# **STAT 177, CLASS 0**

**Richard Waterman** 

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# WELCOME TO THE COURSE

#### **OBJECTIVES**

- Learn how to code Python.
- Focus on applications for data science.
- Familiarize with popular data science libraries.
- Learn about the statistical underpinnings of the analyses.
- Have some fun!

## **SYLLABUS REVIEW**

- You read it.
- I'll review it.
- Resources (including Canvas and Piazza).
- Deliverables.
- Schedule.
- Grading.

# **PYTHON AND "DATA SCIENCE"**

#### WHAT IS DATA SCIENCE ANYWAY?

- A Data Scientist is usually involved in various parts of the analytics pipeline:
  - Data capture.
  - Data maintenance.
  - Data exploration and analysis.
  - Reporting and presentation.
- What's the difference between a data scientist and a statistician?

### TOP TEN REASONS FOR LIKING PYTHON

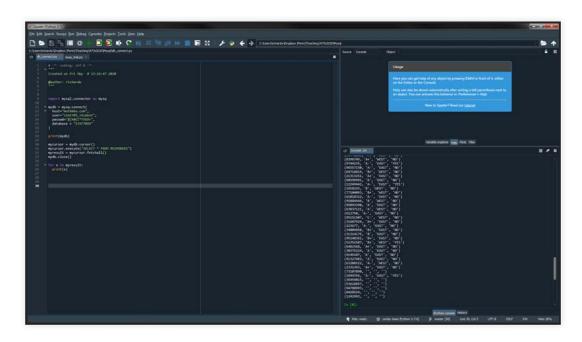
- 1. The most popular language for data science (but R is great too).
- 2. In demand/preferred skill.
- 3. All purpose software engineering framework (good at many different things).
- 4. Great for scientific computing (e.g. machine learning).
- 5. Scripting (formalizing a set of repetitive/boring tasks).

# TOP TEN REASONS FOR LIKING PYTHON (CTD.)

- 6. Open source.
- 7. Extensible (add-on packages for specific activities, e.g visualization).
- 8. Large support community.
- 9. Generous learning curve.
- 10. At least 10 more reasons!

# BY THE END OF THE COURSE ...

# **CODE SOME PYTHON**



Python coding

# IN 5 WEEKS, THE CODE BELOW SHOULD MAKE SENSE!

```
# Recode the levels of a categorical variable
print(carTable['Transmission'].value_counts())
recodes = {'A':'Automatic','AS':'Automatic','AV':'Automatic', 'M':'Manual'}
carTable['Transmission'] = carTable['Transmission'].map(recodes)
print(carTable.Transmission.value_counts())
```

# RETRIEVE DATA (FROM VARIOUS SOURCES)

```
import pandas as pd
import os
from sklearn import tree

# This is where the data lives:
os.chdir('C:\\Users\\richardw\\Dropbox (Penn)\\Teaching\\477s2020\\DataSets')
print(os.getcwd())

carTable = pd.read_csv("Car08_just_499.csv")
print(carTable)
```

```
C:\Users\richardw\Dropbox (Penn)\Teaching\477s2020\DataSets
    Make/Model MPG City MPG Hwy Weight(lb) Seating
                                                      Horsepower HP/Pound \
0
      Acura RL
                               24
                                         4014
                                                              290 0.072247
      Acura TL
                      17
                               26
                                         3674
                                                              286 0.077844
     Acura TL
                      18
                                         3559
                                                              286 0.080360
     Acura TSX
                      20
                               2.8
                                         3345
                                                              205 0.061285
                     19
                               28
                                         3257
                                                              205 0.062941
     Acura TSX
    Volvo V50
494
                      20
                               28
                                         3321
                                                              168 0.050587
495
     Volvo V70
                               24
                                       3527
                                                              235 0.066629
    Volvo XC90
496
                      14
                               20
                                       4356
                                                              235 0.053949
                               22
497
    Volvo XC70
                      15
                                         4092
                                                              235 0.057429
    Volvo XC90
498
                      13
                               19
                                         4826
                                                              311 0.064443
     Displacement Cylinders Origin Transmission
                                                   EPA Class Length Fuel \
```

#### RETRIEVE FROM A DATA BASE

```
import mysql.connector as mysq

mydb = mysq.connect(
  host="mathmba.com",
  user="stat705_student",
  passwd="$[h0CC*TtKO~",
  database = "STAT705X"
)

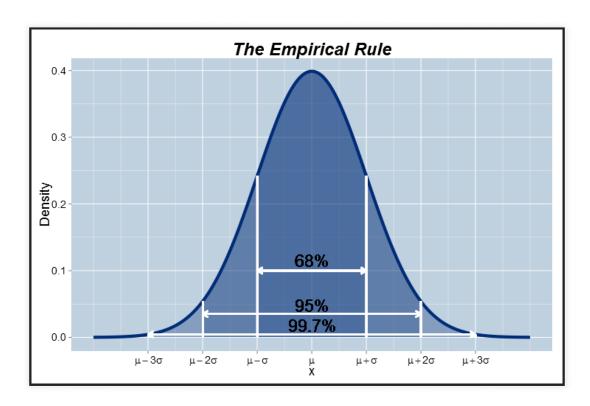
mycursor = mydb.cursor()
mycursor.execute("SELECT * FROM RESPONSES")
myresult = mycursor.fetchall()
mydb.close()
```

```
(98616981, 'B', 'EAST', 'YES')
(31004395, 'A', 'EAST', 'NO')
(42889147, 'A-', 'EAST', 'NO')
(3215038, 'A', 'EAST', 'NO')
(21698690, 'B+', 'EAST', 'NO')
(22549780, 'A-', 'EAST', 'NO')
(26222822, 'A-', 'WEST', 'NO')
(23218140, 'B', 'EAST', 'NO')
(98114241, 'A+', 'EAST', 'NO')
(55528599, 'A-', 'EAST', 'NO')
(42039091, 'B+', 'EAST', 'NO')
(91812638, 'A+', 'EAST', 'NO')
```

```
(90446266, 'A+', 'WEST', 'NO')
(8396749, 'A+', 'WEST', 'NO')
```

## **LEARN SOME CORE STATS IDEAS**

#### Making inferences:



The Empirical Rule

#### **CLEAN AND WRANGLE DATA**

```
# Recode the levels of a categorical variable
print(carTable['Transmission'].value_counts())
recodes = {'A':'Automatic', 'AS':'Automatic', 'AV':'Automatic', 'M':'Manual'}
carTable['Transmission'] = carTable['Transmission'].map(recodes)
print(carTable.Transmission.value_counts())
```

```
A 217
AS 129
M 122
AV 31
Name: Transmission, dtype: int64
Automatic 377
Manual 122
Name: Transmission, dtype: int64
```

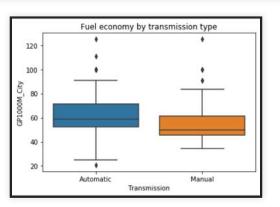
4.

#### **GRAPHICAL EXPLORATION OF DATA**

```
import matplotlib.pyplot as plt
import seaborn as sns

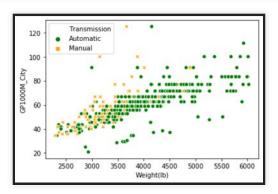
sns.boxplot(x = 'Transmission', y = 'GP1000M_City', data = carTable)
plt.title('Fuel economy by transmission type')
```

Text(0.5, 1.0, 'Fuel economy by transmission type')



## **BIVARIATE RELATIONSHIPS**

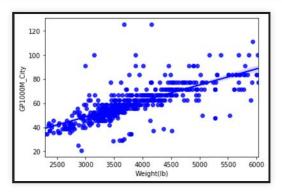
<matplotlib.axes. subplots.AxesSubplot at 0xc06cd88>



## **CREATE PREDICTIVE MODELS**

# A scatterplot with the linear regression line and confidence bands
sns.regplot(carTable['Weight(lb)'], carTable['GP1000M City'], color ='blue')

<matplotlib.axes.\_subplots.AxesSubplot at 0xc116988>



### **FIT A REGRESSION TREE**

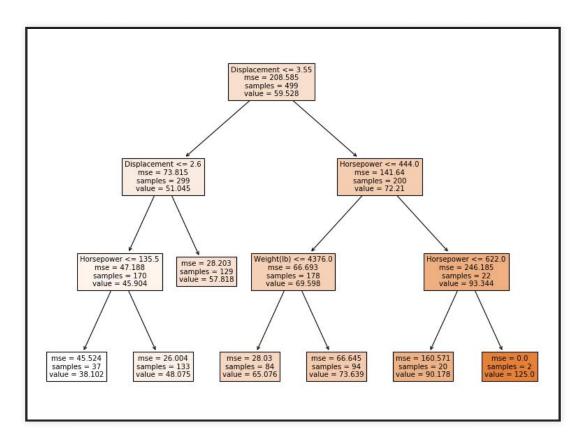
```
from sklearn.tree import DecisionTreeClassifier, plot_tree

X = carTable[['Weight(lb)', 'Displacement', 'Horsepower']]
y = carTable['GP1000M_City']
regtree = tree.DecisionTreeRegressor(max_leaf_nodes = 7)
regtree = regtree.fit(X, y,)
```

## **VISUALIZE MODELS**

```
os.chdir('C:\\Users\\richardw\\Dropbox (Penn)\\Teaching\\477s2020\\Notes')
fig = plt.figure(num=None, figsize=(12, 9), dpi=80, facecolor='w', edgecolor='k')
plot_tree(regtree, filled=True, feature_names=X.columns)
plt.savefig('images/tree_01.png', bbox_inches='tight')
plt.close(fig)
```

## **VISUALIZE MODELS**



Decision tree

# **CLASS SUMMARY**

### **SUMMARY**

- Objectives.
- Syllabus review.
- Deliverables.
- Resources.
- Reasons why Python is so popular.
- Got a sense of the types of data activities we will learn to do.

# **NEXT TIME**

## **NEXT TIME**

- Opening the Spyder IDE
  - Introduction to interactive Python
  - Getting help
- Types of data
- Storing data in variables
- Data structures