analysis.

It fits a **line** (or plane) through the data that **minimizes the sum of squared differences** between the observed values and the model's predicted values.

**Introduction :**

A linear regression model establishes the relation between a dependent variable(**y**) and at least one independent variable(**x**) as :

$$\hat{y} = b_1x + b_0$$

In *OLS* method, we have to choose the values of  $b_1$  and  $b_0$  such that, the total sum of squares of the difference between the calculated and observed values of **y**, is minimised.

**Formula for OLS:**

$$S = \sum_{i=1}^n (y_i - \hat{y}_i)^2 = \sum_{i=1}^n (y_i - b_1x_i - b_0)^2 = \sum_{i=1}^n (\epsilon_i)^2 = \text{min}$$

Where,

$\hat{y}_i$ = predicted value for the *i*th observation

$y_i$ = actual value for the *i*th observation

$\epsilon_i$ = error/residual for the *i*th observation

*n* = total number of observations

To get the values of  $b_0$  and  $b_1$  which minimise *S*, we can take a partial derivative for each coefficient and equate it to zero.

# Part III : statsmodels.regression.linear\_model.OLS

What OLS does

Estimates how **changes** in predictor variables affect the **outcome variable**.

Provides estimates for:

**Coefficients ( $\beta$ )** → strength and direction of relationships.

**Intercept** → expected value of the outcome when predictors are 0.

**R<sup>2</sup> and F-statistic** → overall model fit and significance.

**p-values** → how likely it is that each effect (or the model overall) occurred by chance.

**Introduction :**

A linear regression model establishes the relation between a dependent variable( $y$ ) and at least one independent variable( $x$ ) as :

$$\hat{y} = b_1x + b_0$$

In *OLS* method, we have to choose the values of  $b_1$  and  $b_0$  such that, the total sum of squares of the difference between the calculated and observed values of  $y$ , is minimised.

**Formula for OLS:**

$$S = \sum_{i=1}^n (y_i - \hat{y}_i)^2 = \sum_{i=1}^n (y_i - b_1x_i - b_0)^2 = \sum_{i=1}^n (\epsilon_i)^2 = \min$$

Where,

$\hat{y}_i$ = predicted value for the  $i$ th observation

$y_i$ = actual value for the  $i$ th observation

$\epsilon_i$ = error/residual for the  $i$ th observation

$n$  = total number of observations

To get the values of  $b_0$  and  $b_1$  which minimise  $S$ , we can take a partial derivative for each coefficient and equate it to zero.

# Part III : statsmodels.regression.linear\_model.OLS

What  $R^2$  means

Proportion of variance in  
the outcome explained by  
the model (0–1).

Higher  $R^2 \Rightarrow$  better  
in-sample fit (for the same  
outcome and dataset).

OLS Regression Results						
Dep. Variable:	value_eur	R-squared:	0.452			
Model:	OLS	Adj. R-squared:	0.452			
Method:	Least Squares	F-statistic:	4038.			
Date:	Mon, 03 Nov 2025	Prob (F-statistic):	0.00			
Time:	16:50:50	Log-Likelihood:	-82386.			
No. Observations:	4902	AIC:	1.648e+05			
Df Residuals:	4900	BIC:	1.648e+05			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]
Intercept	-3.674e+07	6.28e+05	-58.534	0.000	-3.8e+07	-3.55e+07
overall	5.991e+05	9428.092	63.542	0.000	5.81e+05	6.18e+05
=====						
Omnibus:	5334.671	Durbin-Watson:	0.117			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	543549.845			
Skew:	5.433	Prob(JB):	0.00			
Kurtosis:	53.429	Cond. No.	607.			
=====						

Notes:

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

# Part III : statsmodels.regression.linear\_model.OLS

# What the p-value (F-statistic) means

Tests whether the model as a whole explains a significant amount of variance.

Null hypothesis: all regression coefficients = 0 (no predictive value).

A small p-value (e.g.,  $< 0.05$ )  $\Rightarrow$   
at least one predictor significantly  
improves model fit.

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OLS Regression Results
=====
Dep. Variable: value_eur R-squared: 0.452
Model: OLS Adj. R-squared: 0.452
Method: Least Squares F-statistic: 4038.
Date: Mon, 03 Nov 2025 Prob (F-statistic): 0.00
Time: 16:50:50 Log-Likelihood: -82386.
No. Observations: 4902 AIC: 1.648e+05
Df Residuals: 4900 BIC: 1.648e+05
Df Model: 1
Covariance Type: nonrobust
=====

            coef    std err          t      P>|t|      [0.025      0.975]
Intercept -3.674e+07  6.28e+05     -58.534     0.000    -3.8e+07    -3.55e+07
overall    5.991e+05  9428.092      63.542     0.000     5.81e+05    6.18e+05
=====

Omnibus: 5334.671 Durbin-Watson: 0.117
Prob(Omnibus): 0.000 Jarque-Bera (JB): 543549.845
Skew: 5.433 Prob(JB): 0.00
Kurtosis: 53.429 Cond. No. 607.
=====

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

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# Part III : model comparison

How to **compare** two OLS models

1. Compare **R<sup>2</sup>** – higher means better overall fit.
2. Check **Prob (F-statistic)** – must be < 0.05 for the model to be statistically significant.

This model is the winner!

Dep. Variable:	log_value_eur	R-squared:	0.890
Model:	OLS	Adj. R-squared:	0.890
Method:	Least Squares	F-statistic:	3.973e+04
Date:	Mon, 03 Nov 2025	Prob (F-statistic):	0.00
Time:	16:50:50	Log-Likelihood:	669.09
No. Observations:	4902	AIC:	-1334.
Df Residuals:	4900	BIC:	-1321.
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
Intercept	0.5053	0.028	18.373	0.000	0.451	0.559
overall	0.0823	0.000	199.314	0.000	0.082	0.083

Omnibus:	6866.292	Durbin-Watson:	0.729
Prob(Omnibus):	0.000	Jarque-Bera (JB):	6398339.823
Skew:	-7.659	Prob(JB):	0.00
Kurtosis:	179.328	Cond. No.	607.

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

OLS Regression Results						
Dep. Variable:	value_eur	R-squared:	0.452			
Model:	OLS	Adj. R-squared:	0.452			
Method:	Least Squares	F-statistic:	4038.			
Date:	Mon, 03 Nov 2025	Prob (F-statistic):	0.00			
Time:	16:50:50	Log-Likelihood:	-82386.			
No. Observations:	4902	AIC:	1.648e+05			
Df Residuals:	4900	BIC:	1.648e+05			
Df Model:	1					
Covariance Type:	nonrobust					

	coef	std err	t	P> t	[0.025	0.975]
Intercept	-3.674e+07	6.28e+05	-58.534	0.000	-3.8e+07	-3.55e+07
overall	5.991e+05	9428.092	63.542	0.000	5.81e+05	6.18e+05

Omnibus:	5334.671	Durbin-Watson:	0.117
Prob(Omnibus):	0.000	Jarque-Bera (JB):	543549.845
Skew:	5.433	Prob(JB):	0.00
Kurtosis:	53.429	Cond. No.	607.

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

### OLS Regression Results

Dep. Variable:	potential	R-squared:	0.009			
Model:	OLS	Adj. R-squared:	0.009			
Method:	Least Squares	F-statistic:	32.28			
Date:	Wed, 05 Nov 2025	Prob (F-statistic):	7.18e-33			
Time:	22:24:18	Log-Likelihood:	-55680.			
No. Observations:	17244	AIC:	1.114e+05			
Df Residuals:	17238	BIC:	1.114e+05			
Df Model:	5					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
Intercept	72.7900	0.173	419.910	0.000	72.450	73.130
side[T.Left]	-1.2958	0.216	-5.997	0.000	-1.719	-0.872
side[T.Right]	-1.0691	0.307	-3.487	0.000	-1.670	-0.468
preferred_foot[T.Right]	-1.0401	0.190	-5.476	0.000	-1.412	-0.668
side[T.Left]:preferred_foot[T.Right]	1.6812	0.263	6.392	0.000	1.166	2.197
side[T.Right]:preferred_foot[T.Right]	0.0619	0.329	0.188	0.850	-0.582	0.706
Omnibus:	179.576	Durbin-Watson:	0.039			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	185.092			
Skew:	0.254	Prob(JB):	6.43e-41			
Kurtosis:	2.981	Cond. No.	15.5			

# Section Materials

Section materials can be accessed at:

[https://github.com/JasonC1217/COGS108\\_FA25\\_B07-B08](https://github.com/JasonC1217/COGS108_FA25_B07-B08)



SCAN ME

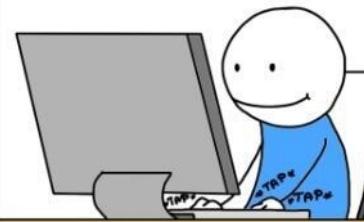
# THANKS!

Questions on EdStem or office hours

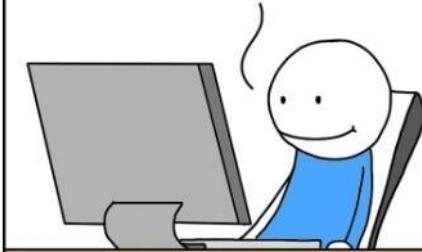
Office hours: Tue, 3-5 PM

## UNFINISHED WORK

FRIDAY EVENING



PERFECT!  
I'LL FINISH  
THIS ON  
MONDAY



MONDAY MORNING...

