MSAS Tutorial Sequence -Module Three-

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LET'S **GET** READY TO WORK



Goals of Today

- 1. Finish up the basics of Python
 - Booleans
 - 2. If/Else
 - 3. Lists
 - 4. Loops
 - 5. Dictionaries
- 2. Understand Pandas
- 3. Getting some I/O with dataframes
 - 1. Reading from Excel
 - 2. Webscraping
 - 3. Basic dataframe operations



More Python

Booleans

- Booleans represent a "True" or "False" variable (can only hold these two values)
- Often used for comparison purposes
- Declared just like any other variable
 - o is_starter = True
 - temp = False
- Can also be used in inequalities and expressions
 - More on this in a bit

Operators

- The only three operators you should need to know for python are the following:
 - "and"
 - "or"
 - "not"
 - All three operators are used in their plaintext form (no need for "&&" or "| " or "!")

• Truth tables:

Α	Operand	В	Result
True	and	True	True
True	and	False	False
False	and	False	False
True	or	True	True
True	or	False	True
False	or	False	False

Conditions

- Recap: Python supports the usual logical conditions from mathematics:
 - ∘ Equals: a == b
 - ∘ Not Equals: a != b
 - Less than: a < b
 - Less than or equal to: a <= b</p>
 - Greater than: a > b
 - Greater than or equal to: a >= b
- "If" statements allow us to check "if" a condition is true or false
 - If the statement evaluates to True, we will go into to that "if" block and run what is inside it

```
In [1]: if 2 < 3:
    print("hello")
hello</pre>
```

```
In [2]: brady_touchdowns = 25
    rivers_touchdowns = 20

if brady_touchdowns > rivers_touchdowns:
    print("Brady is better")

Brady is better
```

```
In [4]:
        home = True
        sun = True
        if home and sun:
            print("We should win!")
        We should win!
        home = False
In [5]:
        sun = True
        if home or sun:
            print("We have a chance")
        We have a chance
        road = False
In [6]:
        if not road:
            print("We are playing at home")
        We are playing at home
```

Else statements

- Must be directly paired with an "if" statement
- •The "else" keyword catches anything that isn't caught by the if statement
- •If a condition is not hit, the else statement will run.
 Otherwise it will not.
 - Yes, an if/else clause is literally an if/else clause

```
In [1]: lebron = 27.4
        harden = 36.1
        Harden averaged more ppg than lebron
In [2]: # Don't do this
        if lebron > harden:
             print("Lebron averaged more ppg than harden")
        if harden > lebron:
             print("Harden averaged more ppg than lebron")
        Harden averaged more ppg than lebron
In [3]: # Do this
        if lebron > harden:
             print("Lebron averaged more ppg than harden")
         else:
            print("Harden averaged more ppg than lebron")
        Harden averaged more ppg than lebron
```

```
if goals_scored == 0:
    print("shutout")
elif goals_scored == 1:
    print("scored")
elif goals_scored == 3:
    print("Hat Trick!")
else:
    print("multiple scored")
```

```
In [5]: goals_scored = 2

if goals_scored == 0:
    print("shutout")
elif goals_scored == 1:
    print("scored")
elif goals_scored == 3:
    print("Hat Trick!")
else:
    print("multiple scored")
```

If/Elif/Else

- •"Elif" clauses allow for multiple if statements to be checked (short for "else if")
- First, the "if" statement is checked
- Next, each "elif" clause will be checked until one is hit
- •Finally, if none of the "elif" clauses are hit, the "else" clause is run

INDENTATION MATTERS

```
In [1]: a = 10
b = 5

if a > b:
print("here")

File "<ipython-input-1-c0905ae9b858>", line 5
    print("here")

IndentationError: expected an indented block
```

Lists

- Lists are a collection of items that is ordered and can be changed
- Lists are indicated by square brackets: []
- We can declare a list in three different ways:
 - ∘ An empty list □ my list = []
 - Fill it with elements \(\sum \text{my_list_2} = ["Ravens", "Saints", "Patriots"]
 - $\,^\circ\,$ This is a list of strings, but we can have a list of any datatype
 - You can mix and match your datatypes within the list, but this is considered bad practice
 - ∘ Set it equal to another list □ my_list_3 = my_list_2
- You can get the number of elements in the list by calling the "len" function
 - o my_list = [1, 2, 3, 4]
 - ∘ print(len(my_list)) □ 4
- Clear the list by calling the "clear function"
 - my_list.clear()
 - ∘ print(my list) □ ???

Accessing elements of a list

- You can access list items by referring to the index number
- Lists in Python are "0-index based"
- Example:
 - stats = ["pts", "reb", "ast"]
 - print(stats[0]) "pts"
 - print(stats[1]) "reb"
 - Print(stats[2]) "ast"
 - Print(stats[3]) Error! Why?
 - Print(stats[-1]) ???
- You can access a range of indexes by specifying where to start and where to end (called slicing)
 - Use the following format: list[start:end]
 - qbs = ["Brady", "Rivers", "Jackson", "Stafford", "Cousins"]
 - slice = qbs[2:4]
 - print(slice) ["Jackson", "Stafford", "Cousins"]

Updating a list element

 To update a list element, simply index into the list and reassign its value

```
In [9]: points = [10, 7, 13, 21, 15]
    print(points)
    print(points[3])
    points[3] = 19
    print(points)
    print(points[3])

[10, 7, 13, 21, 15]
    21
    [10, 7, 13, 19, 15]
    19
```

Adding to a list

```
In [1]: ppg = [27.4, 31.5, 21.5, 18.9, 29.7]
In [2]: # append to the end of a list
        ppg.append(15.7)
        ppg
Out[2]: [27.4, 31.5, 21.5, 18.9, 29.7, 15.7]
In [3]: # add another list
        more ppg = [11.5, 20.1, 16.7]
        ppg += more_ppg
        ppg
Out[3]: [27.4, 31.5, 21.5, 18.9, 29.7, 15.7, 11.5, 20.1, 16.7]
In [4]: # insert at a specific position
        ppg.insert(3, 10.1)
        ppg
Out[4]: [27.4, 31.5, 21.5, 10.1, 18.9, 29.7, 15.7, 11.5, 20.1, 16.7]
```

Range-based for loops (one way to loop)

- A "for" loop is used for iterating over a sequence
 - For our purposes, this will usually be a list
- We will only cover range based loops in this module
 - This is a brief sneak peak... for more information
 - Other loops aren't used traditionally in data science unless you are exploring a niche
 - Don't need to index into array
- Use cases:
 - When you want to view every element of a list in sequential order
 - Save time from indexing into specific elements
- Syntax: for element in list: do something

Basic loop examples

```
In [6]: simple_list = [0, 3, 2, 5]
                      for number in simple list:
                          print(number)
In [5]: AFC_West = ["Broncos", "Chargers", "Chiefs", "Raiders"]
        for team in AFC_West:
            print(team)
        Broncos
        Chargers
        Chiefs
        Raiders
```

Basic loop applications

```
In [7]: rebounds = [1, 4, 3, 2, 6, 1, 3]
    sum = 0

for game in rebounds:
        sum = sum + game

    print(sum / len(rebounds))
# What does this value represent?

2.857142857142857
```

```
In [18]: rebounds = [1, 4, 3, 2, 6, 1, 3]
    good_games = []

for game in rebounds:
    if game <= 3:
        bad_games.append(game)
    else:
        good_games.append(game)

print(bad_games)
print(good_games)

[1, 3, 2, 1, 3]
[4, 6]</pre>
```

Dictionaries

- Arguably one of the most important data structures in all of computer science
- •A dictionary is a collection of items which is unordered and indexed.
- •In Python, dictionaries are written with curly brackets and have keys and values
 - Keys are how you access values within the data structure
 - In a list, you index into the elements by numerical index
 - In a dictionary, you index into values with a key!
 - Every key is unique, but keys can have the same value
- •Why are dictionaries useful?

Example Dictionaries

```
players = {
    "LeBron James": "PF",
    "Anthony Davis": "C",
    "Steph Curry" : "PG",
    "Luka Doncic" : "SF",
    "Trae Young" : "PG",
    "Pau Gasol" : "C"
}
```

```
team_names = {
    "DET" : "Lions",
    "GB" : "Packers",
    "CHI" : "Bears",
    "MIN" : "Vikings"
}
```

```
In [1]: players = {
             "LeBron James": "PF",
             "Anthony Davis": "C",
             "Steph Curry": "PG",
             "Luka Doncic" : "SF",
             "Trae Young" : "PG",
             "Pau Gasol" : "C"
In [2]: players["LeBron James"]
Out[2]: 'PF'
In [3]: players["LeBron James"] = "C"
        players
Out[3]: {'LeBron James': 'C',
         'Anthony Davis': 'C',
          'Steph Curry': 'PG',
          'Luka Doncic': 'SF',
          'Trae Young': 'PG',
          'Pau Gasol': 'C'}
In [4]: players["Kevin Love"]
                                                   Traceback (most recent call last)
        KevError
        <ipython-input-4-b16d7fa527bf> in <module>
        ----> 1 players["Kevin Love"]
        KeyError: 'Kevin Love'
```

Accessing elements in a dictionary

```
In [14]: mlb_teams = {
    "NYY" : "Yankees",
    "BAL" : "Orioles",
    "BOS" : "Red Sox",
    "TB" : "Rays"
}

In [15]: if "TB" in mlb_teams:
    print("The value of TB is: " + mlb_teams["TB"])

The value of TB is: Rays

In [16]: if "TOR" not in mlb_teams:
    print("Forgot that one")
    mlb_teams["TOR"] = "Blue Jays"
    print(mlb_teams)

Forgot that one
{'NYY': 'Yankees', 'BAL': 'Orioles', 'BOS': 'Red Sox', 'TB': 'Rays', 'TOR': 'Blue Jays'}
```

Checking if an element is in a dictionary (try it on lists too!)

Questions about anything we talked about?

IF NOT, TAKE A COUPLE MINUTES TO WORK WITH THESE CONCEPTS







Pandas

Welcome to REAL Data Science

Libraries in Python

- A python library is a collection of functions that allows you to perform specific actions with your code
- More often than not, a library is designed for a specific feature. There are libraries to...
 - Move and modify files on your computer
 - Create graphical interfaces for your code
 - Webscrape (more on that in the next module)
 - Perform data science operations
 - And much more!
- Libraries can be imported with the following syntax:
 - import library (imports entire library)
 - import library as lib (imports entire library as an alias, usually used to not clutter up code)
 - from library, import function a, b (imports functions a and b from the library, saves memory)
- To access functions of your imported library, you will need to use the dot operator
 - We will see this throughout the module sequence

What is Pandas?

•From their documentation:

"pandas is an open source, BSD-licensed library providing high-performance, easy-to-use data structures and data analysis tools for the Python programming language."

- Very powerful data science library
- •The basis of the rest of this module, as well as the next three!
- Import pandas at the top of your jupyter notebok
 import pandas as pd

Series vs DataFrames

- •The primary two components of pandas are Series and DataFrames
- •Series is a column in your table, and a DataFrame is a multidimensional table comprised of a collection of Series

Series Series DataFrame apples oranges apples oranges 0 0 0 3 0 1 2 1 3 0 2 0 3 0 2 7 2 0 7 3 1 2 3 1

Reading in Files

import url lib.request urllib.request.urlretrieve('https://github.com/gfolbe318/MSAS_tutorials/raw/master/3_Module/nfl_stats.xlsx', 'nfl_stats.xlsx')

```
In [1]: import pandas as pd
In [5]: # Change the filename if you aren't in the same directory
    file_name = "nfl_stats.xlsx"
    # Read in the file from excel
    df = pd.read_excel(file_name)
```

Printing a DataFrame

```
In [6]: # This makes things ugly
        print(df)
                                                                         Cmp%
                              Player
                                       Tm Age
                                                       G GS
                                                              Cmp Att
                                                                              Yds \
              Rk
                   Ben Roethlisberger
                                     PIT
                                                      16 16
                                                              452
                                                                   675
                                                                         67.0 5129
                                            36
                                                                        67.3 4593
                        Andrew Luck* IND
                                                      16 16
                                                              430
                           Matt Ryan ATL
                                                                        69.4 4924
                                                      16
                                                          16
                                                                   608
                        Kirk Cousins MIN
                                                         16
                                                              425
                                                                   606
                                                                         70.1
                                                                              4298
                      Aaron Rodgers*
                                                         16
                                                              372 597
                                                                         62.3 4442
                         Case Keenum DEN
                                                          16
                                                                         62.3
                                                              365
                                                                   586
                                                                              3890
                   Patrick Mahomes*+ KAN
                                                                   580
                                                                              5097
                                                          16
                                                              383
                                                                         66.0
                         Eli Manning NYG
                                                         16
                                                                   576
                                                                         66.0 4299
                                            37
                                                      16
                                                              380
              9
                          Tom Brady*
                                     NWE
                                                         16
                                                              375
                                                                  570
                                                                         65.8
                                                                              4355
                         Jared Goff*
              10
                                      LAR
                                                      16
                                                          16
                                                                   561
                                                                              4688
                                                              364
                                                                         64.9
        10
                    Matthew Stafford DET
                                                         16
                                                                   555
             11
                                                              367
                                                                         66.1
                                                                              3777
                          Derek Carr OAK
        11
              12
                                           27
                                                         16
                                                              381
                                                                   553
                                                                         68.9
                                                                              4049
                       Dak Prescott*
        12
              13
                                     DAL
                                                          16
                                                              356
                                                                   526
                                                                         67.7
                                                                              3885
                      Philip Rivers*
                                                                   508
        13
              14
                                      LAC
                                                      16
                                                         16
                                                              347
                                                                        68.3
                                                                              4308
                     Deshaun Watson*
        14
             15
                                                      16 16
                                                              345
                                                                   505
                                                                         68.3 4165
        15
             16
                         Drew Brees*
                                                      15 15
                                                              364
                                                                   489
                                                                        74.4 3992
                      Baker Mayfield
             17
                                                                              3725
        16
                                                      14 13
                                                             310
                                                                   486
                                                                         63.8
        17
              18
                          Cam Newton
                                                      14 14 320 471
                                                                         67.9 3395
                  Mitchell Truhisky* CHT
                                                  OR 1/1 1/1 289 /3/1
                                                                         66 6 3223
```

Displaying a DataFrame

In [7]: # This makes things pretty df Out[7]: Rk Player G GS Cmp Att Cmp% Yds TD Int Y/A Y/C Rate QBR Sk Ben Roethlisberger 36 16 16 452 675 67.0 5129 34 16 7.6 11.3 320.6 71.0 24 Andrew Luck* IND QB 16 430 639 4593 39 15 7.2 10.7 287.1 69.4 18 Matt Ryan ATL 33 QB 16 422 608 8.1 11.7 307.8 68.5 42 16 Kirk Cousins MIN 425 58.2 40 30 606 7.1 10.1 268.6 QB 16 Aaron Rodgers* 7.4 11.9 277.6 **GNB** 372 597 Case Keenum DEN 30 365 586 15 6.6 10.7 243.1 45.5 34 Patrick Mahomes*+ KAN 23 383 580 8.8 13.3 318.6 80.4 26 QB 16 16 50 12 113.8 380 Eli Manning NYG 576 21 11 7.5 11.3 268.7 Tom Brady* NWE 16 375 7.6 11.6 272.2 66.6 21 QB 16 570 4355 29 11 10 Jared Goff* LAR 8.4 12.9 293.0 QB 16 561 32 12 101.1 63.1 33 11 Matthew Stafford 10 DET 555 66.1 3777 21 11 6.8 10.3 236.1 89.9 48.4 40 12 Derek Carr OAK QB 16 16 19 10 7.3 10.6 253.1 46.9 51 11 381 553 68.9 4049

Displaying the head of a DataFrame

Out[8]:

	Rk	Player	Tm	Age	Pos	G	GS	Cmp	Att	Cmp%	Yds	TD	Int	Y/A	Y/C	Y/G	Rate	QBR	Sk
0	1	Ben Roethlisberger	PIT	36	QB	16	16	452	675	67.0	5129	34	16	7.6	11.3	320.6	96.5	71.0	24
1	2	Andrew Luck*	IND	29	QB	16	16	430	639	67.3	4593	39	15	7.2	10.7	287.1	98.7	69.4	18
2	3	Matt Ryan	ATL	33	QB	16	16	422	608	69.4	4924	35	7	8.1	11.7	307.8	108.1	68.5	42
3	4	Kirk Cousins	MIN	30	QB	16	16	425	606	70.1	4298	30	10	7.1	10.1	268.6	99.7	58.2	40
4	5	Aaron Rodgers*	GNB	35	QB	16	16	372	597	62.3	4442	25	2	7.4	11.9	277.6	97.6	54.4	49

Display a Specific Number of Rows

```
In [10]: # Print out n number of rows
    num_rows = 10
    df.head(num_rows)
```

Out[10]:

	Rk	Player	Tm	Age	Pos	G	GS	Cmp	Att	Cmp%	Yds	TD	Int	Y/A	Y/C	Y/G	Rate	QBR	Sk
0	1	Ben Roethlisberger	PIT	36	QB	16	16	452	675	67.0	5129	34	16	7.6	11.3	320.6	96.5	71.0	24
1	2	Andrew Luck*	IND	29	QB	16	16	430	639	67.3	4593	39	15	7.2	10.7	287.1	98.7	69.4	18
2	3	Matt Ryan	ATL	33	QB	16	16	422	608	69.4	4924	35	7	8.1	11.7	307.8	108.1	68.5	42
3	4	Kirk Cousins	MIN	30	QB	16	16	425	606	70.1	4298	30	10	7.1	10.1	268.6	99.7	58.2	40
4	5	Aaron Rodgers*	GNB	35	QB	16	16	372	597	62.3	4442	25	2	7.4	11.9	277.6	97.6	54.4	49
5	6	Case Keenum	DEN	30	QB	16	16	365	586	62.3	3890	18	15	6.6	10.7	243.1	81.2	45.5	34
6	7	Patrick Mahomes*+	KAN	23	QB	16	16	383	580	66.0	5097	50	12	8.8	13.3	318.6	113.8	80.4	26
7	8	Eli Manning	NYG	37	QB	16	16	380	576	66.0	4299	21	11	7.5	11.3	268.7	92.4	48.7	47
8	9	Tom Brady*	NWE	41	QB	16	16	375	570	65.8	4355	29	11	7.6	11.6	272.2	97.7	66.6	21
9	10	Jared Goff*	LAR	24	QB	16	16	364	561	64.9	4688	32	12	8.4	12.9	293.0	101.1	63.1	33

```
In [11]: # Get dimensions of dataframe in the format of (rows, columns)
         df.shape
Out[11]: (106, 19)
In [12]: # Get the names of of the columns
         df.columns
Out[12]: Index(['Rk', 'Player', 'Tm', 'Age', 'Pos', 'G', 'GS', 'Cmp', 'Att', 'Cmp%',
                'Yds', 'TD', 'Int', 'Y/A', 'Y/C', 'Y/G', 'Rate', 'QBR', 'Sk'],
               dtype='object')
In [13]: # Get the tyeps of each column
         df.dtypes
Out[13]: Rk
                     int64
                    object
         Player
         Tm
                    object
                     int64
         Age
                    object
         Pos
                     int64
         G
                     int64
         GS
                     int64
         Cmp
         Att
                     int64
         Cmp%
                   float64
         Yds
                     int64
                     int64
         TD
         Int
                     int64
         Y/A
                   float64
         Y/C
                   float64
         Y/G
                   float64
         Rate
                   float64
                   float64
         OBR
                     int64
         Sk
         dtype: object
```

Get Information about the DataFrame

Extract Certain Columns

```
In [15]: # Pick out specifc columns that your want
    new_df = df[["Player"]]
    new_df.head()
```

Out[15]:

	Player
0	Ben Roethlisberger
1	Andrew Luck*
2	Matt Ryan
3	Kirk Cousins
4	Aaron Rodgers*

```
In [17]: # Get more than one column
    new_df = df[["Player", "Tm", "TD", "Yds"]]
    new_df.head()
```

Out[17]:

	Player	Tm	TD	Yds
0	Ben Roethlisberger	PIT	34	5129
1	Andrew Luck*	IND	39	4593
2	Matt Ryan	ATL	35	4924
3	Kirk Cousins	MIN	30	4298
4	Aaron Rodgers*	GNB	25	4442

Write to Excel

```
In [36]: # Write back to an excel sheet
# Be careful, you can only write to a sheet once
new_df.to_excel("important.xlsx")
```

Any Questions?