# Homework 06



## A Before you start A



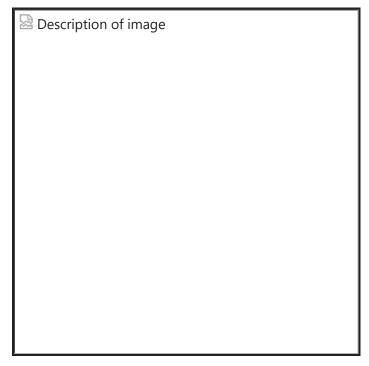
Duplicate this Jupyter Notebook in your week-07 folder (right-click -> Duplicate) and then add your last name to the beginning of it (ie. blevins-hw-06.ipynb - otherwise you risk having all your work overwritten when you try to sync your GitHub repository with your instructor's repository.

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### Overview

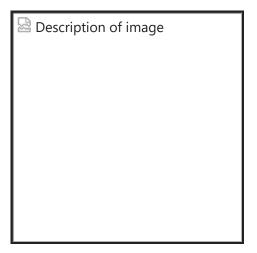
This homework assignment will help you learn how to use the Pandas library to explore tabular data by applying some of the concepts and lessons from Melanie Walsh, Pandas I to a new dataset.

This week we're going to use a spreadsheet of historical data transcribed by CU Denver history major Ryan Hanlon as part of his final project for the course Introduction to Digital Studies in Spring 2021. The data consists of a passenger list from a steamship that arrived in Boston on April 9, 1884 carrying several hundred immigrants.



The Steamship Grecian

The nine-page passenger list from the *Steamship Grecian* was submitted to authorities at the Port of Boston. This document was later scanned by the Church of Latter Day Saints and made available through its FamilySearch online archive.



Passenger list from the Steamship Grecian

Ryan then transcribed the data in Spring 2021 into a spreadsheet formatted as a CSV (comma separated value) file contained in this folder: boston-passenger-list-1884.csv.

For this homework, you will be following many of Melanie Walsh's steps in Pandas I and then adapting them to fit this new dataset.

 Import the Pandas library (use the alias pd ) and read in the CSV file, storing the contents of the file as a dataframe named passengers\_df. Add a second line of code to display the contents of the dataframe (truncated).

```
In [9]: #Your Code Here
import pandas as pd

passengers_df = pd.read_csv('boston-passenger-list-1884.csv', delimiter=",")
passengers_df
```

Out[9]:	first_name		last_name	last_name date		native_country	destination_city	destination_state	
	0	Jno	McNab	09 Apr 1884	21	Scotland	Suelpla	Canada	
	1	Thos	Campbell	09 Apr 1884	24	Scotland	Suelpla	Canada	
	2	Jas	Mitchell	09 Apr 1884	23	Scotland	Detroit	MI	
	3	Don	Cumming	09 Apr 1884	24	Scotland	Detroit	MI	
	4	Jno	McKinlay	09 Apr 1884	24	Scotland	Winnipeg	Canada	
	•••								
	510	Mich	Mulkern	09 Apr 1884	15	Ireland	Boston	MA	
	511	Math	Griffins	09 Apr 1884	2	Ireland	Pittsburg	PA	
	512	Т	McDermott	09 Apr 1884	35	United States	Boston	MA	
	513	Wm	Hewitt	09 Apr 1884	25	United States	Boston	MA	
	514	F	Doherty	09 Apr 1884	27	United States	Boston	MA	

515 rows × 9 columns

### 2. Display the **first 6 rows** of the dataframe.

```
In [11]: #Your Code Here
passengers_df.head(6)
```

Out[11]:		first_name	last_name	date	age	native_country	destination_city	destination_state	ОС
	0	Jno	McNab	09 Apr 1884	21	Scotland	Suelpla	Canada	
	1	Thos	Campbell	09 Apr 1884	24	Scotland	Suelpla	Canada	
	2	Jas	Mitchell	09 Apr 1884	23	Scotland	Detroit	МІ	
	3	Don	Cumming	09 Apr 1884	24	Scotland	Detroit	МІ	В
	4	Jno	McKinlay	09 Apr 1884	24	Scotland	Winnipeg	Canada	
	5	John	Wilson	09 Apr 1884	28	Scotland	Boston	MA	

### 3. Show a **random sample of 10 rows** from your dataframe.

In [13]: #Your Code Here
passengers\_df.sample(10)

Out[13]:	first_name		last_name	last_name date a		native_country	destination_city	destination_state
	297	Pat	Durstan	09 Apr 1884	17	Ireland	Providence	RI
	66	Eliz	Caldwell	09 Apr 1884	40	Ireland	Boston	MA
	343	Bgt	McDonough	09 Apr 1884	44	Ireland	Brooklyn	NY
	418	Thos	Mitchell	09 Apr 1884	34	Ireland	Boston	MA
	16	John	Little	09 Apr 1884	26	Scotland	Woodstock	VT
	302	John	Parsons	09 Apr 1884	30	Ireland	Chicago	IL
	111	Robt	Gilmore	09 Apr 1884	25	Ireland	Boston	MA
	68	Jas	Gallagher	09 Apr 1884	46	Ireland	Boston	MA
	177	Kate	Mellett	09 Apr 1884	18	Ireland	Taunton	MA
	452	Margt	Griffin	09 Apr 1884	17	Ireland	Pittsburg	PA

4. What are **two historical questions** about this list of passengers that you might be able to answer using Pandas?

#### Your answer here.

- 1. Pandas can help determine where most migrants migrated to and where most migrants emigrated from.
- 2. Another thing Pandas could help answer using this dataset is what these occupations were for these immigrants coming to the Americas.

# Analyzing the Data

5. Calculate "summary statistics" for the passenger data.

In [18]: #Your Code Here
passengers\_df.describe(include='all')

	passenger s_ur . uescr ibe(include= all )									
Out[18]:		first_name	last_name	date	age	native_country	destination_city	destinat		
	count	512	515	515	515.000000	515	515			
	unique	102	170	1	NaN	3	56			
	top	Mary	Doherty	09 Apr 1884	NaN	Ireland	Boston			
	freq	59	12	515	NaN	461	107			
	mean	NaN	NaN	NaN	21.440777	NaN	NaN			
	std	NaN	NaN	NaN	13.115392	NaN	NaN			
	min	NaN	NaN	NaN	0.000000	NaN	NaN			
	25%	NaN	NaN	NaN	11.000000	NaN	NaN			
	50%	NaN	NaN	NaN	20.000000	NaN	NaN			
	75%	NaN	NaN	NaN	28.000000	NaN	NaN			
	max	NaN	NaN	NaN	64.000000	NaN	NaN			
<pre>In [19]: passengers_df['last_name'].value_counts().head(1)  Out[19]: last_name</pre>										
In [20]:	passeng	ers_df[' <mark>occ</mark>	upation'].	value_	counts()					
Out[20]: occupation Laborer 189 Child 132 Domestic 116 Wife 64 Farmer 8 Boatman 3 Blacksmith 1 Miller 1 Clerk 1 Name: count, dtype: int64										
In [21]:	passeng	ers_df['age	'].max()							

```
Out[21]: 64
```

- 6. Looking at the summary statistics, answer the following questions:
- What is the most frequently occuring last name?
- How often does the most frequently occurring last name appear?
- How many different kinds of occupations are listed in the data?
- How old is the oldest passenger?

#### Your answers here:

Name: count, dtype: int64

- What is the most frequently occurring last name? The most common last name is Doherty.
- How often does the most frequently occuring last name appear? It appears 12 times.
- How many different kinds of occupations are listed in the data? There were nine types of labeled occupations on the dataset.
- How old is the oldest passenger? The oldest passenger was 64 years old.
- 7. Write code to answer: what was the **median** age of the passengers?

```
In [25]: #Your Code Here
passengers_df['age'].median()
```

Out[25]: 20.0

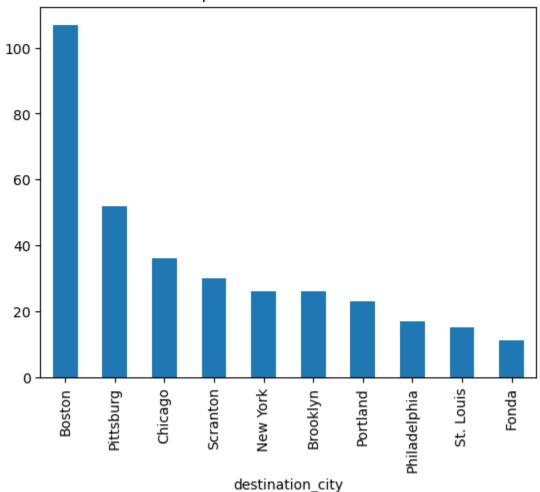
8. What were the **ten most frequent cities** that passengers were traveling to and how many of them were going to each of these cities?

```
In [27]: #Your Code Here
        passengers_df['destination_city'].value_counts().head(10)
Out[27]: destination_city
         Boston 107
         Pittsburg
                        52
         Chicago
                        36
                        30
         Scranton
         New York
                        26
         Brooklyn
                        26
                         23
         Portland
         Philadelphia
                       17
         St. Louis
                         15
         Fonda
                         11
```

Follow Walsh's example and adapt her code to make a bar chart of the top ten most frequent destination cities based on how many passengers were going to each of them.

```
In [29]: #Your Code Here
    passengers_df['destination_city'].value_counts()[:10].plot(kind='bar', title='Passe
Out[29]: <Axes: title={'center': 'Passengers:\nTop 10 Destination Cities'}, xlabel='destination_city'>
```

Passengers: Top 10 Destination Cities

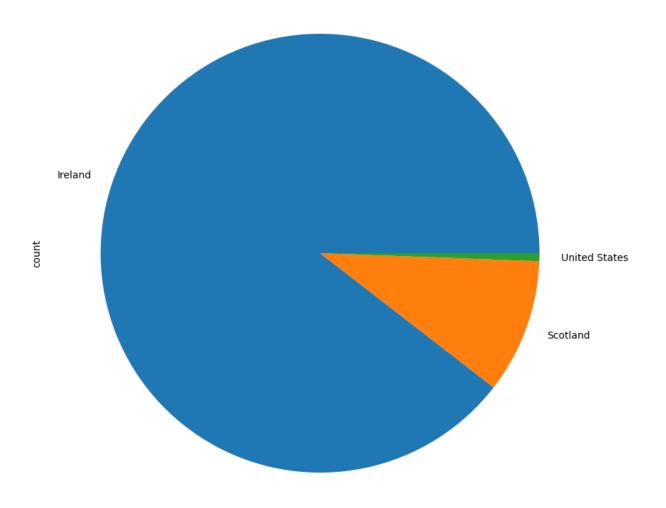


10. Where were passengers coming from? Print out the most frequent countries they were immigrating from and how many passengers were coming from each country. Hint: use value\_counts() and index.

```
In [31]: #Your Code Here
passengers_df['native_country'].value_counts()
```

```
Out[31]: native_country
Ireland 461
Scotland 51
United States 3
Name: count, dtype: int64
```

11. Make a pie chart showing **how many passengers were coming from each country**. Adapt Walsh's example.



12. Create a new variable called children\_filter and assign it a True/False statement to that variable that specifies passengers who were **children**. Then use this new children\_filter to create a new dataframe called children\_df that **only** 

**contains passengers who were children**. Display a sample of **five random rows** from this new dataframe. Hint: look under the occupation column in your dataframe. Hint: Walsh example.

Out[35]:	first_name	last_name	date	age	native_country	destination_city	destination_state
25	Alex	McBride	09 Apr 1884	12	Scotland	Chicago	IL
26	Cath	McBride	09 Apr 1884	11	Scotland	Chicago	IL
27	Wm	McBride	09 Apr 1884	9	Scotland	Chicago	IL
28	Agnes	McBride	09 Apr 1884	7	Scotland	Chicago	IL
29	Maggie	McBride	09 Apr 1884	4	Scotland	Chicago	IL
•••							
499	Ellen	Deran	09 Apr 1884	8	Ireland	Pittsburg	PA
503	Pat	Kyne	09 Apr 1884	9	Ireland	Portland	ME
504	John	Kyne	09 Apr 1884	3	Ireland	Portland	ME
505	Mich	Kyne	09 Apr 1884	0	Ireland	Portland	ME
511	Math	Griffins	09 Apr 1884	2	Ireland	Pittsburg	PA

132 rows × 9 columns

```
In [36]: children_df = passengers_df[children_filter]
```

children\_df.sample(5)

5]:		first_name	last_name	date	age	native_country	destination_city	destination_state
	49	Sarah	Watson	09 Apr 1884	4	Scotland	Auburn	NY
	447	John	Mulkern	09 Apr 1884	9	Ireland	Pittsburg	PA
	312	Mary	Divine	09 Apr 1884	11	Ireland	Fonda	NY
	468	Ann	Kerrigan	09 Apr 1884	4	Ireland	Winchester	MA
	354	Maggie	Brennan	09 Apr 1884	9	Ireland	New York	NY

13. Create a **new CSV file** named passenger-list-children.csv that only contains records for passengers who were children. Hint: you'll be printing the contents of children\_df to a CSV file using to\_csv() method. Walsh example. To check to make sure you successfully created the file, add a line of code that reads in the newly created CSV file using pd.read\_csv().

## **Bonus Questions**

What was the cut-off age for classifying a passenger as a child? le. What was **the oldest a passenger could be to still be considered a child**? Write code that prints out the answer to this question.

```
In [41]: #Your Code Here
children_df['age'].max()
```

Out[41]: 12

Out[36

Age Comparison: Calculate and write print() statements that show:

- The average age of passengers from Ireland
- The average age of passengers from **Scotland**.
- The difference in years between these average

```
In [43]: #Your Code Here
         ire_avg_age = passengers_df[passengers_df['native_country'] == 'Ireland']['age'].me
         scot_avg_age = passengers_df[passengers_df['native_country'] == 'Scotland']['age'].
In [44]: | age_diff = abs(ire_avg_age - scot_avg_age)
In [45]: print(ire_avg_age)
        20.98698481561822
In [46]: print(scot_avg_age)
        25.098039215686274
In [47]: print(age_diff)
        4.111054400068053
         Save a Filtered Dataset: Create a new CSV file that contains data for: only adult passengers
         (age 18 and over) who were heading to Boston.
         #Your Code Here
In [49]:
         adult_boston_filter = (passengers_df['age'] >= 18) & (passengers_df['destination_ci
         adult_boston_df = passengers_df[adult_boston_filter]
In [50]: | adult_boston_df.to_csv('passengers_adult_boston.csv', encoding='utf-8', index=False
 In [ ]:
```