

# Machine Learning with Python: A Hands-On Introduction

[https://github.com/cbrownley/2024MLWEEK\\_MLWITHPYTHON](https://github.com/cbrownley/2024MLWEEK_MLWITHPYTHON)

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# Agenda (too much...we'll take our time : )

8:30-8:45 – Setup and Overview

8:45-9:30 – **Data preprocessing**

9:30-10:00 – Hands-on: Data preprocessing

10:00-10:30 – **Cross-validation**

10:30-11:00 – Hands-on: Cross-validation

11:00-11:30 – Hands-on: K-fold cross-validation

11:30-12:00 – **Classification** (breast tumor diagnosis)

12:00-12:30 – Hands-on: Classification

12:30-1:00 – Hands-on: Decision Trees

1:00-1:30 – **Regression** (california housing)

1:30-2:00 – Hands-on: Regression

2:00-2:30 – Hands-on: **Shrinkage methods**

2:30-3:00 – Classification (credit card fraud)

3:00-3:30 – Regression (cycling counts)

3:30-4:00 – Hands-on: Classification (hotel bookings)

# Prediction vs Causal Inference

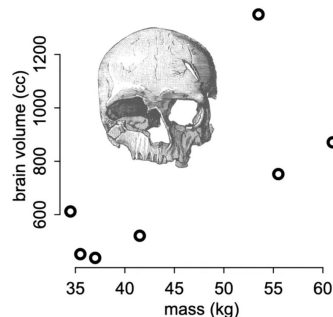
## Problems of Prediction

What function describes these points?  
(fitting, compression)

What function explains these points?  
(causal inference)

What would happen if we changed a point's mass? (intervention)

What is the next observation from the same process? (prediction)



## Good & Bad Controls

**“Control” variable:** Variable introduced to an analysis so that a causal estimate is possible

Common **wrong** heuristics for choosing control variables

Anything in the spreadsheet **YOLO!**

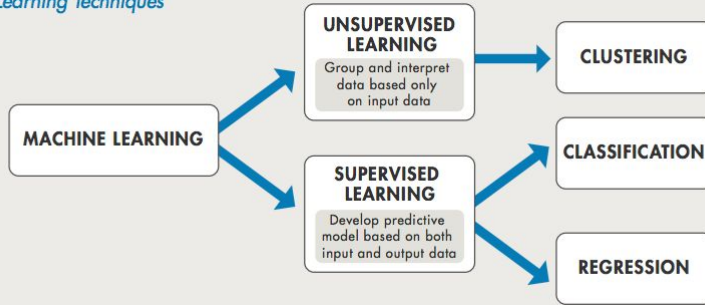
Any variables not highly **collinear**

Any **pre-treatment** measurement (baseline)



# Supervised vs Unsupervised

*Machine Learning Techniques*



## Supervised Learning

X <sub>1</sub>	X <sub>2</sub>	X <sub>3</sub>	X <sub>p</sub>	Y

Target

## Un-Supervised Learning

X <sub>1</sub>	X <sub>2</sub>	X <sub>3</sub>	X <sub>p</sub>	Y

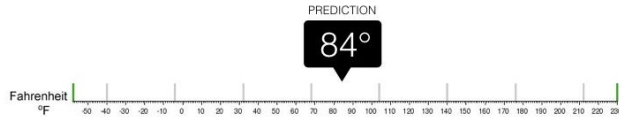
No Target

# Regression vs Classification



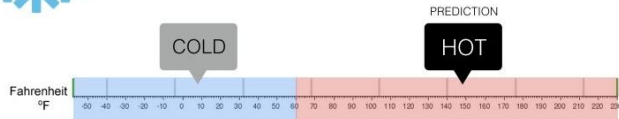
## Regression

What is the temperature going to be tomorrow?



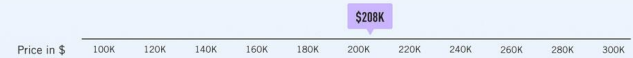
## Classification

Will it be Cold or Hot tomorrow?



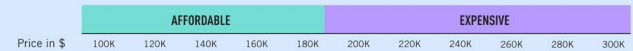
## Regression

What will house prices be like in my town next year?

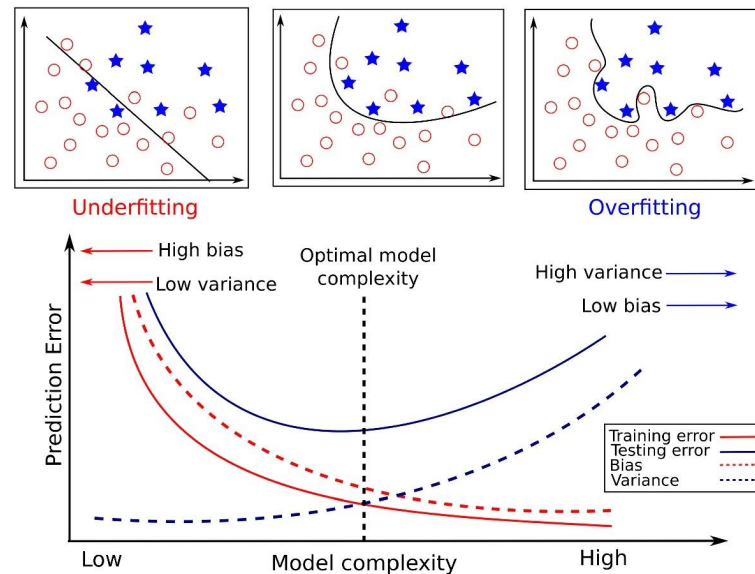
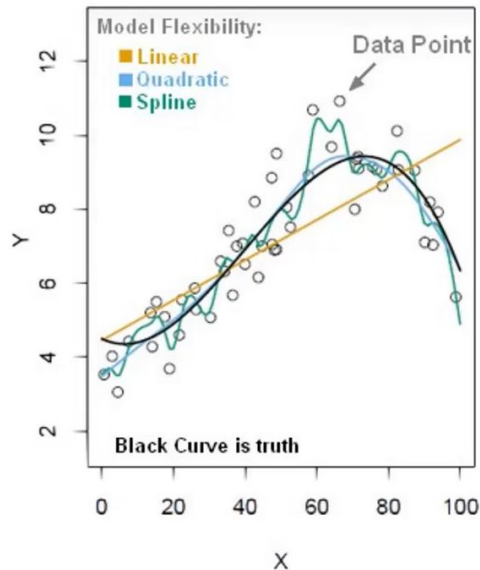


## Classification

Will houses be affordable in my town next year?



# The Bias-Variance Trade-off



$$E[(y - \hat{f}(x))^2] = \text{Bias}[\hat{f}(x)]^2 + \text{Var}[\hat{f}(x)] + \sigma^2$$

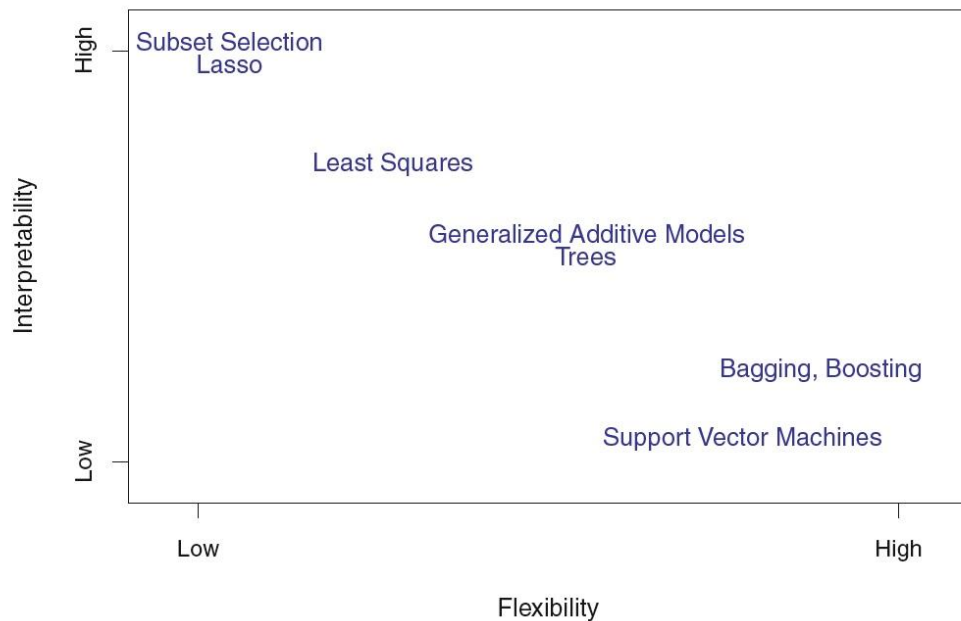
Where:

$$\text{Bias}[\hat{f}(x)] = E[\hat{f}(x) - f(x)]$$

and

$$\text{Var}[\hat{f}(x)] = E[\hat{f}(x)^2] - E[\hat{f}(x)]^2$$

# Flexibility vs Interpretability



**FIGURE 2.7.** *A representation of the tradeoff between flexibility and interpretability, using different statistical learning methods. In general, as the flexibility of a method increases, its interpretability decreases.*