This notebook is an exercise in the Pandas (https://www.kaggle.com/learn/pandas) course. You can reference the tutorial at this link (https://www.kaggle.com/residentmario/summary-functions-and-maps).

# Introduction

Now you are ready to get a deeper understanding of your data.

Run the following cell to load your data and some utility functions (including code to check your answers).

```
import pandas as pd
pd.set_option("display.max_rows", 5)
reviews = pd.read_csv("../input/wine-reviews/winemag-data-130k-v2.csv",
index_col=0)

from learntools.core import binder; binder.bind(globals())
from learntools.pandas.summary_functions_and_maps import *
print("Setup complete.")

reviews.head()
```

#### Setup complete.

#### Out[1]:

|   | country  | description   | designation                                 | points | price | province             | region 1                  | region 2             | taste         |
|---|----------|---|---|--------|-------|----------------------|---------------------------|----------------------|---------------|
| С | Italy    | Aromas include tropical fruit, broom, brimston                | Vulkà<br>Bianco                             | 87     | NaN   | Sicily &<br>Sardinia | Etna                      | NaN                  | Kerii<br>OʻKe |
| 1 | Portugal | This is ripe<br>and fruity,<br>a wine that<br>is<br>smooth    | Avidagos                                    | 87     | 15.0  | Douro                | NaN                       | NaN                  | Rog           |
| 2 | US       | Tart and<br>snappy,<br>the flavors<br>of lime<br>flesh and    | NaN   | 87     | 14.0  | Oregon               | Willamette<br>Valley      | Willamette<br>Valley | Paul<br>Greç  |
| 3 | US       | Pineapple<br>rind, lemon<br>pith and<br>orange<br>blossom     | Reserve<br>Late<br>Harvest                  | 87     | 13.0  | Michigan             | Lake<br>Michigan<br>Shore | NaN                  | Alex<br>Pear  |
| 4 | US       | Much like<br>the regular<br>bottling<br>from<br>2012,<br>this | Vintner's<br>Reserve<br>Wild Child<br>Block | 87     | 65.0  | Oregon               | Willamette<br>Valley      | Willamette<br>Valley | Paul<br>Greç  |

# **Exercises**

#### 1.

What is the median of the points column in the reviews DataFrame?

```
In [2]:
    median_points = reviews.points.median()

# Check your answer
q1.check()
```

Correct

```
In [3]:
    #q1.hint()
    #q1.solution()
```

## 2.

What countries are represented in the dataset? (Your answer should not include any duplicates.)

```
In [4]:
    countries = reviews.country.unique()

# Check your answer
    q2.check()
```

Correct

```
In [5]:
    #q2.hint()
    #q2.solution()
```

#### 3.

How often does each country appear in the dataset? Create a Series reviews\_per\_country mapping countries to the count of reviews of wines from that country.

```
In [6]:
    reviews_per_country = reviews.country.value_counts()

# Check your answer
q3.check()
```

Correct

```
In [7]:
    #q3.hint()
    #q3.solution()
```

## 4.

Create variable centered\_price containing a version of the price column with the mean price subtracted.

(Note: this 'centering' transformation is a common preprocessing step before applying various machine learning algorithms.)

```
In [8]:
    centered_price = reviews.price - reviews.price.mean()

# Check your answer
q4.check()
```

Correct

```
In [9]: #q4.hint() #q4.solution()
```

### 5.

I'm an economical wine buyer. Which wine is the "best bargain"? Create a variable bargain\_wine with the title of the wine with the highest points-to-price ratio in the dataset.

```
In [10]:
    bargain_wine = reviews.loc[ (reviews.points / reviews.price).idxmax(),
    'title']

# Check your answer
q5.check()
```

Correct

```
In [11]:
    #q5.hint()
    #q5.solution()
```

#### 6.

There are only so many words you can use when describing a bottle of wine. Is a wine more likely to be "tropical" or "fruity"? Create a Series descriptor\_counts counting how many times each of these two words appears in the description column in the dataset.

```
In [12]:
    n_trop = reviews.description.map(lambda desc: "tropical" in desc).sum()
    n_fruity = reviews.description.map(lambda desc: "fruity" in desc).sum()
    descriptor_counts = pd.Series([n_trop, n_fruity], index=['tropical', 'f
    ruity'])

# Check your answer
    q6.check()
```

Correct

```
In [13]:
    #q6.hint()
    #q6.solution()
```

#### 7.

We'd like to host these wine reviews on our website, but a rating system ranging from 80 to 100 points is too hard to understand - we'd like to translate them into simple star ratings. A score of 95 or higher counts as 3 stars, a score of at least 85 but less than 95 is 2 stars. Any other score is 1 star.

Also, the Canadian Vintners Association bought a lot of ads on the site, so any wines from Canada should automatically get 3 stars, regardless of points.

Create a series star\_ratings with the number of stars corresponding to each review in the dataset.

```
In [14]:
    def star_num(index):
        if index.country == 'Canada':
            return 3
        elif index.points >= 95:
            return 3
        elif index.points >= 85:
            return 2
        else:
            return 1

    star_ratings = reviews.apply(star_num, axis='columns')

# Check your answer
q7.check()
```

Correct

```
In [16]:
    #q7.hint()
    #q7.solution()
```

# Keep going

Continue to grouping and sorting (https://www.kaggle.com/residentmario/grouping-and-sorting).

Have questions or comments? Visit the Learn Discussion forum (https://www.kaggle.com/learn-forum/161299) to chat with other Learners.

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