



Filtering

Filter DataFrames for values that fit certain conditions.

Chapter Goals:

- Understand how to filter a DataFrame based on filter conditions
- Write code to filter a dataset of MLB statistics

A. Filter conditions

In the Data Manipulation section, we used relation operations on NumPy arrays to create *filter conditions*. These filter conditions returned boolean arrays, which represented the locations of the elements that pass the filter.

In pandas, we can also create filter conditions for DataFrames. Specifically, we can use relation operations on a DataFrame's column features, which will return a boolean Series representing the DataFrame rows that pass the filter.

The code below demonstrates how to use relation operations as filter conditions.

```
1  df = pd.DataFrame({
2    'playerID': ['bettsmo01', 'canoro01', 'cruzne02', 'ortizda01', 'cruzne02'],
3    'yearID': [2016, 2016, 2016, 2017],
4    'teamID': ['BOS', 'SEA', 'SEA', 'BOS', 'SEA'],
5    'HR': [31, 39, 43, 38, 39]})
6
7  print('{}\n'.format(df))
8
9  cruzne02 = df['playerID'] == 'cruzne02'
10  print('{}\n'.format(cruzne02))
11
12  hr40 = df['HR'] > 40
13  print('{}\n'.format(hr40))
14
15  notbos = df['teamID'] != 'BOS'
16  print('{}\n'.format(notbos))
```

```
X
                                                                  0.696s
Output
    HR
        playerID teamID yearID
    31 bettsmo01
                    BOS
                          2016
 1
   39
        canoro01
                   SEA
                          2016
   43
        cruzne02
                   SEA
                          2016
    38 ortizda01
                   BOS
                          2016
 3
    39
        cruzne02
                   SEA
                          2017
 0
     False
 1
     False
```

In the code above, we created filter conditions for df based on the columns labeled 'playerID', 'HR', and 'teamID'. The boolean Series outputs have True for the rows that pass the filter, and False for the rows that don't.

B. Filters from functions

Apart from relation operations, pandas provides various functions for creating specific filter conditions. For columns with string values, we can use str.startswith (https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.Series.str.startswith.html), str.endswith (https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.Series.str.endswith.html), and str.contains (https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.Series.str.contains.html) to filter for specific strings. These functions work the exact same as their namesakes from the Python standard library.

The code below shows various examples of string filter conditions. In the final example using str.contains, we prepend the ~ operation, which negates the filter condition. This means our final filter condition checked for player IDs that *do not* contain 'o'.

```
1
   df = pd.DataFrame({
      'playerID': ['bettsmo01', 'canoro01', 'cruzne02', 'ortizda01', 'cru ခြ
 2
      'yearID': [2016, 2016, 2016, 2016, 2017],
 3
      'teamID': ['BOS', 'SEA', 'SEA', 'BOS', 'SEA'],
 4
 5
      'HR': [31, 39, 43, 38, 39]})
 6
 7
    print('{}\n'.format(df))
 8
 9
    str_f1 = df['playerID'].str.startswith('c')
    print('{}\n'.format(str_f1))
10
11
12
   str_f2 = df['teamID'].str.endswith('S')
    print('{}\n'.format(str_f2))
13
14
15
   str_f3 = ~df['playerID'].str.contains('o')
16 print('{}\n'.format(str_f3))
                                                              \triangleright
  X
Output
                                                                      0.931s
    HR
         playerID teamID yearID
   31 bettsmo01
                     BOS
                            2016
 0
 1
   39
         canoro01
                     SEA
                            2016
 2
   43
         cruzne02
                     SEA
                            2016
   38 ortizda01
                     BOS
                            2016
 3
         cruzne02
    39
                     SEA
                            2017
      False
 0
 1
       True
```

We can also create filter conditions that check for values in a specific set, by using the <code>isin</code> (https://pandas.pydata.org/pandas-docs/stable/reference/api/pandas.Series.isin.html) function. The function only takes in one argument, which is a list of values that we want to filter for.

The code below demonstrates how to use the isin function for filter conditions.

```
'playerID': ['bettsmo01', 'canoro01', 'cruzne02', 'ortizda01', 'cruzne0
 2
       'yearID': [2016, 2016, 2016, 2016, 2017],
      'teamID': ['BOS', 'SEA', 'SEA', 'BOS', 'SEA'],
      'HR': [31, 39, 43, 38, 39]})
 5
 6
 7
    print('{}\n'.format(df))
 8
 9
    isin_f1 = df['playerID'].isin(['cruzne02',
                                    'ortizda01'])
10
    print('{}\n'.format(isin_f1))
11
12
    isin_f2 = df['yearID'].isin([2015, 2017])
13
    print('{}\n'.format(isin_f2))
14
                                                              X
                                                                       0.706s
Output
    HR
         playerID teamID yearID
    31 bettsmo01
                     BOS
                            2016
 1
    39
         canoro01
                     SEA
                            2016
                     SEA
 2
   43
         cruzne02
                            2016
    38 ortizda01
                     BOS
                            2016
    39
         cruzne02
                            2017
                     SEA
 0
      False
 1
      False
```

In pandas, when a Series or DataFrame has a missing value at a location, it is represented by NaN. The NaN value in pandas is equivalent to np.nan in NumPy.

Similar to Numpy, we cannot use a relation operation to create a filter condition for NaN values. Instead, we use the isna

```
(https://pandas.pydata.org/pandas-
docs/stable/reference/api/pandas.Series.isna.html) and notna
(https://pandas.pydata.org/pandas-
docs/stable/reference/api/pandas.Series.notna.html) functions.
```

```
1 df = pd.DataFrame({
```

```
'playerID': ['bettsmo01', 'canoro01', 'doejo01'],
 2
 3
       'yearID': [2016, 2016, 2017],
 4
       'teamID': ['BOS', 'SEA', np.nan],
 5
       'HR': [31, 39, 99]})
 6
 7
    print('{}\n'.format(df))
 8
    isna = df['teamID'].isna()
 9
10
    print('{}\n'.format(isna))
11
12
    notna = df['teamID'].notna()
    print('{}\n'.format(notna))
13
                                                                \triangleright
  X
Output
                                                                         0.703s
    HR
         playerID teamID yearID
    31 bettsmo01
                      BOS
                             2016
    39
         canoro01
                      SEA
                             2016
 2
    99
          doejo01
                      NaN
                             2017
 0
      False
 1
      False
       True
 Name: teamID, dtype: bool
```

The isna function returns True in the locations that contain NaN and False in the locations that don't, while the notna function does the opposite.

C. Feature filtering

It is really easy to filter a DataFrame's rows based on filter conditions. Similar to direct indexing of a DataFrame, we use square brackets. However, the inside of the square brackets will now contain a filter condition.

When applying filter conditions within square brackets, we retrieve the rows of the DataFrame that pass the filter condition (i.e. the rows for which the filter condition is True).

The code below shows how to filter using square brackets and filter conditions.

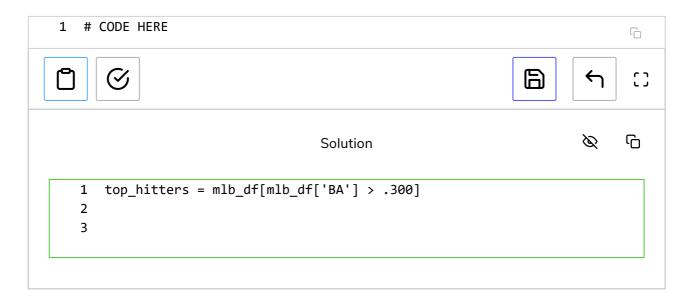
```
1 df = pd.DataFrame({
      'playerID': ['bettsmo01', 'canoro01', 'cruzne02', 'ortizda01', 'bettsmo01'],
 2
      'yearID': [2016, 2016, 2016, 2016, 2015],
      'teamID': ['BOS', 'SEA', 'SEA', 'BOS', 'BOS'],
      'HR': [31, 39, 43, 38, 18]})
 7
   print('{}\n'.format(df))
 8
 9 hr40_df = df[df['HR'] > 40]
10 print('{}\n'.format(hr40_df))
11
12 not_hr30_df = df[~(df['HR'] > 30)]
13 print('{}\n'.format(not_hr30_df))
14
15 str_df = df[df['teamID'].str.startswith('B')]
16 print('{}\n'.format(str_df))
                                                            \leftarrow
  X
                                                                    0.642s
Output
    HR
         playerID teamID yearID
 0 31 bettsmo01
                    BOS
                           2016
 1 39 canoro01
                    SEA
                           2016
 2 43 cruzne02
                    SEA
                           2016
 3 38 ortizda01
                    BOS
                           2016
 4 18 bettsmo01
                    BOS
                           2015
    HR playerID teamID yearID
 2 43 cruzne02
                   SEA
                          2016
```

Time to Code!

In this chapter's code exercises, we'll apply various filters to a predefined DataFrame, mlb_df, which contains MLB statistics.

We'll first filter mlb_df for the top MLB hitting seasons in history, which we define as having a batting average above .300.

Set top_hitters equal to mlb_df[] applied with mlb_df['BA'] > .300 as the filter condition.



Next we filter for the players whose player ID *does not* start with the letter **a**.

Set exclude_a equal to mlb_df[] applied with the negation of mlb_df['playerID'].str.startswith('a') as the filter condition.



We'll now retrieve the statistics for two specific players. Their player IDs are 'bondsba01' and 'troutmi01'.

Set two_ids equal to a list containing the two specified player IDs.

