Probability Theory and Statistics



ALY6010, WINTER 2022

Module 2 Final Project – Milestone 1

Week-2

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Module 2

Week 2

Introduction: Final Project Milestone-1

Exploratory Data Analysis

Introduction:

Wine tasting is as old as the industry itself, with a more systematic approach gradually emerging from the 14th century onwards. To define a wine's spectrum of perceived flavors, aromas, and overall attributes, professional wine tasters (such as sommeliers or retail consumers) use a constantly developing specialized lexicon. Similar terminology could be used in more casual, recreational evaluations, which often require a considerably less objective manner for a more general, portion of one's personality.

This project is intended to use exploratory data analysis (EDA) approaches to investigate correlations between 14 variables, as well as to find the solutions for visualizations, distributions, outliers in a selected wine taste data set.

There are 1000 observations and 14 variables in this dataset. This dataset, which includes the variables price, designation, description, province, region 1, region 2, flavor, taster, taster name, Twitter handle, title, variety, and winery, will help in offering insights into the wine taste. Here, will show performing statistical analysis while performing command in R.

🔾 wine_data	1000	obs. of 14 variables
\$ ï	: int	[1:1000] 0 1 2 3 4 5 6 7 8 9
\$ country	: chr	[1:1000] "Italy" "Portugal" "US" "US"
<pre>\$ description</pre>	: chr	[1:1000] "Aromas include tropical fruit, broom, brimstone and
<pre>\$ designation</pre>	: chr	[1:1000] "Vulkâ\210šâ\200 Bianco" "Avidagos" "" "Reserve Late
\$ points	: int	[1:1000] 87 87 87 87 87 87 87 87 87 87
\$ price	: int	[1:1000] NA 15 14 13 65 15 16 24 12 27
<pre>\$ province</pre>	: chr	[1:1000] "Sicily & Sardinia" "Douro" "Oregon" "Michigan"
<pre>\$ region_1</pre>	: chr	[1:1000] "Etna" "" "Willamette Valley" "Lake Michigan Shore"
\$ region_2	: chr	[1:1000] "" "" "Willamette Valley" ""
<pre>\$ taster_name</pre>	: chr	[1:1000] "Kerin Oâ\200šÅ"Å´Keefe" "Roger Voss" "Paul Gregutt"
<pre>\$ taster_twitter_handle</pre>	: chr	[1:1000] "@kerinokeefe" "@vossroger" "@paulgwineA-â\200 " ""
\$ title	: chr	[1:1000] "Nicosia 2013 Vulkâ\210šâ\200 Bianco (Etna)" "Quint
<pre>\$ variety</pre>	: chr	[1:1000] "White Blend" "Portuguese Red" "Pinot Gris" "Riesling
<pre>\$ winery</pre>	: chr	[1:1000] "Nicosia" "Quinta dos Avidagos" "Rainstorm" "St. Juli

It is defined with the country where the wine has been tasted, with the description according to the taste of the wine like Tropical fruit, broom, brimstone, and dried herb are among the aromas. The tongue is understated, with unripened apple, citrus, and dry sage accented by crisp acidity. Moreover, the designation as the name of the wine like Levele, Cignale, Rosado, with we can see points which help in defining the rating of the wine as per the rating price of the wine will be available in the dataset. We can see the availability of wine in various countries. This dataset will give an understanding of which province,

region (reigion_1 and region_2) the wine is available depending on the consumption of the wine in a particular country. Apart from this dataset gives insights about the taste of the wine with taster_name and information about the social media site where they handle the wine taste follower on Twitter. Likewise, this dataset defines the variety of wines available in various regions and provinces around different parts of the country. However, the most interesting thing we can retrieve the production details of the wine with winery details available in the dataset. Like sparkling blend is available in Unites States (US) which is available in Iron Horse winery.

Wine_ Tasting data has various 1000 rows and 14 columns which contain 'id' as an integer (int), there are many character variables (chr) like, description, designation, province, region_1, region_2, taster_name, taster_twitter_handle, country title, variety, and winery are defined as a character. Whereas Point and Price is a continuous variable is defined as integer(int).

Purpose of Dataset:

To get an understanding of the wine price in different countries with the specific brand and availability as per the rating (points) of the dataset. This Exploratory data analysis helps in visualizing the statistics through, bar scatterplot, boxplot, histogram, chart, and analysis of the dataset in the tabular form.

Task 1: Dataset Imported in R: In this task wine _tasting contains 14 variables and 1000 observations which help in analyzing and visualizing the data in meaningful data available in the dataset.

1 4	□ Y Fit	r							Q,			Q
ī.,	country	description	designation	points	price	province	region_1	region_2	taster_name	title	variety	winery
	D Italy	Aromas include tropical fluit, broom, brimstone and dried h	VulkᰚဠBianco	8	7 NA	Sicily & Sardinia	Etna		Kerin Oštšá, á Keefe	Nicosia 2013 Vulká*šá€ Bianco (Etna)	White Blend	Nicosia
	1 Portugal	This is ripe and fruity, a wine that is smooth while still struct	Avidagos	8	7 15	Douro			Roger Voss	Quinta dos Avidagos 2011 Avidagos Red (Douro)	Portuguese Red	Quinta dos Avidagos
	2 US	Tart and snappy, the flavors of lime flesh and rind dominate		8	7 14	Oregon	Willamette Valley	Wilamette Valley	Paul Gregutt	Rainstorm 2013 Pinot Gris (Willamette Valley)	Pinot Gris	Rainstorm
	3 US	Pineapple rind, lemon pith and orange biossom start off the	Reserve Late Harvest	8	7 13	Michigan	Lake Michigan Shore		Alexander Peartree	St. Julian 2013 Reserve Late Harvest Riesling (Lake Michigan	Riesling	St. Julian
-	4 US	Much like the regular bottling from 2012, this comes across	Vintner's Reserve Wild Child Block		65	Oregon	Williamette Valley	Williamette Valley	Paul Gregutt	Sweet Cheeks 2012 Vintner's Reserve Wild Child Block Pinot	Pinot Noir	Sweet Cheeks
	5 Spain	Blackberry and raspberry aromas show a typical Navarran w	Ars In Vitro	8	7 15	Northern Spain	Navarra		Michael Schachner	Tandem 2011 Ars in Vitro Tempranillo-Meriot (Navarra)	Tempranilio-Meriot	Tandem
	6 Italy	Here's a bright, informal red that opens with aromas of can	Beisito	8	7 16	Sicily & Sardinia	Vittoria		Kerin OátšÁ, Á Keefe	Terre di Giurfo 2013 Beisito Frappato (Vittoria)	Frappato	Terre di Giurfo
	7 France	This dry and restrained wine offers spice in profusion. Balan	•	8	7 24	Aisace	Alsace		Roger Voss	Trimbach 2012 Gewurztraminer (Alsace)	Gewä ⁻ šÅ ^o rztraminer	Trimbach
-	8 Germany	Savory dried thyme notes accent sunnier flavors of preserve	Shine	8	7 12	Rheinhessen			Anna Lee C. Iljima	Heinz Eifel 2013 Shine Gewä [*] šÄ ⁿ rztraminer (Rheinhessen)	Gewä'śÄ°rztraminer	Heinz Eifel
-	9 France	This has great depth of flavor with its fresh apple and pear f	Les Natures	8	7 27	Alsace	Alsace		Roger Voss	Jean-Baptiste Adam 2012 Les Natures Pinot Gris (Alsace)	Pinot Gris	Jean-Baptiste Adam
10	0 US	Soft, supple plum envelopes an oaky structure in this Caber	Mountain Cuvá'šÁ®e	8	7 15	California	Napa Valley	Napa	Virginie Boone	Kirkland Signature 2011 Mountain Cuvá*šÁ®e Cabernet Sau	Cabernet Sauvignon	Kirkland Signature
- 11	1 France	This is a dry wine, very spicy, with a tight, taut texture and st		8	7 30	Alsace	Alsace		Roger Voss	Leon Beyer 2012 Gewurztraminer (Alsace)	Gewä [*] šÅ ^a rztraminer	Leon Beyer
12	2 US	Slightly reduced, this wine offers a chalky, tannic backbone t		8	7 34	California	Alexander Valley	Sonoma	Virginie Boone	Louis M. Martini 2012 Cabernet Sauvignon (Alexander Valley)	Cabernet Sauvignon	Louis M. Martini
10	3 Italy	This is dominated by oak and oak-driven aromas that includ	Rosso	8	7 NA	Sicily & Sardinia	Etna		Kerin DátšÁ, Á Keete	Masseria Setteporte 2012 Rosso (Etna)	Nerello Mascalese	Masseria Setteporte
14	4 US	Building on 150 years and six generations of winemaking tr		8	7 12	California	Central Coast	Central Coast	Matt Kettmann	Mirassou 2012 Chardonnay (Central Coast)	Chardonnay	Mirassou
15	5 Germany	Zesty orange peels and apple notes abound in this sprightly	Devon	8	7 24	Mosel			Anna Lee C. Ijima	Richard Bá'śá' cking 2013 Devon Riesling (Mosel)	Riesling	Richard 8á°šá°,cking
16	6 Argentina	Baked plum, molasses, balsamic vinegar and cheesy oak aro	Felix	8	7 30	Other	Cafayate		Michael Schachner	Felix Lavaque 2010 Felix Malbec (Cafayate)	Melbec	Felix Lavaque
17	7 Argentina	Raw black-cherry aromas are direct and simple but good. Th	Winemaker Selection	8	7 13	Mendoza Province	Mendoza		Michael Schachner	Gaucho Andino 2011 Winemaker Selection Malbec (Mendoz	Malbec	Gaucho Andino
10	8 Spain	Desiccated blackberry, leather, charred wood and mint aro	Vendimia Seleccionada Finca Valdelayegua Single Vineyard	8	7 28	Northern Spain	Ribera del Duero		Michael Schachner	Pradorey 2010 Vendimia Seleccionada Finca Valdelayegua Si	Tempranilio Biend	Pradorey
11	9 US	Red fruit aromas pervade on the nose, with cigar box and m		8	7 32	Virginia	Virginia		Alexander Peartree	Quiă SÁ®vremont 2012 Meritage (Virginia)	Meritage	Quiá'šÁ®vremont
2	D US	Ripe aromas of dark berries mingle with ample notes of blac	Vin de Malson	8	7 23	Virginia	Virginia		Alexander Peartree	Quiă SÁ Evremont 2012 Vin de Maison Red (Virginia)	Red Blend	Quíá'šÁ©vremont
2	1 US	A sleek mix of tart berry, stem and herb, along with a hint of		8	7 20	Oregon	Oregon	Oregon Other	Paul Gregutt	Acrobat 2013 Pinot Noir (Oregon)	Pinot Noir	Acrobat
z	2 Italy	Delicate aromas recall white flower and citrus. The palate off	Ficilgno	8	7 19	Sicily & Sardinia	Sicilia		Kerin OátšÁ, Á Keete	Baglio di Pianetto 2007 Ficiligno White (Sicila)	White Blend	Baglio di Planetto
2	3 US	This wine from the Geneseo district offers aromas of sour pl	Signature Selection		7 22	California	Paso Robies	Central Coast	Matt Kettmann	Blanchi 2011 Signature Selection Meriot (Paso Robles)	Meriot	Blanchi
2	4 Italy	Aromas of prune, blackcurrant, toast and oak carry through	Aynat	8	35	Sicily & Sardinia	Sidila		Kerin OátšÁ, Á Keefe	Canicattá ŠÁ 2009 Aynat Nero d'Avola (Sicila)	Nero d'Avola	Canicattá'šÁ'
2	s US	Oak and earth intermingle around robust aromas of wet for	King Ridge Vineyard		65	California	Sonoma Coast	Sonoma	Virginie Boone	Castello di Amorosa 2011 King Ridge Vineyard Pinot Noir (S	Pinot Neir	Castello di Amorosa
2	6 Italy	Pretty aromas of yellow flower and stone fruit lead the nose	Dalla	8	7 13	Sicily & Sardinia	Terre Siciliane		Kerin OSKSÅ, Ä'Keefe	Stemmari 2013 Dalla White (Terre Siciliane)	White Blend	Stemmari
z	7 Itely	Aromas recall ripe dark berry, toast and a whiff of cake spice		8	7 10	Sicily & Sardinia	Terre Sicilane		Kerin OátšÁ, Á Keefe	Stemmari 2013 Nero d'Avola (Terre Siciliane)	Nero d'Avola	Stemmari
25	B Italy	Aromas suggest mature berry, scorched earth, animal, toast	Mascaria Barricato	8	7 17	Sicily & Sardinia	Cerasuolo di Vittoria		Kerin Oátšá, ÁlKeefe	Terre di Giurfo 2011 Mascaria Barricato (Cerasuolo di Vittoria)	Red Blend	Terre di Giurfo
25	9 US	Clarksburg is becoming a haven for Chenin Blanc in Californ		8	5 16	California	Clarksburg	Central Valley	Virginie Boone	Clarksburg Wine Company 2010 Chenin Blanc (Clarksburg)	Chenin Blanc	Clarksburg Wine Company
3	0 France	Red cherry fruit comes laced with light tannins, giving this b	Nouveau	8	5 NA	Beaujolais	Beaujolais-Villages		Roger Voss	Domaine de la Madone 2012 Nouveau (Beaujolais-Villages)	Gamay	Domaine de la Madone
3	1 Italy	Meriot and Nero d'Avoia form the base for this easy red win	Calaná"šÁ"ca Nero d'Avola-Merlot	8	S NA	Sicily & Sardinia	Sicilia			Duca di Salaparuta 2010 Calană SĂ Ca Nero d'Avola-Meriot	Red Blend	Duca di Salaparuta
_												

Task 2: Data cleaning: Create Duplicate Dataset to fill missing value with NA to make the data in an evaluation formate.

wine_new	100	0 obs. of 14 variables
\$ ï	: int	0 1 2 3 4 5 6 7 8 9
\$ country	: chr	"Italy" "Portugal" "US" "US"
<pre>\$ description</pre>	: chr	"Aromas include tropical fruit, broom, brimstone and dried herb. The pa
<pre>\$ designation</pre>	: chr	"Vulkâ\210šâ\200 Bianco" "Avidagos" NA "Reserve Late Harvest"
\$ points	: int	87 87 87 87 87 87 87 87 87 87 87
\$ price	: int	NA 15 14 13 65 15 16 24 12 27
\$ province	: chr	"Sicily & Sardinia" "Douro" "Oregon" "Michigan"
\$ region_1	: chr	"Etna" NA "Willamette Valley" "Lake Michigan Shore"
\$ region_2		NA NA "Willamette Valley" NA
<pre>\$ taster_name</pre>	: chr	"Kerin Oâ\200šÅ"Å´Keefe"•"Roger Voss" "Paul Gregutt" "Alexander Peartre
<pre>\$ taster_twitter_handle</pre>	e: chr	"@kerinokeefe" "@vossroger" "@paulgwineŬâ\200 " NA
\$ title	: chr	"Nicosia 2013 Vulkâ\210šâ\200 Bianco (Etna)" "Quinta dos Avidagos 201
\$ variety	: chr	"White Blend" "Portuguese Red" "Pinot Gris" "Riesling"
\$ winery	: chr	"Nicosia" "Quinta dos Avidagos" "Rainstorm" "St. Julian"

^ ī		country	description	designation	points	price	province	region_1 ÷	region_2	taster_name ÷	taster_twitter_handle
1	0	Italy	Aromas include tropical fruit, broom, brimstone and dried h	VulkᰚဠBlanco	87	N/	Sicily & Sardinia	Etna	NA	Kerin OယÃ,Ã′Keefe	@kerinokeefe
2	- 1	Portugal	This is ripe and fruity, a wine that is smooth while still struct	Avidagos	87	15	Douro	NA	NA	Roger Voss	@vossroger
3	2	US	Tart and snappy, the flavors of lime flesh and rind dominate	NA .	87	14	Oregon	Willamette Valley	Willamette Valley	Paul Gregutt	@paulgwineÂ⇒â€
4	3	US	Pineapple rind, lemon pith and orange blossom start off the	Reserve Late Harvest	87	13	Michigan	Lake Michigan Shore	NA	Alexander Peartree	NA
5	4	US	Much like the regular bottling from 2012, this comes across	Vintner's Reserve Wild Child Block	87	65	Oregon	Willamette Valley	Willamette Valley	Paul Gregutt	@paulgwine¬â€
6	5	Spain	Blackberry and raspberry aromas show a typical Navarran w	Ars In Vitro	87	15	Northern Spain	Navarra	NA	Michael Schachner	@wineschach
7	6	Italy	Here's a bright, informal red that opens with aromas of can	Belsito	87	16	Sicily & Sardinia	Vittoria	NA	Kerin O'Ã,Ã′Keefe	@kerinokeefe
8	7	France	This dry and restrained wine offers spice in profusion. Balan	NA	87	24	Alsace	Alsace	NA	Roger Voss	@vossroger
9	8	Germany	Savory dried thyme notes accent sunnier flavors of preserve	Shine	87	12	Rheinhessen	NA	NA	Anna Lee C. Iijima	NA
10	9	France	This has great depth of flavor with its fresh apple and pear f	Les Natures	87	27	Alsace	Alsace	NA	Roger Voss	@vossroger
11	10	US	Soft, supple plum envelopes an oaky structure in this Caber	Mountain Cuvă^šÂ©e	87	19	California	Napa Valley	Napa	Virginie Boone	@vboone
12	11	France	This is a dry wine, very spicy, with a tight, taut texture and st	NA .	87	30	Alsace	Alsace	NA	Roger Voss	@vossroger
13	12	US	Slightly reduced, this wine offers a chalky, tannic backbone t	NA .	87	34	California	Alexander Valley	Sonoma	Virginie Boone	@vboone
14	13	Italy	This is dominated by oak and oak-driven aromas that includ	Rosso	87	N/	Sicily & Sardinia	Etna	NA	Kerin O'Ã,Ã′Keefe	@kerinokeefe
15	14	US	Building on 150 years and six generations of winemaking tr	NA .	87	12	California	Central Coast	Central Coast	Matt Kettmann	@mattkettmann
16	15	Germany	Zesty orange peels and apple notes abound in this sprightly	Devon	87	24	Mosel	NA	NA	Anna Lee C. Iijima	NA
17	16	Argentina	Baked plum, molasses, balsamic vinegar and cheesy oak aro	Felix	87	30	Other	Cafayate	NA	Michael Schachner	@wineschach
18	17	Argentina	Raw black-cherry aromas are direct and simple but good. Th	Winemaker Selection	87	13	Mendoza Province	Mendoza	NA	Michael Schachner	@wineschach
19	18	Spain	Desiccated blackberry, leather, charred wood and mint aro	Vendimia Seleccionada Finca Valdelayegua Single Vineyard	87	21	Northern Spain	Ribera del Duero	NA	Michael Schachner	@wineschach
20	19	US	Red fruit aromas pervade on the nose, with cigar box and m	NA .	87	32	Virginia	Virginia	NA	Alexander Peartree	NA
21	20	US	Ripe aromas of dark berries mingle with ample notes of blac	Vin de Maison	87	2	Virginia	Virginia	NA	Alexander Peartree	NA
22	21	US	A sleek mix of tart berry, stem and herb, along with a hint of	NA .	87	20	Oregon	Oregon	Oregon Other	Paul Gregutt	@paulgwineÅ ¬â€
23	22	Italy	Delicate aromas recall white flower and citrus. The palate off	Ficiligno	87	19	Sicily & Sardinia	Sicilia	NA	Kerin OယÃ,Ã′Keefe	@kerinokeefe
24	23	US	This wine from the Geneseo district offers aromas of sour pl	Signature Selection	87	2	California	Paso Robles	Central Coast	Matt Kettmann	@mattkettmann
25	24	Italy	Aromas of prune, blackcurrant, toast and oak carry through	Aynat	87	35	Sicily & Sardinia	Sicilia	NA	Kerin O'Ã,Ã′Keefe	@kerinokeefe
26	25	US	Oak and earth intermingle around robust aromas of wet for	King Ridge Vineyard	87	69	California	Sonoma Coast	Sonoma	Virginie Boone	@vboone
27	26	Italy	Pretty aromas of yellow flower and stone fruit lead the nose	Dalila	87	13	Sicily & Sardinia	Terre Siciliane	NA	Kerin O'Ã,Ã′Keefe	@kerinokeefe
28	27	Italy	Aromas recall ripe dark berry, toast and a whiff of cake spice	NA .	87	10	Sicily & Sardinia	Terre Siciliane	NA	Kerin O'Ã,Ã′Keefe	@kerinokeefe
29	28	Italy	Aromas suggest mature berry, scorched earth, animal, toast	Mascaria Barricato	87	17	Sicily & Sardinia	Cerasuolo di Vittoria	NA	Kerin Oâ€ŝÃ,Ã′Keefe	@kerinokeefe
30	29	US	Clarksburg is becoming a haven for Chenin Blanc in Californ	NA	86	16	California	Clarksburg	Central Valley	Virginie Boone	@vboone
31	30	France	Red cherry fruit comes laced with light tannins, giving this b	Nouveau	86	N/	Beaujolais	Beaujolais-Villages	NA	Roger Voss	@vossroger
32	31	Italy	Meriot and Nero d'Avola form the base for this easy red win	Calanâ*šÂ"ca Nero d'Avola-Merlot	86	N/	Sicily & Sardinia	Sicilia	NA	NA	NA
22				and the second of				er or			

```
> wine_new <- wine_data
> wine_new[wine_new == ""] <- NA</pre>
> wine_new
   i.. country
0 Italy
1 Portugal
    2
3
                 US
                 US
5
                US
    5
          Spain
   6
7
            İtaly
          France
8
9
    8 Germany
10 9
11 10
12 11
          France
          France
```

Task 3: In this task change the name of the data variable available in dataset

```
> wine_aly_rename1 <- rename_with(wine_new, toupper)
   Ϊ..
        COUNTRY
          Italy
       Portugal
    1
4
    3
             US
    4
             US
6
          Spain
     6
          Italy
8
         France
9
    8
        Germany
10
   9
         France
11 10
             US
12
   11
          France
13 12
             US
14 13
          Italy
15
   14
16 15
        Germany
17 16 Argentina
18 17 Argentina
```

^ ï	. ÷	COUNTRY [‡]	DESCRIPTION	DESIGNATION	POINTS [‡]	PRICE [‡]	PROVINCE [‡]	REGION_
1	0	Italy	Aromas include tropical fruit, broom, brimstone and dried h	Vulkâ^šã€ Bianco	87	NA	Sicily & Sardinia	Etna
2	1	Portugal	This is ripe and fruity, a wine that is smooth while still struct	Avidagos	87	15	Douro	NA
3	2	US	Tart and snappy, the flavors of lime flesh and rind dominate	NA	87	14	Oregon	Willamett
4	3	US	Pineapple rind, lemon pith and orange blossom start off the	Reserve Late Harvest	87	13	Michigan	Lake Mich
5	4	US	Much like the regular bottling from 2012, this comes across	Vintner's Reserve Wild Child Block	87	65	Oregon	Willamett
6	5	Spain	Blackberry and raspberry aromas show a typical Navarran w	Ars In Vitro	87	15	Northern Spain	Navarra
7	6	Italy	Here's a bright, informal red that opens with aromas of can	Belsito	87	16	Sicily & Sardinia	Vittoria
8	7	France	This dry and restrained wine offers spice in profusion. Balan	NA	87	24	Alsace	Alsace
9	8	Germany	Savory dried thyme notes accent sunnier flavors of preserve	Shine	87	12	Rheinhessen	NA
10	9	France	This has great depth of flavor with its fresh apple and pear f	Les Natures	87	27	Alsace	Alsace
11	10	US	Soft, supple plum envelopes an oaky structure in this Caber	Mountain Cuvâ^šÂ©e	87	19	California	Napa Vall
12	11	France	This is a dry wine, very spicy, with a tight, taut texture and st	NA	87	30	Alsace	Alsace

Task 4: In this task dropped variable to do our analysis precise

> #Drop
> wine_aly_drop <- subset(wine_aly_rename1, select = -c (DESCRIPTION, TITLE, TASTER_NAME, TASTER_TWITTER_HANDLE, WINERY, REGION_1, REGION_2))

1 2 3 4 5 6 7 8 9	1 0 1 2 3 4 5 6 7 8	COUNTRY Italy Portugal US US Spain Italy France Germany	Vintner's Res	vulkâî Reserve L erve Wild (A)		PROVINCE sicily & Sardinia Douro Oregon Michigan Oregon Northern Spain Sicily & Sardinia Alsace Rheinhessen	
^	ï ÷	COUNTRY [‡]	DESIGNATION	POINTS [‡]	PRICE [‡]	PROVINCE	VARIETY
1	0	Italy	Vulkâ^šå€ Bianco	87	NA	Sicily & Sardinia	White Blend
2	1	Portugal	Avidagos	87	15	Douro	Portuguese Red
3	2	US	NA	87	14	Oregon	Pinot Gris
4	3	US	Reserve Late Harvest	87	13	Michigan	Riesling
5	4	US	Vintner's Reserve Wild Child Block	87	65	Oregon	Pinot Noir
6	5	Spain	Ars In Vitro	87	15	Northern Spain	Tempranillo-Merlot
7	6	Italy	Belsito	87	16	Sicily & Sardinia	Frappato
8	7	France	NA	87	24	Alsace	Gewâ^šÂ°rztraminer
9	8	Germany	Shine	87	12	Rheinhessen	Gewâ^šÂ°rztraminer
10	9	France	Les Natures	87	27	Alsace	Pinot Gris
11	10	US	Mountain Cuvâ^šÂ®e	87	19	California	Cabernet Sauvignon
12	11	France	NA	87	30	Alsace	Gewâ^šÂ°rztraminer
13	12	US	NA	87	34	California	Cabernet Sauvignon
14	13	Italy	Rosso	87	NA	Sicily & Sardinia	Nerello Mascalese
15	14	US	NA	87	12	California	Chardonnay
16	15	Germany	Devon	87	24	Mosel	Riesling
17	16	Argentina	Felix	87	30	Other	Malbec
18	17	Argentina	Winemaker Selection	87	13	Mendoza Province	Malbec
19	18	Spain	Vendimia Seleccionada Finca Valdelayegua Single Vineyard	87	28	Northern Spain	Tempranillo Blend
20	19	HS	NΔ	87	32	Virninia	Meritane

Task 4: In this task structure of the dataset after dropping a few variables.

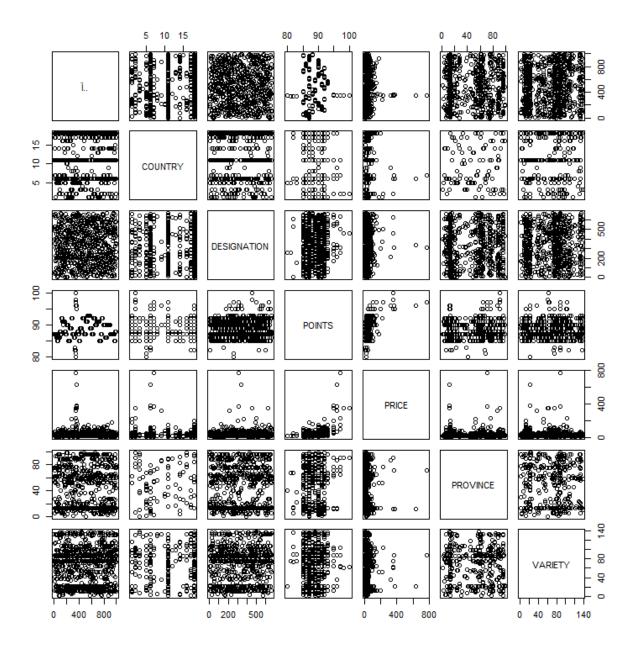
```
> str(wine_aly_1)
'data.frame': 1000 obs. of 7 variables:
$\tilde{I}\) : int 0 1 2 3 4 5 6 7 8 9 ...
$\tilde{S}\) COUNTRY : chr "Italy" "Portugal" "US" "US" ...
$\tilde{S}\) DESIGNATION: chr "Vulk\(\frac{A}{2}\) 10\(\frac{A}{2}\) 200 Bianco" "Avidagos" NA "Reserve Late Harvest" ...
$\tilde{P}\) POINTS : int 87 87 87 87 87 87 87 87 87 87 ...
$\tilde{P}\) PRICE : int NA 15 14 13 65 15 16 24 12 27 ...
$\tilde{P}\) PROVINCE : chr "Sicily & Sardinia" "Douro" "Oregon" "Michigan" ...
$\tilde{V}\) VARIETY : chr "White Blend" "Portuguese Red" "Pinot Gris" "Riesling" ...
```

Task 5: In this task as the point has min. value 80 and max. value 100 with the mean and median 88, 88.58 respectively. Moreover, the minimum price of the wine is 7, and the maximum price is 775

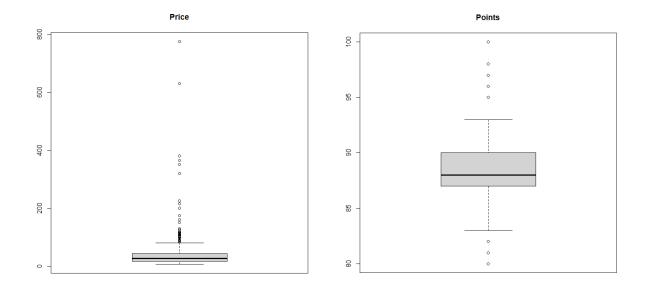
```
> summary(wine_aly_1)
Ϊ..
Min. : 0.0
                       COUNTRY
                                                                         POINTS
                                                                                              PRICE
                                                                                         Min. : 7.00
1st Qu.: 17.00
                                                                    Min.
                    Length:1000
                                            Length:1000
Class :character
                                                                    Min. : 80.00
1st Qu.: 87.00
                                                                                                               Length:1000
                                                                                                                                      Length:1000
1st Qu.:249.8
                     Class :character
                                                                                                               Class :character
                                                                                                                                      Class :character
                                                                                         Median : 27.00
Mean : 37.35
Median :499.5
Mean :499.5
                    Mode :character Mode :character
                                                                    Median : 88.00
Mean : 88.58
                                                                                                               Mode :character
                                                                                                                                      Mode :character
                                                                    3rd Qu.: 90.00
Max. :100.00
                                                                                         3rd Qu.: 43.50
Max. :775.00
 3rd Qu.:749.2
         :999.0
мах.
                                                                                                  :57
```

Task 6: In this task before analyzing the facts about the dataset- glimpse()

Task 7: In this task to create the scatterplot for understanding the check the correlation between the variables of the wine_tasting dataset. We can see the correlation between price and points. Let's discuss this



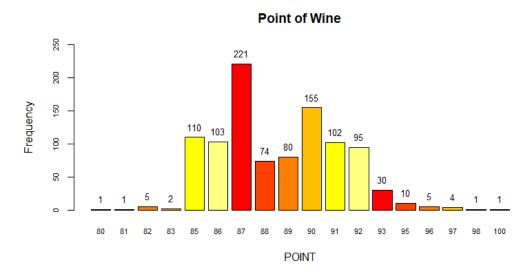
Task 8: In this task, the boxplot shows the outlier of the price variable, which shows around 84 to 94 points the price of the wine is below 100, and above the price, boxplot shows the outlier which above price 400.



Task 9: In this task shows the price of the wine with frequency



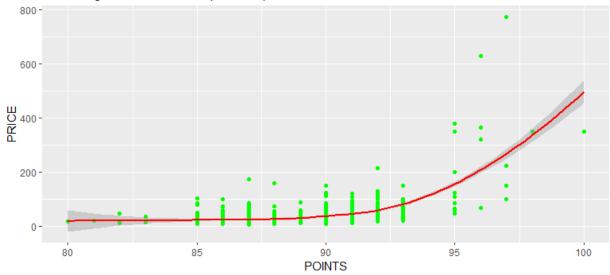
Task 10: In this task points if the wine shows at point 87 maximum number of wine are available i.e. 221



Task 13: In this task we can see the relationaship between the wine price and the rating of the wine, higher the prices in ascending order with very fewer reviews(points) ggplot()

Wine Ratings: Price vs. Points

Wine ratings and their relationship with the price of wine



Data published by WineEnthusiast

Summary:

In this wine_tasting dataset, prices depend on the points of the wine higher the review higher the price. All the variables in this data provide analysis above the graphs. Explain what the tables and visualizations tell you about the data.

This research aims to study relationships between 14 variables, as well as find solutions for visualizations, distributions, and outliers in a selected wine taste data set, using exploratory data analysis (EDA) methodologies.

This dataset has 1000 observations and 14 variables. This dataset will aid in providing insights into the wine taste by including the variables designation, description, flavor, price, province, region 1, region 2, taster name, tastera twitter handle, title, variety, and winery. Here, we'll demonstrate how to perform the statistical analysis while using the R.

There are many character variables (chr) such as country, description, designation, province, region 1, region 2, taster name, taster twitter handle, title, variety, and winery that are defined as a character. Wine_ Tasting data has various 1000 rows and 14 columns that contain 'id' as an integer (int). Point and Price, on the other hand, is a continuous variable that is defined as an integer (int).

It is described by the country in which the wine was tasted, as well as a description of the wine's flavor, such as tropical fruit, broom, brimstone, and dry herb, among other scents. Unripened apple, citrus, and dry sage are enhanced by crisp acidity on the tongue. Furthermore, the designation as the name of the wine, such as Levele, Cignale, or Rosado, will be available in the dataset, along with points that assist in determining the rating of the wine based on the rating price of the wine.

The availability of wine in many countries may be seen. This dataset will show which provinces and regions (reigion 1 and region 2) the wine is available in, based on the wine's consumption in a given country. Apart from that, this dataset contains information about the wine taster name and the social media site where they manage the wine taste follower on Twitter.

Reference:

[1]Create a new ggplot

https://ggplot2.tidyverse.org/reference/ggplot.html

[2] Exploring, cleaning, and analysing data in R - YouTube

A replay of a non-technical livestream that walked through how to explore, clean, and analyse data in R, using the 'starwars' dataset that is built into the ...

https://www.youtube.com/watch?v=Ap1Q2fkqO_I

[3] How To Make Frequency Table in R - Programming R Tutorials

https://www.programmingr.com/statistics/frequency-table/

[4]Wine tasting

https://en.wikipedia.org/wiki/Wine_tasting#:~:text=There%20are%20five%20basic%20steps,expressiveness%2C%20complexity%2C%20and%20connectedness.

[5]R in Action

https://www.google.com/books/edition/R_in_Action/1TkzEAAAQBAJ?hl=en&gbpv=1&printsec=frontcover

Appendix: R code

```
install.packages("tidyverse")
install.packages("dplyr")
install.packages("plyr")
install.packages("pipisplay")
install.packages("gmodels")
install.packages("gmodels")
 library(psych)
library(epiDisplay)
library(gmodels)
library(ggplot2)
wine_data <- read.csv("C:\\Abhinav _ NEU BOSTON\\ALY6010 Probability Theory & Intro Statistics\\Module 1 Submission\\Wine_tasting_20220222.csv")
wine_data
wine_new <- wine_data
wine_new == ""] <- NA
wine_new == ""] <- NA
wine_aly_rename1 <- rename_with(wine_new, toupper)
wine_alv_rename1
wine_aly_drop <- subset(wine_aly_rename1, select = -c (DESCRIPTION, TITLE, TASTER_NAME, TASTER_TWITTER_HANDLE, WINERY, REGION_1, REGION_2))
wine_aly_drop
wine_aly_1 <- wine_aly_drop
#Structure of the dataset str(wine_aly_1)
#Summary of the dataset
summary(wine_aly_1)
#Glimpse of the dataset
glimpse(wine_aly_1)
# BOX PLOT
boxplot(wine_aly_1$PRICE)
# Through ggplot find highest price of provices
wine_aly_3 <- wine_aly_drop[!(is.na(wine_aly_drop$PROVINCE) | wine_aly_drop$PROVINCE==""), ]</pre>
wine_alv_3
province_reviews_5 <- wine_aly_3 %>% group_by(PROVINCE) %>% filter(n() >= 5)
avg_price_5 <- province_reviews_5 %>% group_by(PROVINCE) %>% drop_na() %>% summarize(mean_price = mean(PRICE))
arrange(avg_price_5, -mean_price)
ggplot(data = wine_aly_3) + geom_point(mapping=aes(x=POINTS,y=PRICE),color="Green") +
geom_smooth(mapping=aes(x=POINTS,y=PRICE),color="Red") +
labs(title="wine Ratings: Price vs. Points",
subtitle="wine ratings and their relationship with the price of wine", caption="Data published by WineEnthusiast")
```