### Intro to pandas DataFrame iteration

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#### pandas recap

- See pandas overview in Intermediate Python
- Library used for data analysis
- Main data structure is the DataFrame
  - Tabular data with labeled rows and columns
  - Built on top of the NumPy array structure
- Chapter Objective:
  - Best practice for iterating over a pandas DataFrame

#### **Baseball stats**

```
import pandas as pd
baseball_df = pd.read_csv('baseball_stats.csv')
print(baseball_df.head())
```

	Tean	Leagu	е	Year	RS	RA	W	G	Playoffs
0	AR]	. N	L	2012	734	688	81	162	0
1	. ATL	. N	L	2012	700	600	94	162	1
2	BAL	. А	L	2012	712	705	93	162	1
3	B05	A	L	2012	734	806	69	162	0
4	CHO	N	L	2012	613	759	61	162	0

#### **Baseball stats**

Team
O ARI
1 ATL
2 BAL
3 BOS
4 CHC











#### **Baseball stats**

	Team L	eague	Year	RS	RA	W	G	Playoffs
0	ARI	NL	2012	734	688	81	162	0
1	ATL	NL	2012	700	600	94	162	1
2	BAL	AL	2012	712	705	93	162	1
3	BOS	AL	2012	734	806	69	162	0
4	CHC	NL	2012	613	759	61	162	0

#### Calculating win percentage

```
import numpy as np
def calc_win_perc(wins, games_played):
    win_perc = wins / games_played
    return np.round(win_perc,2)
win_perc = calc_win_perc(50, 100)
print(win_perc)
```

0.5

#### Adding win percentage to DataFrame

```
win_perc_list = []
for i in range(len(baseball_df)):
    row = baseball_df.iloc[i]
    wins = row['W']
    games_played = row['G']
    win_perc = calc_win_perc(wins, games_played)
    win_perc_list.append(win_perc)
baseball_df['WP'] = win_perc_list
```

#### Adding win percentage to DataFrame

```
print(baseball_df.head())
```

```
Team League
             Year
                    RS
                          RA
                             W
                                       Playoffs
                                                    WP
                                    G
ARI
             2012
                   734
                         688
                              81
                                                  0.50
                                  162
ATL
             2012
                   700
                         600
                              94
                                  162
                                                  0.58
BAL
             2012
                   712
                         705
                              93
                                  162
                                                  0.57
             2012
                   734
                         806
BOS
                              69
                                  162
                                                  0.43
CHC
             2012
                   613
                         759
                              61
                                  162
                                                  0.38
```

#### Iterating with .iloc

```
%%timeit
win_perc_list = []
for i in range(len(baseball_df)):
    row = baseball_df.iloc[i]
    wins = row['W']
    games_played = row['G']
    win_perc = calc_win_perc(wins, games_played)
    win_perc_list.append(win_perc)
baseball_df['WP'] = win_perc_list
```

```
183 ms \pm 1.73 ms per loop (mean \pm std. dev. of 7 runs, 10 loops each)
```



#### Iterating with .iterrows()

```
win_perc_list = []
for i,row in baseball_df.iterrows():
    wins = row['W']
    games_played = row['G']
    win_perc = calc_win_perc(wins, games_played)
    win_perc_list.append(win_perc)
baseball_df['WP'] = win_perc_list
```

#### Iterating with .iterrows()

```
%%timeit
win_perc_list = []
for i,row in baseball_df.iterrows():
    wins = row['W']
    games_played = row['G']
    win_perc = calc_win_perc(wins, games_played)
    win_perc_list.append(win_perc)
baseball_df['WP'] = win_perc_list
```

```
95.3 ms \pm 3.57 ms per loop (mean \pm std. dev. of 7 runs, 10 loops each)
```



# Practice DataFrame iterating with .iterrows()

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# Another iterator method: .itertuples()

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#### Team wins data

```
print(team_wins_df)
```

```
Year
     Team
0
      ARI
           2012
                  81
      ATL
           2012
                  94
      BAL
          2012
                  93
      BOS
          2012
                  69
          2012
      CHC
                  61
```

```
for row_tuple in team_wins_df.iterrows():
   print(row_tuple)
   print(type(row_tuple[1]))
(0, Team
          ARI
Year
       2012
         81
Name: 0, dtype: object)
<class 'pandas.core.series.Series'>
(1, Team ATL
Year
       2012
         94
Name: 1, dtype: object)
<class 'pandas.core.series.Series'>
```

#### Iterating with .itertuples()

```
for row_namedtuple in team_wins_df.itertuples():
    print(row_namedtuple)
Pandas(Index=0, Team='ARI', Year=2012, W=81)
Pandas(Index=1, Team='ATL', Year=2012, W=94)
print(row_namedtuple.Index)
print(row_namedtuple.Team)
ATL
```



#### Comparing methods

```
%%timeit
for row_tuple in team_wins_df.iterrows():
    print(row_tuple)
527 ms \pm 41.1 ms per loop (mean \pm std. dev. of 7 runs, 1 loop each)
%%timeit
for row_namedtuple in team_wins_df.itertuples():
    print(row_namedtuple)
```

```
7.48 ms \pm 243 \mus per loop (mean \pm std. dev. of 7 runs, 100 loops each)
```

```
for row_tuple in team_wins_df.iterrows():
    print(row_tuple[1]['Team'])
ARI
ATL
for row_namedtuple in team_wins_df.itertuples():
    print(row_namedtuple['Team'])
TypeError: tuple indices must be integers or slices, not str
for row_namedtuple in team_wins_df.itertuples():
    print(row_namedtuple.Team)
ARI
ATL
```



## Let's keep iterating!

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## pandas alternative to looping

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```
print(baseball_df.head())
```

```
Team League
                RS RA W
                             G
                                Playoffs
          Year
       NL 2012
ARI
                734
                    688
                        81 162
                                      0
ATL
       NL 2012
                700
                    600
                        94 162
       AL 2012 712 705 93 162
BAL
       AL 2012 734 806
                                      0
BOS
                        69 162
CHC
       NL 2012 613 759
                        61 162
                                      0
```

```
def calc_run_diff(runs_scored, runs_allowed):
    run_diff = runs_scored - runs_allowed
    return run_diff
```

#### Run differentials with a loop

```
run_diffs_iterrows = []

for i,row in baseball_df.iterrows():
    run_diff = calc_run_diff(row['RS'], row['RA'])
    run_diffs_iterrows.append(run_diff)

baseball_df['RD'] = run_diffs_iterrows
print(baseball_df)
```

```
Team League
              Year
                         RA
                                  G
                                    Playoffs
                    RS
                                              RD
           NL 2012
                    734
                        688
                             81 162
0
    ARI
                                              46
    ATL
           NL 2012
                    700 600
                             94 162
                                          1 100
    BAL AL 2012 712 705
                             93 162
```

#### pandas .apply() method

- Takes a function and applies it to a DataFrame
  - Must specify an axis to apply (0 for columns; 1 for rows)
- Can be used with anonymous functions (lambda functions)
- Example:

```
baseball_df.apply(
    lambda row: calc_run_diff(row['RS'], row['RA']),
    axis=1
)
```

#### Run differentials with .apply()

```
Team League
              Year
                     RS
                          RA
                                   G
                                      Playoffs
                                                RD
0
     ARI
            NL 2012
                    734
                              81 162
                         688
                                                46
            NL 2012 700
     ATL
                         600
                              94 162
                                               100
     BAL
            AL 2012 712 705
                              93 162
```

#### Comparing approaches

```
%%timeit
run_diffs_iterrows = []

for i,row in baseball_df.iterrows():
    run_diff = calc_run_diff(row['RS'], row['RA'])
    run_diffs_iterrows.append(run_diff)

baseball_df['RD'] = run_diffs_iterrows
```

```
86.8 ms \pm 3 ms per loop (mean \pm std. dev. of 7 runs, 10 loops each)
```

#### Comparing approaches

```
30.1 ms \pm 1.75 ms per loop (mean \pm std. dev. of 7 runs, 10 loops each)
```

# Let's practice using pandas .apply() method!

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# Optimal pandas iterating

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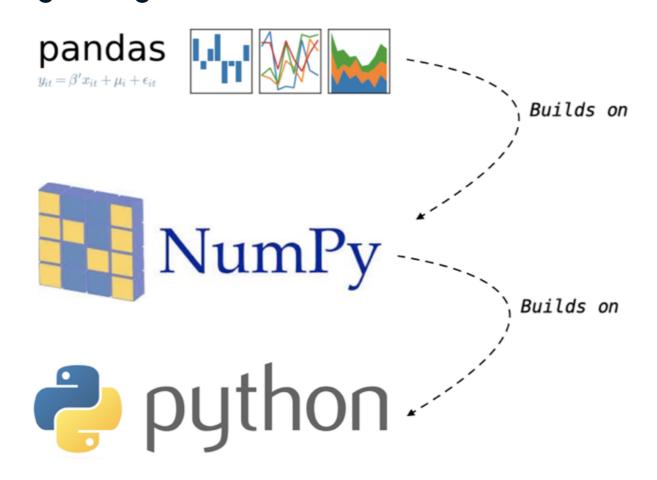
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#### pandas internals

- Eliminating loops applies to using pandas as well
- pandas is built on NumPy
  - Take advantage of NumPy array efficiencies



```
print(baseball_df)
 Team League Year
                   RS
                        RA
                           W
                                    Playoffs
  ARI
         NL 2012 734
                           81 162
                       688
  ATL
      NL 2012 700 600 94 162
      AL 2012 712 705 93 162
  BAL
wins_np = baseball_df['W'].values
print(type(wins_np))
<class 'numpy.ndarray'>
print(wins_np)
[ 81 94 93 ...]
```



#### Power of vectorization

• Broadcasting (vectorizing) is extremely efficient!

```
baseball_df['RS'].values - baseball_df['RA'].values
```

```
array([ 46, 100, 7, ..., 188, 110, -117])
```

#### Run differentials with arrays

```
run_diffs_np = baseball_df['RS'].values - baseball_df['RA'].values
baseball_df['RD'] = run_diffs_np
print(baseball_df)
```

```
Team League
                Year
                       RS
                            RA
                                      G
                                         Playoffs
                                                    RD
             NL
                2012
0
     ARI
                      734
                           688
                                 81 162
                                                    46
             NL 2012
                                                1 100
     ATL
                      700
                           600
                                 94 162
             AL 2012
     BAL
                      712
                           705
                                 93 162
3
                                                0 -72
             AL 2012 734
     BOS
                           806
                                 69 162
     CHC
             NL 2012
                      613
                          759
                                 61 162
                                                0 -146
4
```

#### Comparing approaches

```
%%timeit
run_diffs_np = baseball_df['RS'].values - baseball_df['RA'].values
baseball_df['RD'] = run_diffs_np
```

```
124 \mus \pm 1.47 \mus per loop (mean \pm std. dev. of 7 runs, 10000 loops each)
```

# Let's put our skills into practice!

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## Congratulations!

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#### What you have learned

- The definition of efficient and Pythonic code
- How to use Python's powerful built-in library
- The advantages of NumPy arrays
- Some handy magic commands to profile code
- How to deploy efficient solutions with zip(), itertools, collections, and set theory
- The cost of looping and how to eliminate loops
- Best practices for iterating with pandas DataFrames

### Well done!

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