Analyzing 160,000 Wine Spectator Reviews

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Data Source: https://www.kaggle.com/zynicide/wine-reviews

Objective:

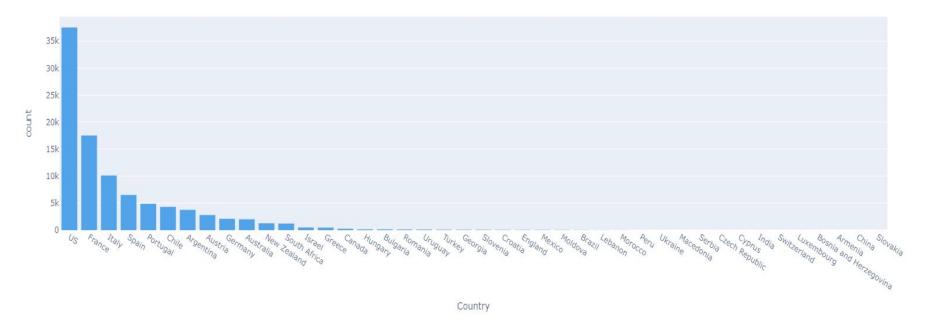
- Explore dataset, find anomalies
- Which regions produce the 'best' wine
- Make wine pairing easier

Data Breakdown

Understanding the Shape of the Data

Represented Countries

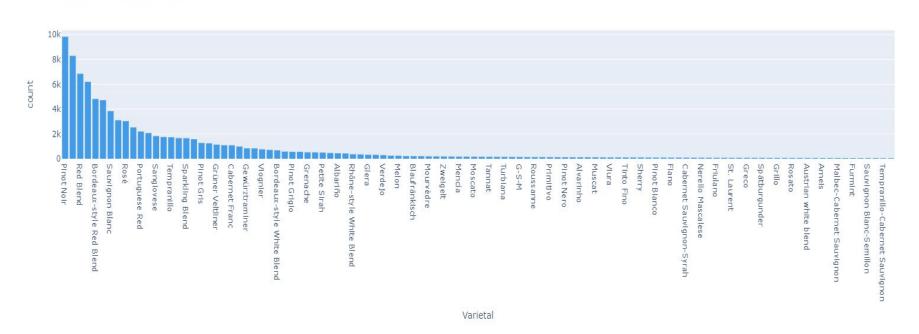
Number of Countries in Dataset



The vast majority of our data is concentrated in a small sample of countries

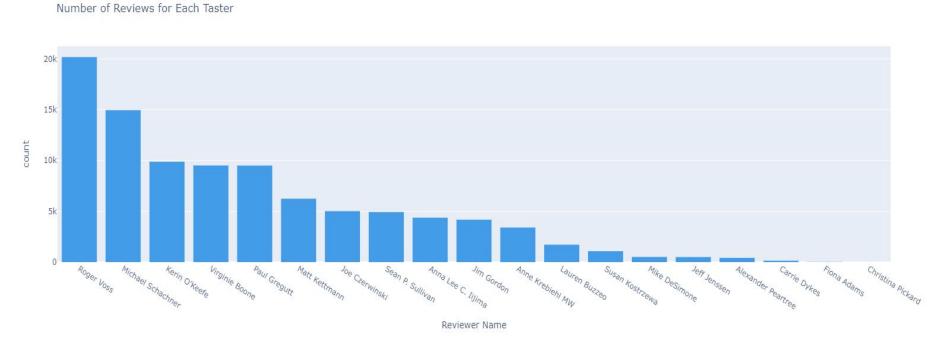
Represented Wine Varietals





Most of the reviews concentrate around a few dozen varieties

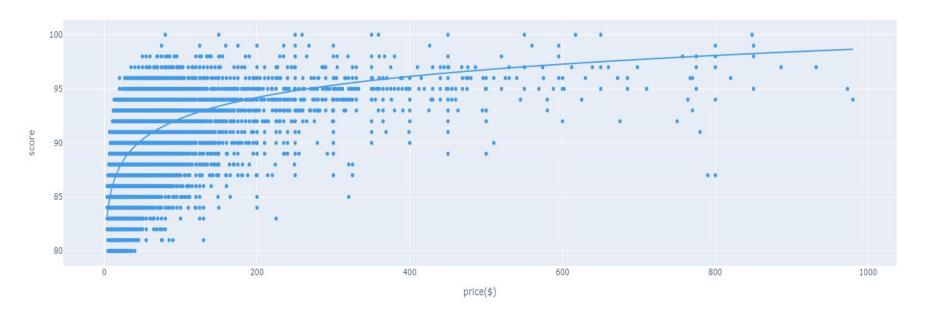
Number of Reviews by Each Taster



There are 19 total Tasters, and about a dozen prolific ones.

Correlation of Price and Review Score

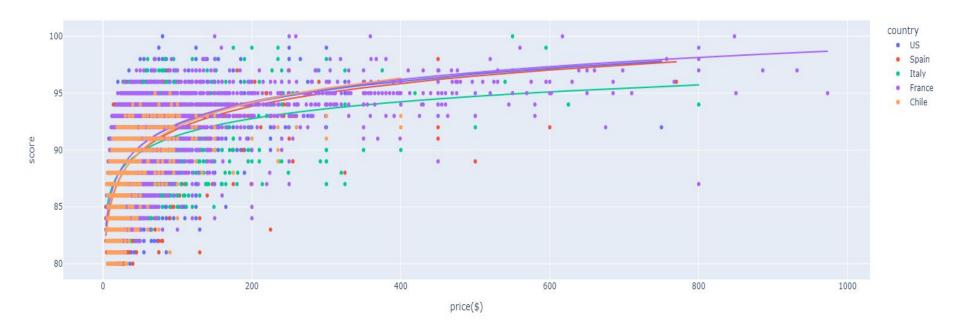
Price of wines less than \$1000 versus Score



A logarithmic relationship seems to exist between these two observations

Price Versus Review Score by Country

Price of wines less than \$1000 versus Score of the 5 most Reviewed Countries



Categorizing Price and Review Score

Defining 'Price' and 'Review Score' Buckets

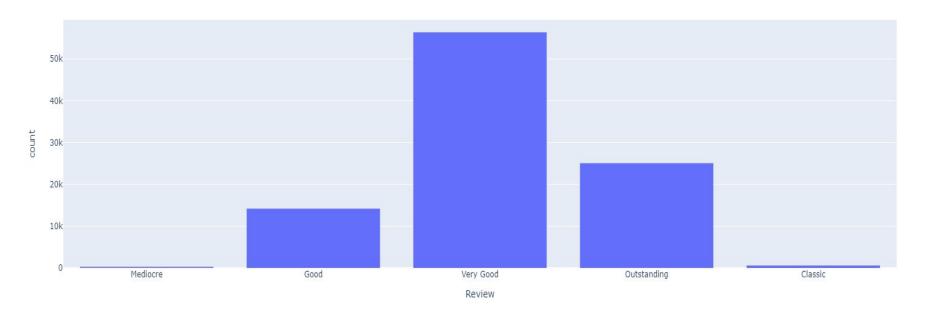
- Cheap: 4-17(\$)
- Inexpensive: 17-25(\$)
- Moderate: 26-42(\$)
- Pricey: 43-79(\$)
- Expensive: 80-175(\$)
- Outlandish: 175(\$) +

- 95-100 Classic: a great wine
- 90-94 Outstanding: a wine of superior character and style
- 85-89 Very good: a wine with special qualities
- 80-84 Good: a solid, well-made wine
- 75-79 Mediocre: a drinkable wine that may have minor flaws
- 50-74 Not recommended

Price broken down by quantile.
Review scores breakdown from Wine Spectator

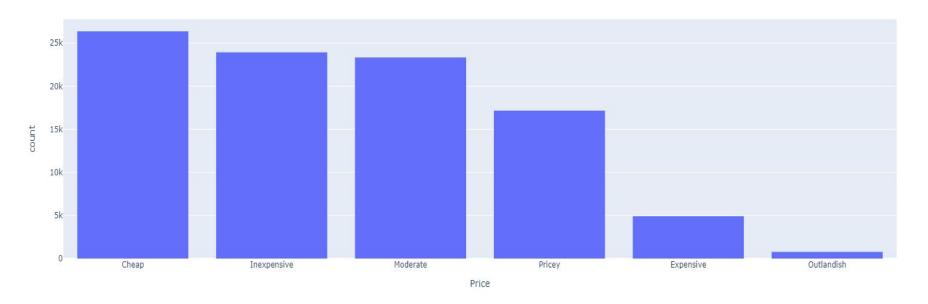
Distribution of Reviews

Distribution of Reviews



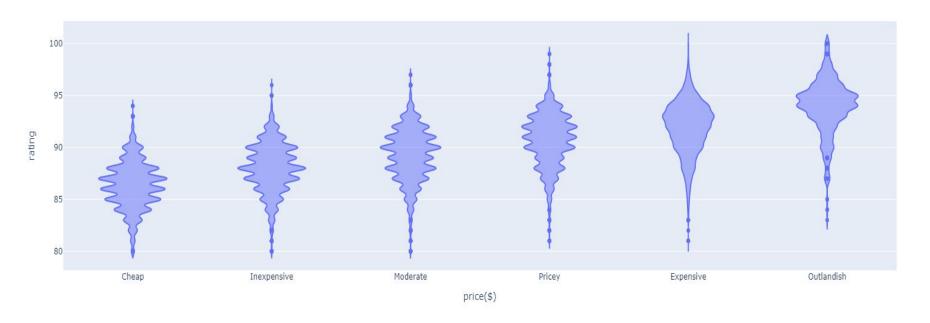
Distribution of Price

Distribution of Wine Price



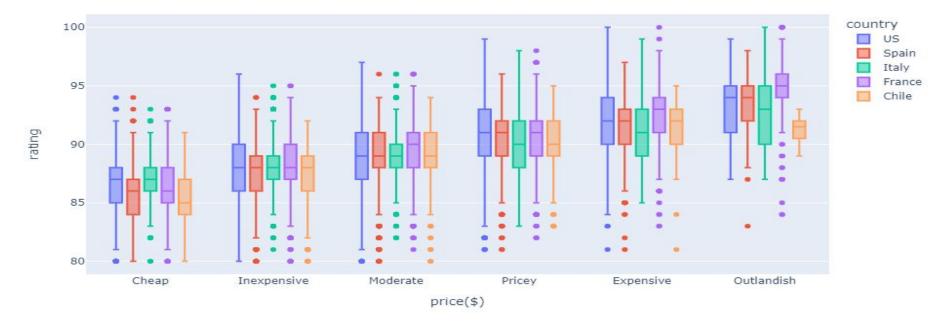
Distribution of Price Versus Review Score

Distribution of Rating vs Price Bucket



Review Score Distribution by Price Bucket and Country

Distribution of Rating vs Price Bucket by Country

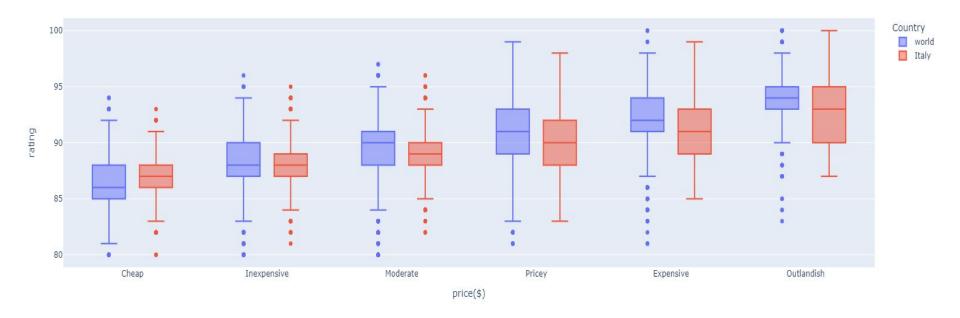


Italy, Are you OK?

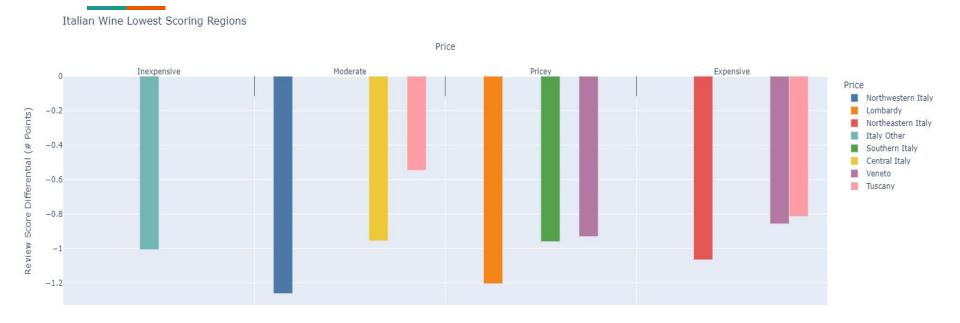
An Analysis of Italian Wines' Review Scores

Italian Wine Reviews Compared with the Rest of the World

Comparison of Review scores of Italy versus the rest of the world



Lowest Rated Italian Wine Regions



Lowest Rated Italian Wine Varieties



Conclusion: There are a handful of Italian regions and varietals pulling down their review scores.

Regions:

- Central Italy
- Northeastern Italy
- Northwestern Italy
- Tuscany
- Veneto
- Other

Varietals:

- Montepulciano
- Primitivo
- Negroamaro
- Tuscany
- Grillo
- Vernaccia

Finding Value in Wine Pairings

An Exploration into what regions produce good value

A Few Definitions

Value:

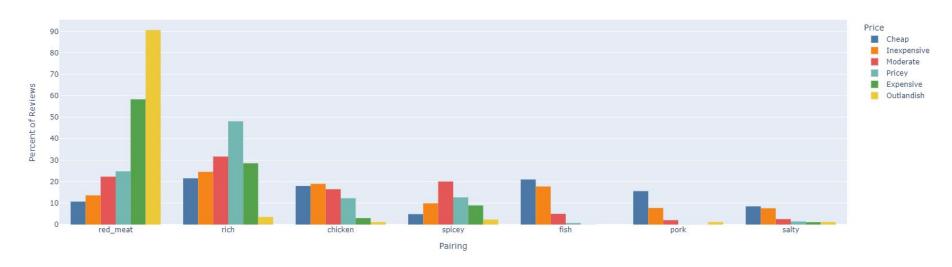
- Value in wine will be defined as a wine's review score divided by its price.
- This gives us a figure of merit in review score per dollars spent
- A higher relative number will be a greater value wine

Wine Pairings:

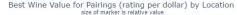
- Wine pairing is more art than science
- I will use my own experience in the field to define the 7 most common entree pairings
 - Chicken
 - Fish
 - Pork
 - Red Meat
 - Rich
 - Salty
 - Spicey
- See Appendix for a list of each wine appropriate for each pairing (and possibly an argument)

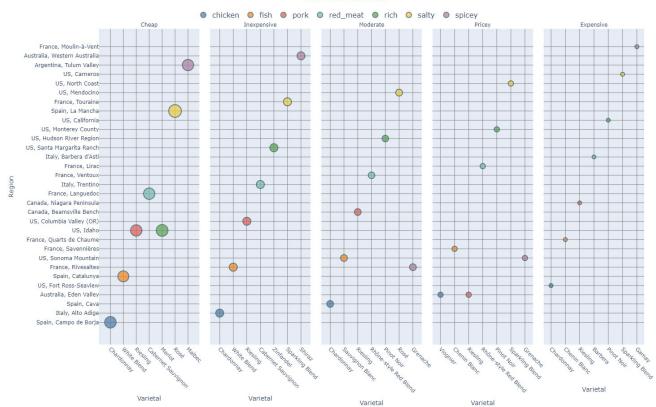
Number of Reviews for each Pairing

Total Percentage of Reviews for Each Pairing by Price

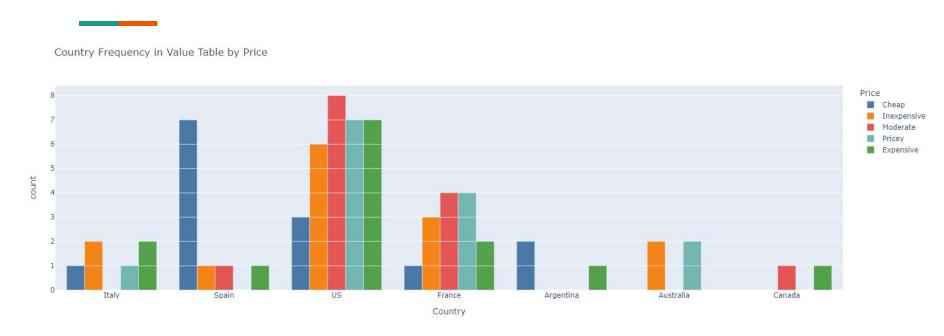


Value Based Wine Selection





Which Countries Produce the Highest Value Wine



The United States and France seem to produce the highest value wine

If Only We Had More Time

What I would do given more hours in the day

- Dive deeper into bias, I noticed there was a large bias to integer review scores on lower priced wine
- 2 tailed T-Test for individual reviewers by country showed very small p-values (~10^-4) against data set as a whole, want to explain
- There is an entire column in the dataset for verbal description of wine taste, would love to try and correlate certain adjectives with price and score
- Dive deeper into countries with smaller numbers of reviews and compare to the world

Appendix

Bonus Section:

Bias in Reviews

Are any of our reviewers biased against certain countries?

Bias Criteria

- A Reviewer must have a statistically significant number of reviews in a given country
- Must have a large population size to compare reviews against
- Correct for wine review score and price. For example, If a reviewer only reviews 'Inexpensive' wine, the scores will not reflect the population mean

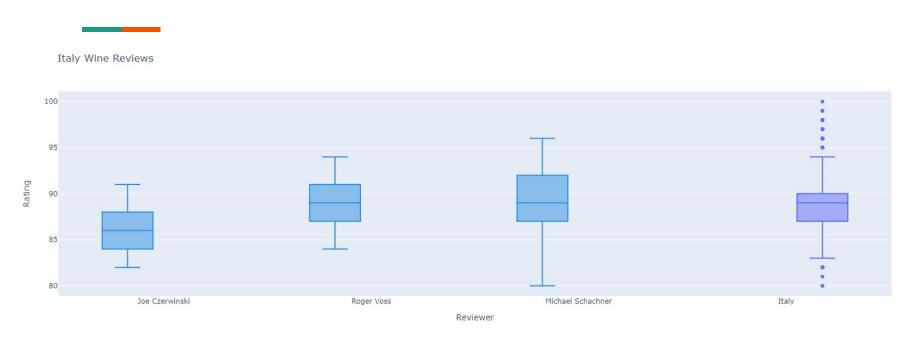
How Wine Spectator Makes This Difficult

- Many countries have a primary reviewer who conducts more than 95% of a countries reviews
- There are only 10 Reviewers with a significant number of reviews to perform statistical analysis
- Most reviewers concentrate their reviews in narrow price ranges outside of their most reviewed country

What can we do?

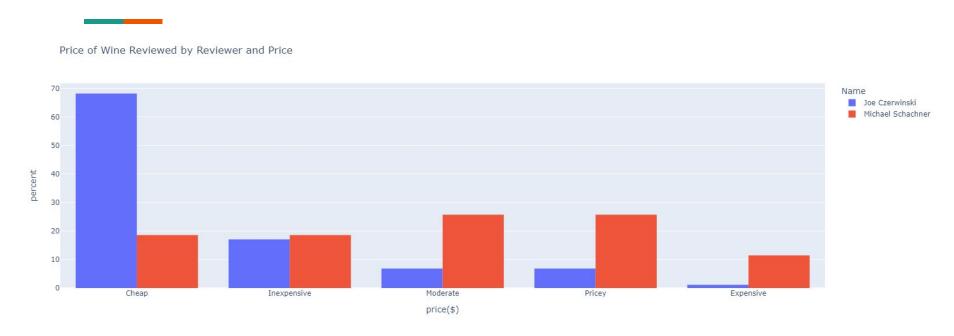
- There are two countries with enough reviewers to perform simple bias exploration (Italy and The US)
- We can look at the distribution of a single reviewer against the US and Italy and attempt to correct for price of wine reviewed

Italian Wine Reviews



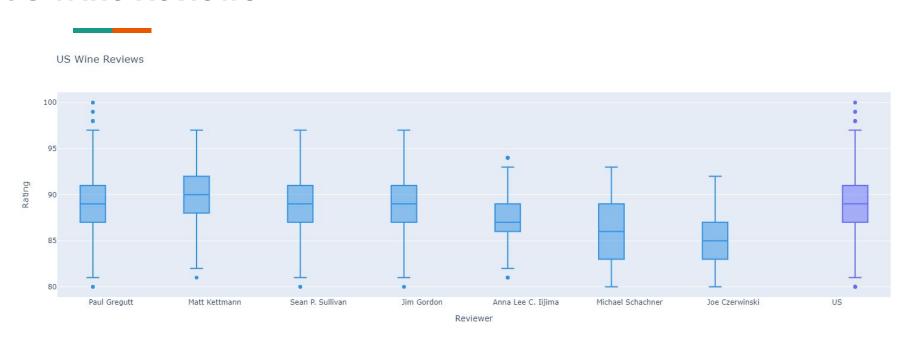
Joe Czerwinski seems to be biased against italian wines, Roger Voss' reviews seem to line up with the general italian wine reviews, while Michael Schachner seems to have a slight bias favoring italian wines.

Italian Wine Reviews by Price Bucket



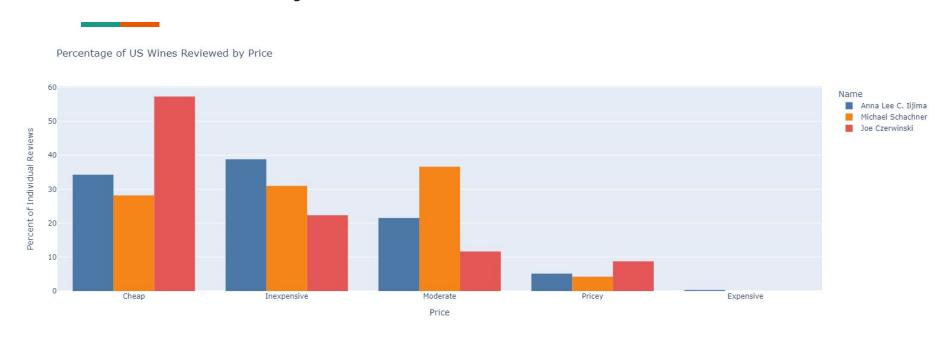
Both tasters bias' can be explained by their lop-sided selection of cheap wine in Joe Czerwinski's case, and more expensive wines in Michael Schachner's case

US Wine Reviews



Anna lijima, Michael Schachner, and Joe Czerwinski all seem to review US wine lower than the mean

US Wine Reviews by Price Bucket



The lower than average rating by these tasters can be explained by wine selection price

All apparent biases can be explained by wine price selection

Description of Numerical Observations by quantile

Description of Entire Dataset

Description of price > 42\$

| | price | points |
|-------|---------------|---------------|
| count | 120975.000000 | 120975.000000 |
| mean | 35.363389 | 88.421881 |
| std | 41.022218 | 3.044508 |
| min | 4.000000 | 80.000000 |
| 25% | 17.000000 | 86.000000 |
| 50% | 25.000000 | 88.000000 |
| 75% | 42.000000 | 91.000000 |
| max | 3300.000000 | 100.000000 |

| 22845.000000 76.115167 73.180289 | 22845.000000 91.070344 |
|--|---------------------------|
| | 91.070344 |
| 72 400200 | |
| 13.100209 | 2.570116 |
| 43.000000 | 81.000000 |
| 50.000000 | 90.000000 |
| 60.000000 | 91.000000 |
| 79.000000 | 93.000000 |
| 3300.000000 | 100.000000 |
| • | 60.000000 79.000000 |

Data Dictionary

Columns:

- **Country**: Country the wine is from
- **Description**: Full text review, includes tasting notes
- **Designation:** Name of the wine
- **Points:** Numerical review score from 50-100
- **Price:** Cost of bottle of wine (US Dollars)
- **Province :** The province or state the wine is from
- Region_1: Name of wine growing region in Province
- Taster_Name: Name of the person conducting the review

Wine Pairings List

- Fish: 'Sauvignon Blanc', 'Pinot Gris', 'White Blend', 'Chenin Blanc', 'Albariño', 'Pinot Blanc'
- Red Meat: 'Cabernet Sauvignon', 'Bordeaux-style Red Blend', 'Nebbiolo', 'Rhône-style Red Blend', 'Cabernet Franc', 'Barbera', 'Verdejo', 'Petit Verdot'
- Salty: 'Rosé', 'Sparkling Blend', 'Champagne Blend', 'Glera'
- **Spicey**: 'Syrah', 'Malbec', 'Tempranillo', 'Gamay','Shiraz', 'Tempranillo Blend', 'Grenache', 'Petite Sirah', 'Garnacha'
- Rich: 'Pinot Noir', 'Merlot', 'Sangiovese', 'Zinfandel', 'Carmenère', 'Torrontés'
- **Pork:** 'Riesling','Grüner Veltliner', 'Gewürztraminer', 'Blaufränkisch'
- Chicken: 'Chardonnay', 'Portuguese White', 'Viognier', 'Bordeaux-style White Blend', 'Rhône-style White Blend'

Code for Represented Countries Graph

Code for Represented Varietals Graph

Code for Tasters Graph

Code for Price V Score Graph

Code of bucketing price and points

Create Buckets

```
cut_labels = ['Cheap', 'Inexpensive', 'Moderate', 'Pricey', 'Expensive', 'Outlandish']
cut_bins = [4, 17, 25, 42, 79, 175, 3300]
# start with fresh DF
reviewsCleanBuckets = reviewsClean
# create series for new row
price buckets = pd.cut(reviewsCleanBuckets['price'], bins=cut bins,
labels=cut labels)
# concat to avoid assigning into a copy of a DF
reviewsCleanBuckets = pd.concat([reviewsCleanBuckets, price buckets], axis = 1)
# Rename columns
reviewsCleanBuckets.columns = ['Unnamed: 0', 'country', 'description', 'designation',
'points'.
   'price', 'province', 'region 1', 'region 2', 'taster name',
   'taster twitter handle', 'title', 'variety', 'winery', 'price bucket']
#drop unneeded column
reviewsCleanBuckets.drop('Unnamed: 0', axis = 1)
```

Code of Review Distribution Graph

Code of Price Distribution Graph

Code of Price versus review Distribution Graph

Code of Price Versus Review Score by Country Graph

Code of Italy Vs World Review Graph

Code of Italy Region Reviews Graph

```
#Create working DF
onlyItalyDF = italyDF.loc[italyDF['country'] == 'Italy']
# group by price and province, aggregate points
dfh = onlyItalyDF.groupby(['price bucket', 'province']).agg(
average_points = ('points', 'mean')
).dropna()
#collapse index, get the top 3 for each group
h = dfh.groupby(level=0, group keys=False).apply(
  lambda x: x.sort values(('average points'),).head(3)).reset index()
h = h.merge(italyDF.groupby('price_bucket')['points'].agg('mean'), how = 'inner', on = 'price_bucket')
# create difference column (world average minus itally average)
h['difference'] = h['average points'] - h['points']
h.sort_values('difference', ascending = False).head(10)
fig = px.bar(h.loc[h['price bucket']!= 'Outlandish'].sort values('difference').head(10), color = 'province', y = 'difference', x = 'price bucket',
     labels = {
       'difference': 'Review Score Differential (# Points)',
       'province': 'Region',
       'price bucket': 'Price'
           category orders = {
        'price bucket': ['Cheap, 'Inexpensive, 'Moderate, 'Pricey, 'Expensive, 'Outlandish']
      barmode = 'group',
      color discrete sequence=px.colors.qualitative.T10,
     title = 'Italian Wine Lowest Scoring Regions')
fig.update xaxes(showgrid = True. ticks='inside', tickson='boundaries',ticklen = 30)
fig.update layout(legend title = 'Price', xaxis = {'side': 'top'}, bargroupgap= 0.0,
       title = {
         'y':.98
fig
```

Code of Italy Varietal Reviews Graph

```
#group by varietal and price
dfg = onlyItalyDF.sort_values('points').groupby(['price_bucket', 'variety']).agg(
average_points = ('points', 'mean'),
 count = ('variety', 'count')
).dropna()
# take the 5 worst rated varietals for each price bucket
g = dfg.loc[dfg['count'] > 5].groupby(level=0, group keys=False).apply(
  lambda x: x.sort_values(('average_points'),).head(5)).reset_index()
#compare to world average by price bucket
g = g.merge(italyDF.groupby('price_bucket')['points'].agg('mean'), how = 'inner', on = 'price_bucket')
g['difference'] = g['average_points'] - g['points']
g.sort values('difference').head(10).sort values('price bucket')
fig = px.bar(g.loc[g['price_bucket']!= 'Outlandish'].sort_values('difference', ascending = False).head(16),
      color = 'variety', y = 'difference', x = 'price bucket',
     labels = {
       'difference': 'Review Score Differential (# Points)'.
       'variety': 'Varietal',
       'price bucket': 'Price'
      barmode='group'.
      color_discrete_sequence=px.colors.qualitative.T10,
     title = 'Italian Wine Lowest Scoring Varietals')
fig.update_layout(legend_title = 'Price', xaxis = {'side': 'top'}, bargroupgap = 0)
fig.update xaxes(showgrid = True, ticks='inside', tickson='boundaries', ticklen = 30)
#fig.update traces(width=.3)
fig
```

Code for pairings by price bucket Graph

```
pairingsDict = {'fish': ['Sauvignon Blanc', 'Pinot Gris', 'White Blend', 'Chenin Blanc', 'Albariño', 'Pinot Blanc'],
'red_meat': ['Cabernet Sauvignon', 'Bordeaux-style Red Blend', 'Nebbiolo', 'Rhône-style Red Blend', 'Cabernet Franc', 'Barbera', 'Verdejo', 'Petit Verdot'],
'salty': ['Rosé', 'Sparkling Blend', 'Champagne Blend', 'Glera'],
'spicey': ['Syrah', 'Malbec', 'Tempranillo,' Gamay, 'Shiraz', 'Tempranillo Blend', 'Grenache', 'Petite Sirah', 'Garnacha'],
'rich': ['Pinot Noir', 'Merlot', 'Sangiovese', 'Zinfandel', 'Carmenère', 'Torrontés'],
'pork': ['Riesling', Grüner Veltliner', 'Gewürztraminer', 'Blaufränkisch'],
'chicken': ['Chardonnay', 'Portuguese White', 'Viognier', 'Bordeaux-style White Blend', 'Rhône-style White Blend']
pairings = []
for i in topVarDF.variety:
 for j in pairingsDict:
   if i in pairingsDict[i]:
     pairings.append(j)
pairings = pd.Series(pairings)
topVarDF = pd.concat([topVarDF, pairings], axis = 1)
topVarDF.columns = ['country', 'description', 'designation', 'points', 'price', 'province', 'region 1', 'region 2',
        'taster_name', 'taster_twitter_handle', 'title', 'variety', 'winery', 'price_bucket', 'pairing']
fig = px.histogram(topVarDF.dropna(), x = 'pairing', color = 'price bucket'.
               category orders = {
        'price bucket': ['Cheap', 'Inexpensive', 'Moderate', 'Pricey', 'Expensive', 'Outlandish']
        color discrete sequence=px.colors.qualitative.T10.
         labels = {
           'pairing': 'Pairing'.
           'price bucket': 'Price'
        title = 'Total Percentage of Reviews for Each Pairing by Price',
        barmode = 'group'.
        histnorm = 'percent'
fig.update xaxes(categoryorder = 'total descending')
fig.update layout(vaxis title = 'Percent of Reviews')
fig
```

Code for pairings by price bucket Graph

```
valueDF['location'] = valueDF['country'] + ', ' + valueDF['region 1']
fig = px.scatter(valueDF, y = 'location', x = 'variety', color = 'pairing', facet_col = 'price_bucket',
       size = 'value',
       labels = {
          'location': 'Region',
         'pairing':",
         'price_bucket': 'Price',
         'variety': 'Varietal',
       title = 'Best Wine Value for Pairings (rating per dollar) by Location <br/> <sup>size of marker is relative value </sup>',
       color discrete sequence=px.colors.gualitative.T10,
       height = 900,
       width = 1500)
fig.update_xaxes(matches=None)
fig.update layout(title x=0.5, title y=.96,
        legend=dict(
 orientation="h".
 yanchor="bottom",
 y=1.02,
 xanchor="right",
 x = .7,
 font=dict(
 size=16
fig.update traces(marker=dict(line=dict(width=2,
                  color='DarkSlateGrev')).
        selector=dict(mode='markers'))
fig.for_each_annotation(lambda a: a.update(text=a.text.split("=")[-1]))
fig.update xaxes(showgrid=True, gridwidth=1, gridcolor='gray')
fig.update yaxes(showgrid=True, gridwidth=1, gridcolor='gray')
fig
```

Code for countries with highest value Graph

Code for Italian wine reviewers Graph

```
graphDF1 = topReviewersDF.loc[(topReviewersDF['taster_name'].isin(['Joe Czerwinski', 'Michael Schachner', 'Roger Voss']))&
             (topReviewersDF['country'] == 'Italy')]
graphDF2 = reviewsClean.loc[reviewsClean['country'] == 'Italy']
fig = px.box(graphDF1, x = 'taster_name', y="points",
     labels = {
         'points': 'Rating',
         'taster_name': 'Reviewer'
     color_discrete_sequence=px.colors.qualitative.Dark24,
     title = 'Italy Wine Reviews'
fig2 = px.box(graphDF2, x='country', y='points',
     labels = {
       'country': 'Total Reviws'
fig.add_trace(fig2.data[0])
fig.show()
```

Code for Italian wine reviewers by Price Graph

```
graphDF1 = popReviewsCleanBuckets.loc[(popReviewsCleanBuckets['taster_name'].isin(
    ['Joe Czerwinski', 'Michael Schachner'])) & (popReviewsCleanBuckets['country'] == 'Italy')]
fig = px.histogram(graphDF1, x = 'price_bucket', color = 'taster_name',
    category_orders = {
        'price_bucket' : ['Cheap', 'Inexpensive', 'Moderate', 'Pricey', 'Expensive', 'Outlandish']
    },
    labels = {
        'price_bucket' : 'price($)',
        'taster_name' : 'Name'
    },
        barmode = 'group',
        histnorm = 'percent',
        title = 'Price of Wine Reviewed by Reviewer and Price'
    )
fig.show()
```

Code for US wine reviewers Graph

```
USReviewers = ['Anna Lee C. lijima', 'Jim Gordon',
      'Joe Czerwinski', 'Matt Kettmann', 'Michael Schachner', 'Paul Gregutt', 'Sean P. Sullivan']
graphDF1 = popReviewsCleanBuckets.loc[(popReviewsCleanBuckets['taster_name'].isin(USReviewers))&
             (popReviewsCleanBuckets['country'] == 'US')]
graphDF2 = reviewsClean.loc[reviewsClean['country'] == 'US']
fig = px.box(graphDF1, x = 'taster_name', y="points",
     labels = {
         'points': 'Rating',
         'taster_name': 'Reviewer'
     color discrete sequence=px.colors.qualitative.Dark24,
     title = 'US Wine Reviews'
fig2 = px.box(graphDF2, x='country', y='points',
      labels = {
       'country': 'Total Reviws'
fig.add_trace(fig2.data[0])
fig.show()
```

Code for US wine reviewers by price Graph

```
USReviewers = ['Anna Lee C. lijima', 'Jim Gordon',
      'Joe Czerwinski', 'Matt Kettmann', 'Michael Schachner', 'Paul Gregutt', 'Sean P. Sullivan']
graphDF1 = popReviewsCleanBuckets.loc[(popReviewsCleanBuckets['taster_name'].isin(USReviewers))&
             (popReviewsCleanBuckets['country'] == 'US')]
graphDF2 = reviewsClean.loc[reviewsClean['country'] == 'US']
fig = px.box(graphDF1, x = 'taster_name', y="points",
     labels = {
         'points': 'Rating',
         'taster_name': 'Reviewer'
     color discrete sequence=px.colors.qualitative.Dark24,
     title = 'US Wine Reviews'
fig2 = px.box(graphDF2, x='country', y='points',
      labels = {
       'country': 'Total Reviws'
fig.add_trace(fig2.data[0])
fig.show()
```