

# Creating, Reading and Writing

```
In [12]: import pandas as pd
```

```
In [13]: pd.DataFrame({'Yes': [50, 21], 'No': [131, 2]})
```

Out[13]:

	Yes	No
0	50	131
1	21	2

```
In [14]: pd.DataFrame({'Bob': ['I liked it.', 'It was awful.'], 'Sue': ['Pretty good.',
```

Out[14]:

	Bob	Sue
0	I liked it.	Pretty good.
1	It was awful.	Bland.

```
In [15]: pd.DataFrame({'Bob': ['I liked it.', 'It was awful.'],
                        'Sue': ['Pretty good.', 'Bland.'],
                        index=['Product A', 'Product B'])
```

Out[15]:

	Bob	Sue
Product A	I liked it.	Pretty good.
Product B	It was awful.	Bland.

```
In [16]: pd.Series([1, 2, 3, 4, 5])
```

Out[16]:

```
0    1
1    2
2    3
3    4
4    5
dtype: int64
```

```
In [17]: pd.Series([30, 35, 40], index=['2015 Sales', '2016 Sales', '2017 Sales'], name
```

Out[17]:

```
2015 Sales    30
2016 Sales    35
2017 Sales    40
Name: Product A, dtype: int64
```

```
In [18]: wine_reviews = pd.read_csv(r"D:\Data Analytics\Python\Kaggle Pandas\winemag-da
```

```
In [19]: wine_reviews.shape
```

```
Out[19]: (65499, 14)
```

```
In [20]: wine_reviews.head()
```

```
Out[20]:
```

	Unnamed: 0	country	description	designation	points	price	province	region_1	region_2	t
0	0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	
1	1	Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	87	15.0	Douro	NaN	NaN	
2	2	US	Tart and snappy, the flavors of lime flesh and...	NaN	87	14.0	Oregon	Willamette Valley	Willamette Valley	
3	3	US	Pineapple rind, lemon pith and orange blossom ...	Reserve Late Harvest	87	13.0	Michigan	Lake Michigan Shore	NaN	
4	4	US	Much like the regular bottling from 2012, this...	Vintner's Reserve Wild Child Block	87	65.0	Oregon	Willamette Valley	Willamette Valley	

```
In [21]: wine_reviews = pd.read_csv(r"D:\Data Analytics\Python\Kaggle Pandas\winemag-da
wine_reviews.head()
```

Out[21]:

	country	description	designation	points	price	province	region_1	region_2	taster_name
0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	Kerin O'Keefe
1	Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	87	15.0	Douro	NaN	NaN	Roger Voss
2	US	Tart and snappy, the flavors of lime flesh and...	NaN	87	14.0	Oregon	Willamette Valley	Willamette Valley	Paul Gregutt
3	US	Pineapple rind, lemon pith and orange blossom ...	Reserve Late Harvest	87	13.0	Michigan	Lake Michigan Shore	NaN	Alexander Peartree
4	US	Much like the regular bottling from 2012, this...	Vintner's Reserve Wild Child Block	87	65.0	Oregon	Willamette Valley	Willamette Valley	Paul Gregutt



## Indexing, Selecting & Assigning

```
In [22]: pd.set_option('display.max_rows', 5)
```

```
In [23]: wine_reviews.country
```

```
Out[23]: 0      Italy
1      Portugal
...
65497    US
65498    Spain
Name: country, Length: 65499, dtype: object
```

```
In [24]: wine_reviews['country']
```

```
Out[24]: 0          Italy
          1      Portugal
          ...
        65497      US
        65498      Spain
          Name: country, Length: 65499, dtype: object
```

```
In [25]: wine_reviews['country'][0]
```

```
Out[25]: 'Italy'
```

## Indexing in pandas

```
In [26]: wine_reviews.iloc[0]
```

```
Out[26]: country          Italy
          description  Aromas include tropical fruit, broom, brimston...
          ...
          variety      White Blend
          winery        Nicosia
          Name: 0, Length: 13, dtype: object
```

```
In [27]: wine_reviews.iloc[:, 0]
```

```
Out[27]: 0          Italy
          1      Portugal
          ...
        65497      US
        65498      Spain
          Name: country, Length: 65499, dtype: object
```

```
In [28]: wine_reviews.iloc[:3, 0]
```

```
Out[28]: 0          Italy
          1      Portugal
          2          US
          Name: country, dtype: object
```

```
In [29]: wine_reviews.iloc[1:3, 0]
```

```
Out[29]: 1      Portugal
          2          US
          Name: country, dtype: object
```

```
In [30]: wine_reviews.iloc[[0, 1, 2], 0]
```

```
Out[30]: 0      Italy
          1    Portugal
          2         US
          Name: country, dtype: object
```

```
In [31]: wine_reviews.iloc[-5:]
```

```
Out[31]:
```

	country	description	designation	points	price	province	region_1	region_2	taster_name
65494	France	Made from young vines from the Vaulorent port...	Fourchaume Premier Cru	90	45.0	Burgundy	Chablis	NaN	Roger Vc
65495	Australia	This is a big, fat, almost sweet-tasting Caber...	NaN	90	22.0	South Australia	McLaren Vale	NaN	J Czerwin
65496	US	Much improved over the unripe 2005, Fritz's 20...	Estate	90	20.0	California	Dry Creek Valley	Sonoma	N
65497	US	This wine wears its 15.8% alcohol better than ...	Block 24	90	31.0	California	Napa Valley	Napa	N
65498	Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	90	10.0	Andalucia	Jerez	NaN	Mich: Schachr

## Label-based selection

```
In [32]: wine_reviews.loc[0, 'country']
```

```
Out[32]: 'Italy'
```

```
In [33]: wine_reviews.loc[:, ['taster_name', 'taster_twitter_handle', 'points']]
```

```
Out[33]:
```

	taster_name	taster_twitter_handle	points
0	Kerin O'Keefe	@kerinokeefe	87
1	Roger Voss	@vossroger	87
...	...	...	...
65497	NaN	NaN	90
65498	Michael Schachner	@wineschach	90

65499 rows × 3 columns

## Manipulating the index

```
In [34]: wine_reviews.set_index("title")
```

```
Out[34]:
```

	country	description	designation	points	price	province	region_1	region_2	taster
title									
Nicosia 2013 Vulkà Bianco (Etna)	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	(
Quinta dos Avidagos 2011 Avidagos Red (Douro)	Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	87	15.0	Douro	NaN	NaN	Rog
...	...	...	...	...	...	...	...	...	...
Hendry 2004 Block 24 Primitivo (Napa Valley)	US	This wine wears its 15.8% alcohol better than ...	Block 24	90	31.0	California	Napa Valley	Napa	
Bodegas Dios Baco S.L. NV Manzanilla Sherry (Jerez)	Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	90	10.0	Andalucia	Jerez	NaN	Sch

65499 rows × 12 columns



## Conditional selection

```
In [35]: wine_reviews.country == 'Italy'
```

```
Out[35]: 0          True
          1          False
          ...
        65497       False
        65498       False
          Name: country, Length: 65499, dtype: bool
```

```
In [36]: wine_reviews.loc[wine_reviews.country == 'Italy']
```

```
Out[36]:
```

	country	description	designation	points	price	province	region_1	region_2	taster
0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	C
6	Italy	Here's a bright, informal red that opens with ...	Belsito	87	16.0	Sicily & Sardinia	Vittoria	NaN	C
...	...	...	...	...	...	...	...	...	...
65477	Italy	Made of 65% Merlot, 25% Cabernet Sauvignon, 5%...	Ruit Hora	88	30.0	Tuscany	Bolgheri	NaN	C
65478	Italy	Aromas suggesting French oak, coconut and spic...	NaN	88	36.0	Tuscany	Vino Nobile di Montepulciano	NaN	C

10005 rows × 13 columns



```
In [37]: wine_reviews.loc[(wine_reviews.country == 'Italy') & (wine_reviews.points >= 9
```

Out[37]:

	country	description	designation	points	price	province	region_1	region_2	taste
120	Italy	Slightly backward, particularly given the vint...	Bricco Rocche Prapó	92	70.0	Piedmont	Barolo	NaN	
130	Italy	At the first it was quite muted and subdued, b...	Bricco Rocche Brunate	91	70.0	Piedmont	Barolo	NaN	
...	...	...	...	...	...	...	...	...	
65365	Italy	Stunning and sophisticated, it leads with inte...	Sanct Valentin	94	40.0	Northeastern Italy	Alto Adige	NaN	(
65399	Italy	Aesthetics and elegance are important values t...	Nectar Dei	91	65.0	Tuscany	Maremma	NaN	

3479 rows × 13 columns





```
In [38]: wine_reviews.loc[(wine_reviews.country == 'Italy') | (wine_reviews.points >= 90),:]  
# Suppose we'll buy any wine that's made in Italy or which is rated above average
```

Out[38]:

	country	description	designation	points	price	province	region_1	region_2	taster_name
0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	Kevin O'Keefe
6	Italy	Here's a bright, informal red that opens with ...	Belsito	87	16.0	Sicily & Sardinia	Vittoria	NaN	Kevin O'Keefe
...	...	...	...	...	...	...	...	...	...
65497	US	This wine wears its 15.8% alcohol better than ...	Block 24	90	31.0	California	Napa Valley	Napa	Natasha
65498	Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	90	10.0	Andalucia	Jerez	NaN	Michael Schachn

31430 rows × 13 columns

```
In [39]: wine_reviews.loc[wine_reviews.country.isin(['Italy', 'France'])]
```

Out[39]:

	country	description	designation	points	price	province	region_1	region_2	taster_name
0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	Ker O'Kee
6	Italy	Here's a bright, informal red that opens with ...	Belsito	87	16.0	Sicily & Sardinia	Vittoria	NaN	Ker O'Kee
...	...	...	...	...	...	...	...	...	...
65492	France	A rounded, fruity wine, packed with yellow pea...	Mont-de-Milieu Premier Cru	90	30.0	Burgundy	Chablis	NaN	Roger Vo:
65494	France	Made from young vines from the Vaulorent porti...	Fourchaume Premier Cru	90	45.0	Burgundy	Chablis	NaN	Roger Vo:

21179 rows × 13 columns



```
In [40]: wine_reviews.loc[wine_reviews.price.notnull()]
```

Out[40]:

	country	description	designation	points	price	province	region_1	region_2	taster_n
1	Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	87	15.0	Douro	NaN	NaN	Roger'
2	US	Tart and snappy, the flavors of lime flesh and...	NaN	87	14.0	Oregon	Willamette Valley	Willamette Valley	Paul Gr
...	...	...	...	...	...	...	...	...	...
65497	US	This wine wears its 15.8% alcohol better than ...	Block 24	90	31.0	California	Napa Valley	Napa	
65498	Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	90	10.0	Andalucia	Jerez	NaN	Mic Schac

60829 rows × 13 columns



## Assigning data

```
In [41]: wine_reviews['critic'] = 'everyone'
         wine_reviews['critic']
```

Out[41]:

```
0      everyone
1      everyone
...
65497  everyone
65498  everyone
Name: critic, Length: 65499, dtype: object
```

```
In [42]: # Or with an iterable of values:
```

```
wine_reviews['index_backwards'] = range(len(wine_reviews), 0, -1)
wine_reviews['index_backwards']
```

```
Out[42]: 0          65499
         1          65498
         ...
         65497         2
         65498         1
         Name: index_backwards, Length: 65499, dtype: int64
```

## Summary Functions and Maps

### Summary functions

```
In [43]: wine_reviews.points.describe()
```

```
Out[43]: count      65499.000000
         mean         88.434037
         ...
         75%         91.000000
         max        100.000000
         Name: points, Length: 8, dtype: float64
```

```
In [44]: wine_reviews.taster_name.describe()
```

```
Out[44]: count          51856
         unique           19
         top      Roger Voss
         freq          13045
         Name: taster_name, dtype: object
```

```
In [45]: wine_reviews.points.mean()
```

```
Out[45]: 88.43403716087269
```

```
In [46]: wine_reviews.taster_name.unique()
```

```
Out[46]: array(['Kerin O'Keefe', 'Roger Voss', 'Paul Gregutt',
                'Alexander Peartree', 'Michael Schachner', 'Anna Lee C. Iijima',
                'Virginie Boone', 'Matt Kettmann', nan, 'Sean P. Sullivan',
                'Jim Gordon', 'Joe Czerwinski', 'Anne Krebiehl\xa0MW',
                'Lauren Buzzeo', 'Mike DeSimone', 'Jeff Jenssen',
                'Susan Kostrzewa', 'Carrie Dykes', 'Fiona Adams',
                'Christina Pickard'], dtype=object)
```

```
In [47]: wine_reviews.taster_name.value_counts()
```

```
Out[47]: taster_name
Roger Voss      13045
Michael Schachner  7752
...
Fiona Adams      11
Christina Pickard   4
Name: count, Length: 19, dtype: int64
```

## Maps

```
In [48]: review_points_mean = wine_reviews.points.mean()
wine_reviews.points.map(lambda p: p - review_points_mean)
```

```
Out[48]: 0      -1.434037
1      -1.434037
...
65497    1.565963
65498    1.565963
Name: points, Length: 65499, dtype: float64
```

```
In [49]: def remean_points(row):
          row.points = row.points - review_points_mean
          return row

          wine_reviews.apply(remean_points, axis='columns')
```

Out[49]:

	country	description	designation	points	price	province	region_1	region_2	taster_
0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	-1.434037	NaN	Sicily & Sardinia	Etna	NaN	O'
1	Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	-1.434037	15.0	Douro	NaN	NaN	Roger
...	...	...	...	...	...	...	...	...	...
65497	US	This wine wears its 15.8% alcohol better than ...	Block 24	1.565963	31.0	California	Napa Valley	Napa	
65498	Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	1.565963	10.0	Andalucia	Jerez	NaN	M Scha

65499 rows × 15 columns



```
In [50]: wine_reviews.head(1)
```

Out[50]:

	country	description	designation	points	price	province	region_1	region_2	taster_name	t
0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	Kerin O'Keefe	



```
In [51]: review_points_mean = wine_reviews.points.mean()
wine_reviews.points - review_points_mean
```

```
Out[51]: 0          -1.434037
         1          -1.434037
         ...
        65497      1.565963
        65498      1.565963
        Name: points, Length: 65499, dtype: float64
```

```
In [52]: wine_reviews.country + " - " + wine_reviews.region_1
```

```
Out[52]: 0          Italy - Etna
         1              NaN
         ...
        65497      US - Napa Valley
        65498      Spain - Jerez
        Length: 65499, dtype: object
```

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 [53]: bargain_wine = wine_reviews.loc[(wine_reviews.points / wine_reviews.price).idxmax()]
bargain_wine
```

```
Out[53]: 'Bandit NV Merlot (California)'
```

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. (For simplicity, let's ignore the capitalized versions of these words.)

```
In [54]: tropical = wine_reviews.description.map(lambda x: "tropical" in x).sum()
         fruity = wine_reviews.description.map(lambda x: "fruity" in x).sum()
         descriptor_counts = pd.Series([tropical, fruity], index=['tropical', 'fruity'])
```

```
In [55]: descriptor_counts
```

```
Out[55]: tropical    1813
         fruity      4632
         dtype: int64
```

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

```
In [56]: def rating(row):
          return '3 stars' if row.points >= 95 or row.country == 'Canada' else ('2 s

star_ratings = wine_reviews.apply(rating, axis='columns')
star_ratings
```

```
Out[56]: 0          2 stars
          1          2 stars
          ...
          65497      2 stars
          65498      2 stars
          Length: 65499, dtype: object
```

## Grouping and Sorting

### Groupwise analysis

```
In [57]: wine_reviews.groupby('points').points.count()
```

```
Out[57]: points
          80      155
          81     305
          ...
          99      15
          100       8
          Name: points, Length: 21, dtype: int64
```

```
In [58]: # to get the cheapest wine in each point value category
          wine_reviews.groupby('points').price.min()
```

```
Out[58]: points
          80      5.0
          81      5.0
          ...
          99     75.0
          100    150.0
          Name: price, Length: 21, dtype: float64
```

```
In [59]: # here's one way of selecting the name of the first wine reviewed from each wi
          wine_reviews.groupby('winery').apply(lambda df: df.title.iloc[0])
```

```
Out[59]: winery
          1+1=3          1+1=3 NV Rosé Sparkling (Cava)
          10 Knots      10 Knots 2010 Viognier (Paso Robles)
          ...
          àMaurice      àMaurice 2013 Fred Estate Syrah (Walla Walla V...
          Štoka          Štoka 2009 Izbrani Teran (Kras)
          Length: 13549, dtype: object
```



```
In [60]: # how we would pick out the best wine by country and province
wine_reviews.groupby(['country', 'province']).apply(lambda df: df.loc[df.points
```

Out[60]:

	country	description	designation	points	price	province	region_1	region_2
country	province							
Argentina	Mendoza Province	Argentina	If you love massive Argentine reds with purity...	95	74.0	Mendoza Province	Mendoza	
	Other	Argentina	This single-vineyard Malbec blend from vineyard...	94	68.0	Other	Calchaquí Valley	
...	...	...	...	...	...	...	...	...
Uruguay	San Jose	Uruguay	Baked, sweet, heavy aromas turn earthy with ti...	87	50.0	San Jose	NaN	
	Uruguay	Uruguay	Cherry and berry aromas are ripe, healthy and ...	91	22.0	Uruguay	NaN	

385 rows × 15 columns



```
In [61]: # we can generate a simple statistical summary of the dataset
wine_reviews.groupby(['country']).price.agg([len, min, max])
```

Out[61]:

	len	min	max
country			
Argentina	1907	4.0	230.0
Armenia	1	14.0	14.0
...	...	...	...
Ukraine	5	6.0	10.0
Uruguay	61	10.0	120.0

41 rows × 3 columns

## Multi-indexes

```
In [62]: countries_reviewed = wine_reviews.groupby(['country', 'province']).description
countries_reviewed
```

Out[62]:

		len
country	province	
Argentina	Mendoza Province	1635
	Other	272
...	...	...
Uruguay	San Jose	3
	Uruguay	7

385 rows × 1 columns

```
In [63]: mi = countries_reviewed.index
type(mi)
```

Out[63]: pandas.core.indexes.multi.MultiIndex

```
In [64]: countries_reviewed.reset_index()
```

Out[64]:

	country	province	len
0	Argentina	Mendoza Province	1635
1	Argentina	Other	272
...	...	...	...
383	Uruguay	San Jose	3
384	Uruguay	Uruguay	7

385 rows × 3 columns

## Sorting

```
In [65]: countries_reviewed = countries_reviewed.reset_index()  
countries_reviewed.sort_values(by='len')
```

Out[65]:

	country	province	len
93	Croatia	Hrvatsko Primorje	1
291	Slovenia	Kras	1
...	...	...	...
375	US	Washington	4308
353	US	California	18122

385 rows × 3 columns

```
In [66]: countries_reviewed.sort_values(by='len', ascending=False)
```

Out[66]:

	country	province	len
353	US	California	18122
375	US	Washington	4308
...	...	...	...
329	South Africa	Vlootenburg	1
139	Greece	Beotia	1

385 rows × 3 columns

```
In [67]: countries_reviewed.sort_index()
```

Out[67]:

	country	province	len
0	Argentina	Mendoza Province	1635
1	Argentina	Other	272
...	...	...	...
383	Uruguay	San Jose	3
384	Uruguay	Uruguay	7

385 rows × 3 columns

```
In [68]: countries_reviewed.sort_values(by=['country', 'len'])
```

Out[68]:

	country	province	len
1	Argentina	Other	272
0	Argentina	Mendoza Province	1635
...	...	...	...
381	Uruguay	Montevideo	10
379	Uruguay	Canelones	24

385 rows × 3 columns

Create a Series whose index is the taster\_twitter\_handle category from the dataset, and whose values count how many reviews each person wrote.

```
In [69]: reviews_written = wine_reviews.groupby('taster_twitter_handle').size()

# or

# reviews_written = wine_reviews.groupby('taster_twitter_handle').taster_twitt
```

What are the minimum and maximum prices for each variety of wine? Create a DataFrame whose index is the variety category from the dataset and whose values are the min and max values thereof.

```
In [70]: price_extremes = wine_reviews.groupby(['variety']).price.agg([min, max])
price_extremes
```

Out[70]:

	min	max
variety		
Abouriou	75.0	75.0
Agiorgitiko	10.0	66.0
...	...	...
Çalkarası	19.0	19.0
Žilavka	15.0	15.0

590 rows × 2 columns

What are the most expensive wine varieties? Create a variable sorted\_varieties containing a copy of the dataframe from the previous question where varieties are sorted in descending order based on minimum price, then on maximum price (to break ties).

```
In [71]: sorted_varieties = price_extremes.sort_values(by=['min', 'max'], ascending = F
sorted_varieties
```

Out[71]:

	min	max
variety		
<b>Terrantez</b>	236.0	236.0
<b>Bual</b>	194.0	230.0
...	...	...
<b>Tempranillo-Malbec</b>	NaN	NaN
<b>Zelen</b>	NaN	NaN

590 rows × 2 columns

What combination of countries and varieties are most common? Create a Series whose index is a MultiIndex of {country, variety} pairs. For example, a pinot noir produced in the US should map to {"US", "Pinot Noir"}. Sort the values in the Series in descending order based on wine count.

```
In [72]: country_variety_counts = wine_reviews.groupby(['country', 'variety']).variety.
```

## Data Types and Missing Values

### Dtypes

```
In [73]: wine_reviews.price.dtype
```

Out[73]: dtype('float64')

```
In [74]: wine_reviews.dtypes
```

```
Out[74]: country          object
description          object
...
critic                object
index_backwards      int64
Length: 15, dtype: object
```

```
In [75]: wine_reviews.points.astype('float64')
```

```
Out[75]: 0      87.0
1      87.0
...
65497   90.0
65498   90.0
Name: points, Length: 65499, dtype: float64
```

```
In [76]: wine_reviews.index.dtype
```

Out[76]: dtype('int64')

Missing data

```
In [77]: wine_reviews[pd.isnull(wine_reviews.country)]
```

Out[77]:

	country	description	designation	points	price	province	region_1	region_2	taster_name
913	NaN	Amber in color, this wine has aromas of peach ...	Asureti Valley	87	30.0	NaN	NaN	NaN	Mik DeSimor
3131	NaN	Soft, fruity and juicy, this is a pleasant, si...	Partager	83	NaN	NaN	NaN	NaN	Roger Voisard
...	...	...	...	...	...	...	...	...	...
59670	NaN	The heady florality of damask rose is joined b...	Steintal	92	38.0	NaN	NaN	NaN	Anr Krebiehl M
60678	NaN	This wine was made for grilled meats, with its...	Dry	86	17.0	NaN	NaN	NaN	Susę Kostrzew

32 rows × 15 columns

```
In [78]: wine_reviews.region_2.fillna("Unknown")
```

Out[78]: 0 Unknown  
1 Unknown  
...  
65497 Napa  
65498 Unknown  
Name: region\_2, Length: 65499, dtype: object

```
In [79]: wine_reviews.taster_twitter_handle.replace("@kerinokeefe", "@kerino")
```

```
Out[79]: 0          @kerino
1        @vossroger
...
65497          NaN
65498    @wineschach
Name: taster_twitter_handle, Length: 65499, dtype: object
```

Create a Series from entries in the points column, but convert the entries to strings. Hint: strings are str in native Python.

```
In [80]: point_strings = wine_reviews.points.astype('str')
```

Sometimes the price column is null. How many reviews in the dataset are missing a price?

```
In [81]: missing_price_reviews = wine_reviews[wine_reviews.price.isnull()]
n_missing_prices = len(missing_price_reviews)
n_missing_prices

# Cute alternative solution: if we sum a boolean series, True is treated as 1
#n_missing_prices = reviews.price.isnull().sum()
# or equivalently:
#n_missing_prices = pd.isnull(reviews.price).sum()
```

```
Out[81]: 4670
```

What are the most common wine-producing regions? Create a Series counting the number of times each value occurs in the region\_1 field. This field is often missing data, so replace missing values with Unknown. Sort in descending order.

```
In [82]: reviews_per_region = wine_reviews.region_1.fillna('Unknown').value_counts().sort_index()
reviews_per_region
```

```
Out[82]: region_1
Unknown          10755
Napa Valley       2226
...
Civitella d'Agliano    1
Vino da Mesa de Toledo  1
Name: count, Length: 1112, dtype: int64
```

## Renaming and Combining

# Renaming

```
In [83]: wine_reviews.rename(columns={'points': 'score'})
```

Out[83]:

	country	description	designation	score	price	province	region_1	region_2	taster_name
0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	Ker O'Kee
1	Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	87	15.0	Douro	NaN	NaN	Roger Vos
...	...	...	...	...	...	...	...	...	...
65497	US	This wine wears its 15.8% alcohol better than ...	Block 24	90	31.0	California	Napa Valley	Napa	Na
65498	Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	90	10.0	Andalucia	Jerez	NaN	Micha Schachn

65499 rows × 15 columns





```
In [84]: wine_reviews.rename(index = {0: 'firstEntry', 1: 'secondEntry'})
```

Out[84]:

	country	description	designation	points	price	province	region_1	region_2	tas
firstEntry	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	
secondEntry	Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	87	15.0	Douro	NaN	NaN	Ri
...	...	...	...	...	...	...	...	...	
65497	US	This wine wears its 15.8% alcohol better than ...	Block 24	90	31.0	California	Napa Valley	Napa	
65498	Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	90	10.0	Andalucia	Jerez	NaN	ε

65499 rows × 15 columns



```
In [85]: wine_reviews.rename_axis("wines", axis='rows').rename_axis("fields", axis='col
```

Out[85]:

	fields	country	description	designation	points	price	province	region_1	region_2	taster_name
wines										
0		Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	Kevin O'Keefe
1		Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	87	15.0	Douro	NaN	NaN	Roger Voorn
...		...	...	...	...	...	...	...	...	...
65497		US	This wine wears its 15.8% alcohol better than ...	Block 24	90	31.0	California	Napa Valley	Napa	Natasha
65498		Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	90	10.0	Andalucia	Jerez	NaN	Michael Schachar

65499 rows × 15 columns

## Combining

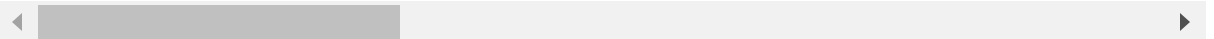
```
In [86]: canadian_youtube = pd.read_csv(r"D:\Data Analytics\Python\Kaggle Pandas\CAvide
british_youtube = pd.read_csv(r"D:\Data Analytics\Python\Kaggle Pandas\GBvideo

pd.concat([canadian_youtube, british_youtube])
```

Out[86]:

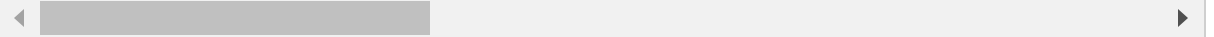
	video_id	trending_date	title	channel_title	category_id	publish_time	
0	n1WpP7iowLc	17.14.11	Eminem - Walk On Water (Audio) ft. Beyoncé	EminemVEVO	10	2017-11-10T17:00:03.000Z	E
1	0dBIkQ4Mz1M	17.14.11	PLUSH - Bad Unboxing Fan Mail	iDubbbzTV	23	2017-11-13T17:00:00.000Z	
...	...	...	...	...	...	...	...
38914	-DRsfNObKIQ	18.14.06	Eleni Foureira - Fuego - Cyprus - LIVE - First...	Eurovision Song Contest	24	2018-05-08T20:32:32.000Z	I
38915	4YFo4bdMO8Q	18.14.06	KYLE - Ikuyo feat. 2 Chainz & Sophia Black [A...	SuperDuperKyle	10	2018-05-11T04:06:35.000Z	t

49508 rows × 16 columns



```
In [93]: left = canadian_youtube.set_index(['title', 'trending_date'])
right = british_youtube.set_index(['title', 'trending_date'])
#left.join(right, lsuffix='_CAN', rsuffix='_UK')
#left.join(right)

# The lsuffix and rsuffix parameters are necessary here because the data has t
```



region\_1 and region\_2 are pretty uninformative names for locale columns in the dataset. Create a copy of reviews with these columns renamed to region and locale, respectively.

```
In [91]: renamed = wine_reviews.rename(columns=dict(region_1='region', region_2='locale',
renamed
```

Out[91]:

	country	description	designation	points	price	province	region	locale	taster_name	t
0	Italy	Aromas include tropical fruit, broom, brimston...	Vulkà Bianco	87	NaN	Sicily & Sardinia	Etna	NaN	Kerin O'Keefe	
1	Portugal	This is ripe and fruity, a wine that is smooth...	Avidagos	87	15.0	Douro	NaN	NaN	Roger Voss	
...	...	...	...	...	...	...	...	...	...	
65497	US	This wine wears its 15.8% alcohol better than ...	Block 24	90	31.0	California	Napa Valley	Napa	NaN	
65498	Spain	A unique take on Manzanilla Sherry, which is o...	Manzanilla	90	10.0	Andalucia	Jerez	NaN	Michael Schachner	

65499 rows × 15 columns



In [ ]: