# **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.

```
df = pd.DataFrame({
   'playerID': ['bettsmo01', 'canoro01', 'cruzne02', 'ortizda01', 'cruzne02'],
   'yearID': [2016, 2016, 2016, 2016, 2017],
   'teamID': ['BOS', 'SEA', 'SEA', 'BOS', 'SEA'],
   'HR': [31, 39, 43, 38, 39]})
print('{}\n'.format(df))
cruzne02 = df['playerID'] == 'cruzne02'
print('{}\n'.format(cruzne02))
hr40 = df['HR'] > 40
print('{}\n'.format(hr40))
notbos = df['teamID'] != 'BOS'
print('{}\n'.format(notbos))
```







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 <a href="mailto:str.endswith">str.endswith</a>, and <a href="mailto:str.contains">str.contains</a> 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 <a href="str.contains">str.contains</a>, we prepend the <a href="operation">operation</a>, which negates the filter condition. This means our final filter condition checked for player IDs that *do not* contain <a href="o">o</a>.

```
df = pd.DataFrame({
   'playerID': ['bettsmo01', 'canoro01', 'cruzne02', 'ortizda01', 'cruzne02'],
   'yearID': [2016, 2016, 2016, 2016, 2017],
   'teamID': ['BOS', 'SEA', 'SEA', 'BOS', 'SEA'],
   'HR': [31, 39, 43, 38, 39]})

print('{}\n'.format(df))

str_f1 = df['playerID'].str.startswith('c')
print('{}\n'.format(str_f1))

str_f2 = df['teamID'].str.endswith('S')
print('{}\n'.format(str_f2))

str_f3 = ~df['playerID'].str.contains('o')
print('{}\n'.format(str_f3))
```

We can also create filter conditions that check for values in a specific set, by using the <code>isin</code> 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.

```
df = pd.DataFrame({
    'playerID': ['bettsmo01', 'canoro01', 'cruzne02', 'ortizda01', 'cruzne02'],
```

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 and notna functions.

```
df = pd.DataFrame({
    'playerID': ['bettsmo01', 'canoro01', 'doejo01'],
    'yearID': [2016, 2016, 2017],
    'teamID': ['BOS', 'SEA', np.nan],
    'HR': [31, 39, 99]})
print('{}\n'.format(df))
isna = df['teamID'].isna()
print('{}\n'.format(isna))
notna = df['teamID'].notna()
print('{}\n'.format(notna))
```

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.

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.

```
df = pd.DataFrame({
    'playerID': ['bettsmo01', 'canoro01', 'cruzne02', 'ortizda01', 'bettsmo01'],
    'yearID': [2016, 2016, 2016, 2015],
    'teamID': ['BOS', 'SEA', 'SEA', 'BOS'],
    'HR': [31, 39, 43, 38, 18]})

print('{}\n'.format(df))

hr40_df = df[df['HR'] > 40]
print('{}\n'.format(hr40_df))

not_hr30_df = df[~(df['HR'] > 30)]
print('{}\n'.format(not_hr30_df))

str_df = df[df['teamID'].str.startswith('B')]
print('{}\n'.format(str_df))
```

## 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['nlaverID'] str\_startswith('a') as the filter condition

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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.

Set two\_players equal to mlb\_df[] applied with mlb\_df['playerID'].isin(two\_ids) as the filter condition.

