

# All about Wine

March 12, 2021

## 1 All about Wine

**Purpose** To find out correlation of prices and points for each wine, and look for which country has the best wine of all. Also, calculate mean, min, and max points for countries Data can be found here: [Kaggle](#)

Key Performance Indicators: - Relationship between price and points - Highest wine points originated country - World choropleth

[ ]:

```
[1]: import numpy as np
import pandas as pd
```

```
[2]: import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
import chart_studio.plotly as py
import plotly.graph_objects as go
from plotly import __version__
from plotly.offline import download_plotlyjs, init_notebook_mode, plot, iplot

init_notebook_mode(connected = True)
```

```
[3]: import cufflinks as cf
```

```
[4]: wine1 = pd.read_csv("winemag-data-first150k.csv")
wine2 = pd.read_csv("winemag-data-130k-v2.csv")
```

```
[5]: wine = [wine1, wine2]
```

```
[6]: winemag = pd.concat(wine)
```

```
[38]: winemag.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 280901 entries, 0 to 129970
Data columns (total 14 columns):
#   Column                                Non-Null Count  Dtype
#   ...
```

```

---  -----
0  Unnamed: 0          280901 non-null  int64
1  country             280833 non-null  object
2  description         280901 non-null  object
3  designation         197701 non-null  object
4  points              280901 non-null  int64
5  price               258210 non-null  float64
6  province            280833 non-null  object
7  region_1            234594 non-null  object
8  region_2            111464 non-null  object
9  variety              280900 non-null  object
10 winery               280901 non-null  object
11 taster_name         103727 non-null  object
12 taster_twitter_handle 98758 non-null  object
13 title               129971 non-null  object
dtypes: float64(1), int64(2), object(11)
memory usage: 32.1+ MB

```

```
[39]: winemag.head()
```

```

[39]:  Unnamed: 0  country          description \
0          0      US  This tremendous 100% varietal wine hails from ...
1          1  Spain  Ripe aromas of fig, blackberry and cassis are ...
2          2      US  Mac Watson honors the memory of a wine once ma...
3          3      US  This spent 20 months in 30% new French oak, an...
4          4  France  This is the top wine from La Bégude, named aft...

          designation  points  price      province \
0          Martha's Vineyard      96  235.0      California
1  Carodorum Selección Especial Reserva      96  110.0  Northern Spain
2          Special Selected Late Harvest      96   90.0      California
3          Reserve      96   65.0          Oregon
4          La Brûlade      95   66.0      Provence

          region_1      region_2      variety \
0      Napa Valley      Napa  Cabernet Sauvignon
1          Toro      NaN      Tinta de Toro
2  Knights Valley      Sonoma  Sauvignon Blanc
3  Willamette Valley  Willamette Valley      Pinot Noir
4          Bandol      NaN  Provence red blend

          winery  taster_name  taster_twitter_handle  title
0          Heitz      NaN      NaN      NaN
1  Bodega Carmen Rodríguez      NaN      NaN      NaN
2          Macauley      NaN      NaN      NaN
3          Ponzi      NaN      NaN      NaN
4  Domaine de la Bégude      NaN      NaN      NaN

```

Filter only for relevant columns and exclude non-numerical or non-categorical columns

```
[7]: winemag = winemag[['country', 'points', 'price', 'province', 'region_1',  
    ↪ 'region_2', 'variety']]
```

```
[65]: #winemag['numbers'] = 1
```

```
[66]: winemag.head()
```

```
[66]:
```

	country	points	price	province	region_1 \
0	US	96	235.0	California	Napa Valley
1	Spain	96	110.0	Northern Spain	Toro
2	US	96	90.0	California	Knights Valley
3	US	96	65.0	Oregon	Willamette Valley
4	France	95	66.0	Provence	Bandol

	region_2	variety	numbers
0	Napa	Cabernet Sauvignon	1
1	NaN	Tinta de Toro	1
2	Sonoma	Sauvignon Blanc	1
3	Willamette Valley	Pinot Noir	1
4	NaN	Provence red blend	1

```
[6]: #winemag = winemag.rename(columns = {'Unnamed: 0': 'num'})
```

```
[42]: winemag.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
Int64Index: 280901 entries, 0 to 129970  
Data columns (total 7 columns):  
#   Column      Non-Null Count  Dtype  
---  -  
0   country     280833 non-null  object  
1   points      280901 non-null  int64  
2   price       258210 non-null  float64  
3   province    280833 non-null  object  
4   region_1    234594 non-null  object  
5   region_2    111464 non-null  object  
6   variety     280900 non-null  object  
dtypes: float64(1), int64(1), object(5)  
memory usage: 17.1+ MB
```

```
[43]: #How many unique countries on data  
winemag['country'].nunique()
```

```
[43]: 50
```

```
[44]: #unique countries array
winemag['country'].unique()
```

```
[44]: array(['US', 'Spain', 'France', 'Italy', 'New Zealand', 'Bulgaria',
        'Argentina', 'Australia', 'Portugal', 'Israel', 'South Africa',
        'Greece', 'Chile', 'Morocco', 'Romania', 'Germany', 'Canada',
        'Moldova', 'Hungary', 'Austria', 'Croatia', 'Slovenia', nan,
        'India', 'Turkey', 'Macedonia', 'Lebanon', 'Serbia', 'Uruguay',
        'Switzerland', 'Albania', 'Bosnia and Herzegovina', 'Brazil',
        'Cyprus', 'Lithuania', 'Japan', 'China', 'South Korea', 'Ukraine',
        'England', 'Mexico', 'Georgia', 'Montenegro', 'Luxembourg',
        'Slovakia', 'Czech Republic', 'Egypt', 'Tunisia', 'US-France',
        'Peru', 'Armenia'], dtype=object)
```

```
[45]: winemag['country'].value_counts()
```

```
[45]: US                116901
      France            43191
      Italy             43018
      Spain             14913
      Portugal          11013
      Chile             10288
      Argentina         9431
      Australia         7286
      Austria           6402
      New Zealand       4739
      Germany           4617
      South Africa      3659
      Greece            1350
      Israel            1135
      Canada            453
      Hungary           377
      Romania           259
      Bulgaria          218
      Uruguay           201
      Slovenia          181
      Croatia           162
      Turkey            142
      Mexico            133
      Moldova           130
      Georgia           129
      England           83
      Brazil            77
      Lebanon           72
      Cyprus            42
      Morocco           40
      Macedonia         28
```

Serbia	26
Ukraine	19
Czech Republic	18
India	17
Peru	16
Luxembourg	15
Switzerland	11
Lithuania	8
Bosnia and Herzegovina	6
Egypt	4
South Korea	4
China	4
Slovakia	4
Armenia	2
Tunisia	2
Albania	2
Montenegro	2
Japan	2
US-France	1

Name: country, dtype: int64

```
[8]: count = winemag['country'].value_counts()
count = count.to_frame().reset_index()
count.rename(columns={'index': 'country', 'country': 'value'}, inplace=True)
count
```

```
[8]:
```

	country	value
0	US	116901
1	France	43191
2	Italy	43018
3	Spain	14913
4	Portugal	11013
5	Chile	10288
6	Argentina	9431
7	Australia	7286
8	Austria	6402
9	New Zealand	4739
10	Germany	4617
11	South Africa	3659
12	Greece	1350
13	Israel	1135
14	Canada	453
15	Hungary	377
16	Romania	259
17	Bulgaria	218
18	Uruguay	201
19	Slovenia	181

20	Croatia	162
21	Turkey	142
22	Mexico	133
23	Moldova	130
24	Georgia	129
25	England	83
26	Brazil	77
27	Lebanon	72
28	Cyprus	42
29	Morocco	40
30	Macedonia	28
31	Serbia	26
32	Ukraine	19
33	Czech Republic	18
34	India	17
35	Peru	16
36	Luxembourg	15
37	Switzerland	11
38	Lithuania	8
39	Bosnia and Herzegovina	6
40	Egypt	4
41	South Korea	4
42	Slovakia	4
43	China	4
44	Japan	2
45	Armenia	2
46	Montenegro	2
47	Tunisia	2
48	Albania	2
49	US-France	1

```
[10]: bywinery_multiple = winemag.groupby(['country']).agg({'points':['mean', 'min', 'max']})
      bywinery_multiple.columns = ['points_mean', 'points_min', 'points_max']
      bywinery_multiple = bywinery_multiple.reset_index()
      bywinery_multiple
```

```
[10]:
```

	country	points_mean	points_min	points_max
0	Albania	88.000000	88	88
1	Argentina	86.283851	80	97
2	Armenia	87.500000	87	88
3	Australia	88.112407	80	100
4	Austria	89.707591	81	98
5	Bosnia and Herzegovina	85.333333	83	88
6	Brazil	84.207792	80	89
7	Bulgaria	87.064220	80	91
8	Canada	88.880795	82	94

9	Chile	86.382290	80	95
10	China	83.750000	82	89
11	Croatia	86.703704	81	91
12	Cyprus	86.214286	80	89
13	Czech Republic	86.777778	84	89
14	Egypt	83.750000	83	84
15	England	91.722892	89	95
16	France	88.884559	80	100
17	Georgia	86.961240	81	92
18	Germany	89.200996	80	98
19	Greece	86.520000	80	93
20	Hungary	88.050398	80	97
21	India	89.000000	82	93
22	Israel	87.752423	80	94
23	Italy	88.481147	80	100
24	Japan	85.000000	85	85
25	Lebanon	86.666667	81	91
26	Lithuania	84.250000	84	85
27	Luxembourg	87.666667	86	90
28	Macedonia	85.678571	81	89
29	Mexico	85.022556	80	92
30	Moldova	85.846154	81	91
31	Montenegro	82.000000	82	82
32	Morocco	88.450000	82	93
33	New Zealand	87.778434	80	95
34	Peru	83.562500	80	86
35	Portugal	88.157178	80	100
36	Romania	85.606178	80	92
37	Serbia	87.615385	86	89
38	Slovakia	84.500000	82	87
39	Slovenia	88.154696	82	92
40	South Africa	87.543591	80	95
41	South Korea	81.500000	81	82
42	Spain	86.932542	80	98
43	Switzerland	88.090909	83	90
44	Tunisia	86.000000	85	87
45	Turkey	88.091549	84	92
46	US	88.166106	80	100
47	US-France	88.000000	88	88
48	Ukraine	84.210526	82	88
49	Uruguay	85.711443	80	92

```
[11]: #Top 20 points by country
bywinery_multiple_20 = bywinery_multiple.nlargest(20, 'points_mean')
bywinery_multiple_20
```

```
[11]:
```

	country	points_mean	points_min	points_max
15	England	91.722892	89	95
4	Austria	89.707591	81	98
18	Germany	89.200996	80	98
21	India	89.000000	82	93
16	France	88.884559	80	100
8	Canada	88.880795	82	94
23	Italy	88.481147	80	100
32	Morocco	88.450000	82	93
46	US	88.166106	80	100
35	Portugal	88.157178	80	100
39	Slovenia	88.154696	82	92
3	Australia	88.112407	80	100
45	Turkey	88.091549	84	92
43	Switzerland	88.090909	83	90
20	Hungary	88.050398	80	97
0	Albania	88.000000	88	88
47	US-France	88.000000	88	88
33	New Zealand	87.778434	80	95
22	Israel	87.752423	80	94
27	Luxembourg	87.666667	86	90

```
[ ]:
```

Choropleth shows map figure but not identifying values – needs improvement and update

```
[12]: data = dict(
    type = 'choropleth',
    locations = winemag['country'],
    z = winemag['price'],
    text = winemag['country'],
    colorbar = {'title' : 'winery spread world'},
)
```

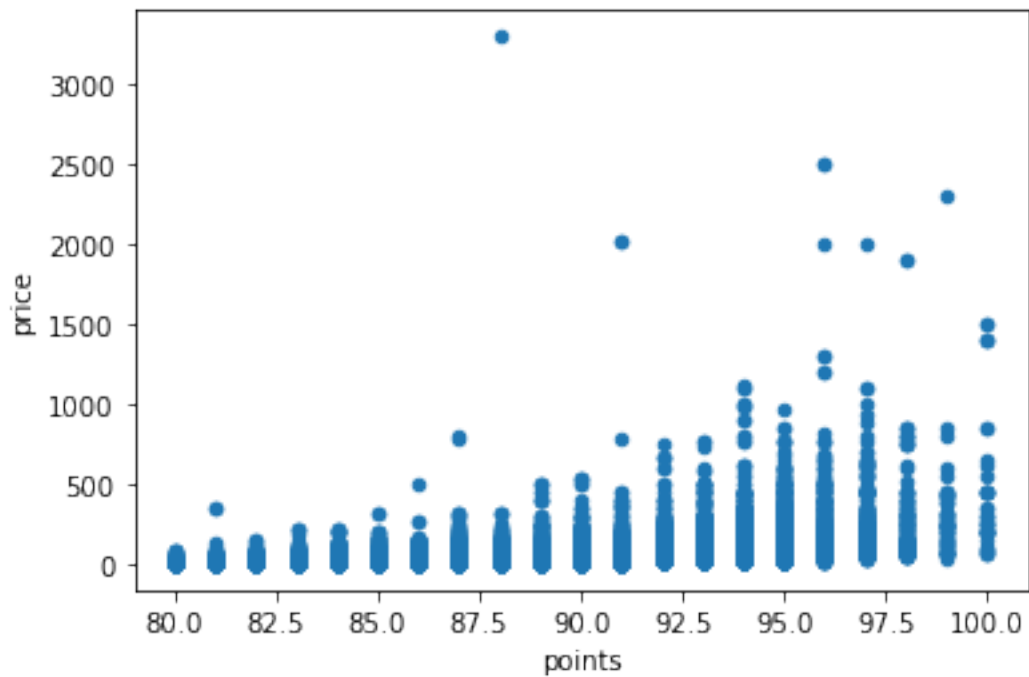
```
[13]: layout = dict(
    title = 'Wine World Map',
    geo = dict(
        showframe = False,
        projection = {'type': 'natural earth'}
    )
)
```

```
[14]: choromap = go.Figure(data = [data], layout = layout)
      iplot(choromap)
```

```
[19]: winemag.plot.scatter(x='points', y='price')
```

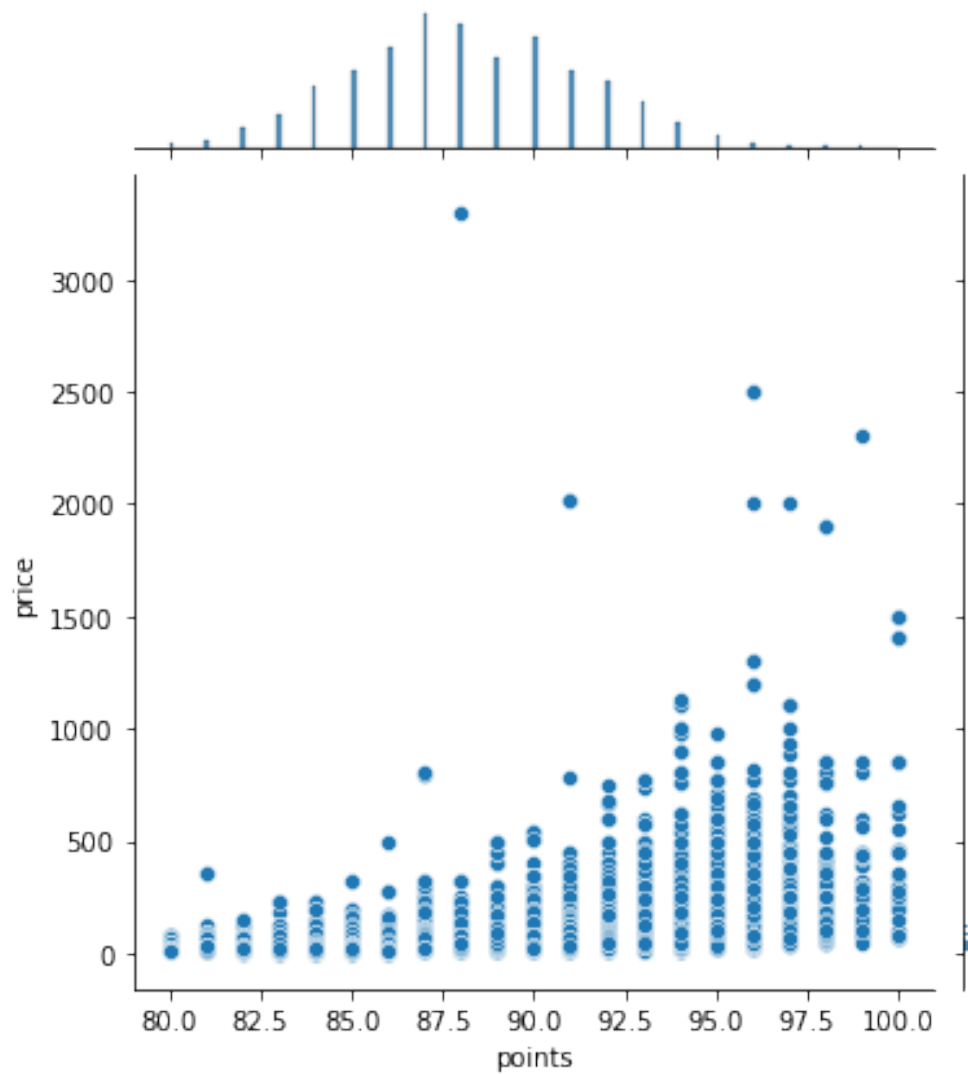


```
[19]: <AxesSubplot:xlabel='points', ylabel='price'>
```



```
[20]: sns.jointplot(x = 'points', y = 'price', data = winemag)
```

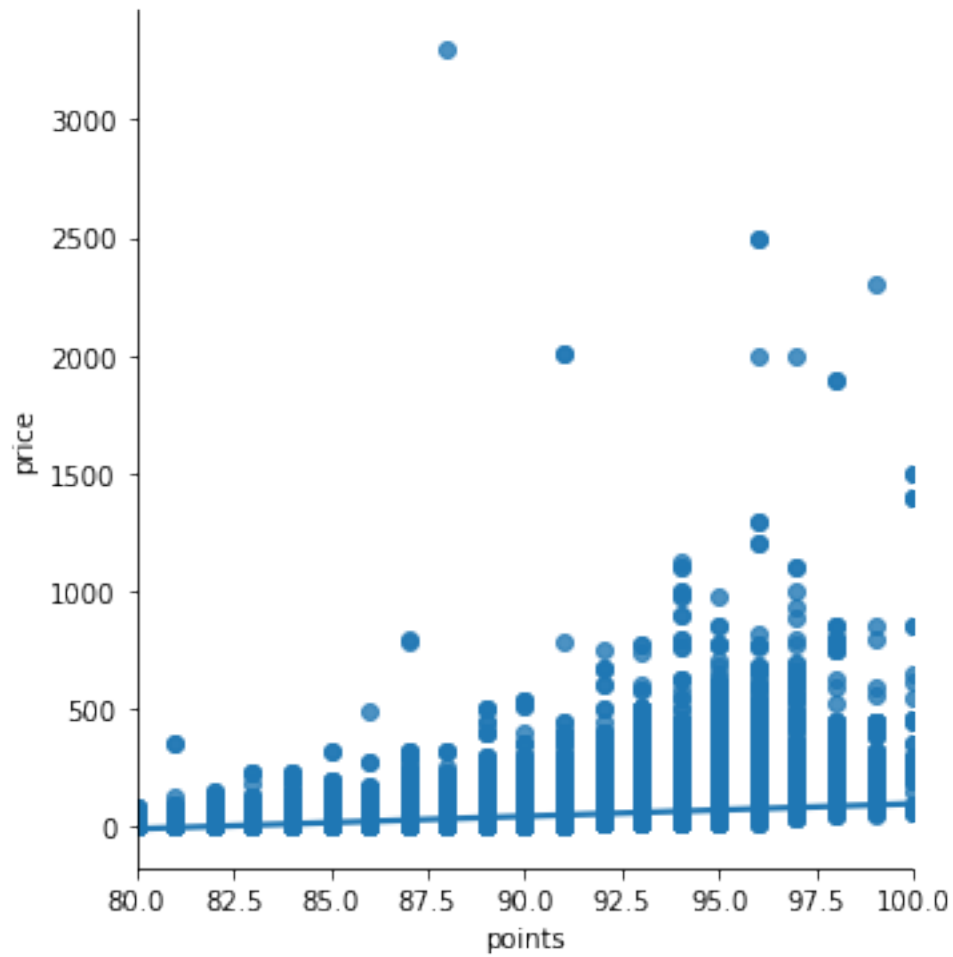
```
[20]: <seaborn.axisgrid.JointGrid at 0x7fa63d129460>
```



```
[26]: #sns.heatmap(winemag)
```

```
[22]: sns.lmplot(x='points',y='price',data=winemag)
```

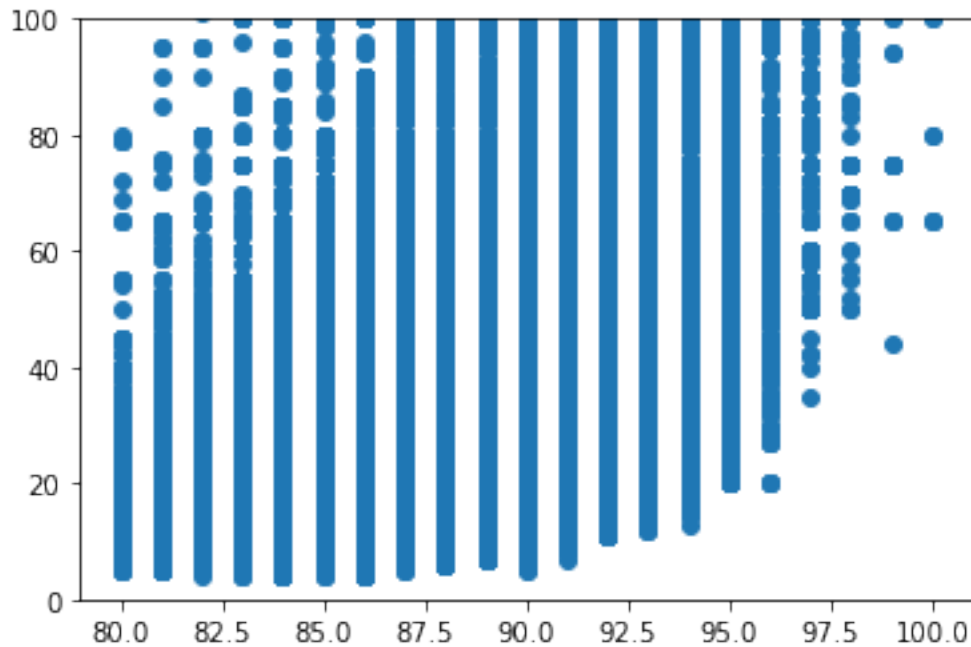
```
[22]: <seaborn.axisgrid.FacetGrid at 0x7fa6250c2d30>
```



```
[24]: fig = plt.figure()

plt.scatter(x = 'points', y = 'price', data = winemag)
plt.ylim([0,100])
```

```
[24]: (0.0, 100.0)
```



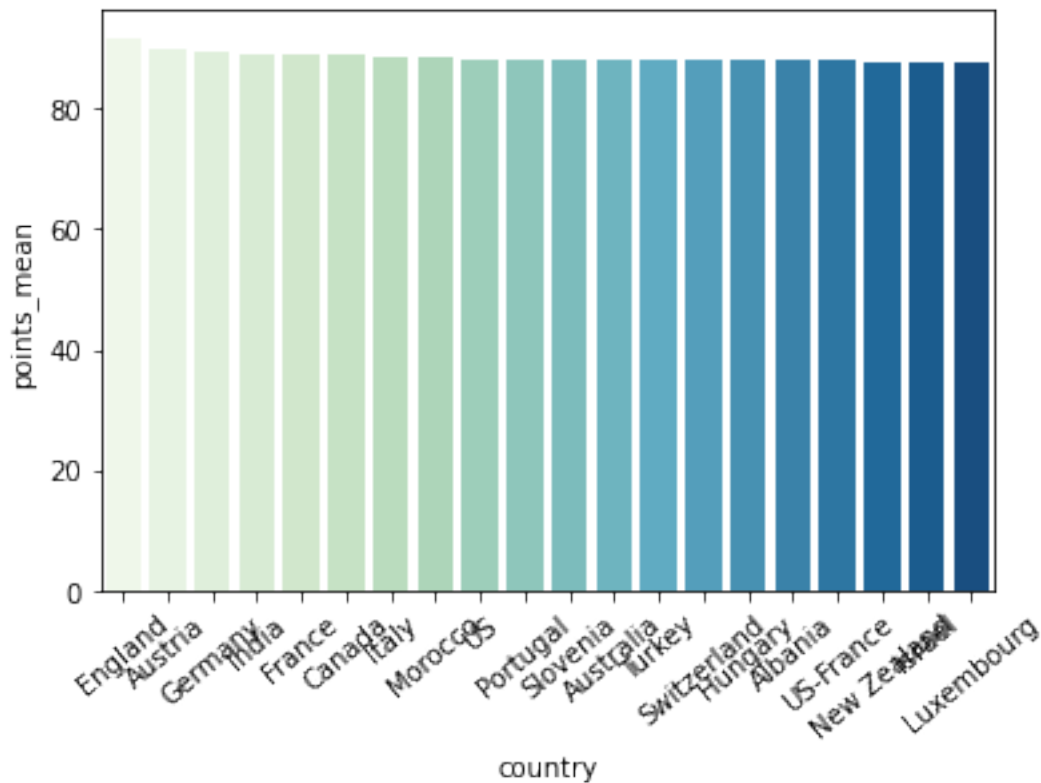
```
[15]: import plotly.express as px
fig = px.bar(bywinery_multiple_20, x = 'country', y = 'points_mean')
fig.show()
```

```
[16]: fig = px.scatter(winemag, x = 'price', y = 'points',
                      hover_name = 'country', log_x = True, size_max = 60)
fig.show()
```

```
[18]: g = sns.barplot(x = 'country', y = 'points_mean', data = bywinery_multiple_20,
                    palette = 'GnBu')
g.set_xticklabels(g.get_xticklabels(), rotation = 40)
```

```
[18]: [Text(0, 0, 'England'),
      Text(1, 0, 'Austria'),
      Text(2, 0, 'Germany'),
      Text(3, 0, 'India'),
      Text(4, 0, 'France'),
      Text(5, 0, 'Canada'),
      Text(6, 0, 'Italy'),
      Text(7, 0, 'Morocco'),
      Text(8, 0, 'US'),
      Text(9, 0, 'Portugal'),
      Text(10, 0, 'Slovenia'),
      Text(11, 0, 'Australia'),
      Text(12, 0, 'Turkey'),
```

```
Text(13, 0, 'Switzerland'),
Text(14, 0, 'Hungary'),
Text(15, 0, 'Albania'),
Text(16, 0, 'US-France'),
Text(17, 0, 'New Zealand'),
Text(18, 0, 'Israel'),
Text(19, 0, 'Luxembourg')]
```



## Conclusion

As you can see from Seaborn scatter plot, joint plot, lm (regression) plot, and Plotly scatter plot, wine price increases with its quality. High-quality wine costs a lot more than low-quality wine. Also, from separate data frame 'bywinery\_multiple\_20', it is shown top 20 points by country. Surprisingly, England takes first place while France and Italy which widely known as famous for wine sit 5th and 7th, respectively. However, when you look into data closely, England only has 83 unique data value counts (France: 43191, Italy: 43018 counts) and that lead to highest points (mean) overall. One thing I couldn't solve was with choropleth map. I was able to show map figure and color scales, but couldn't identify values. To do this, I added columns on dataframe to count, but didn't work out. This part will be updated.

[ ]: