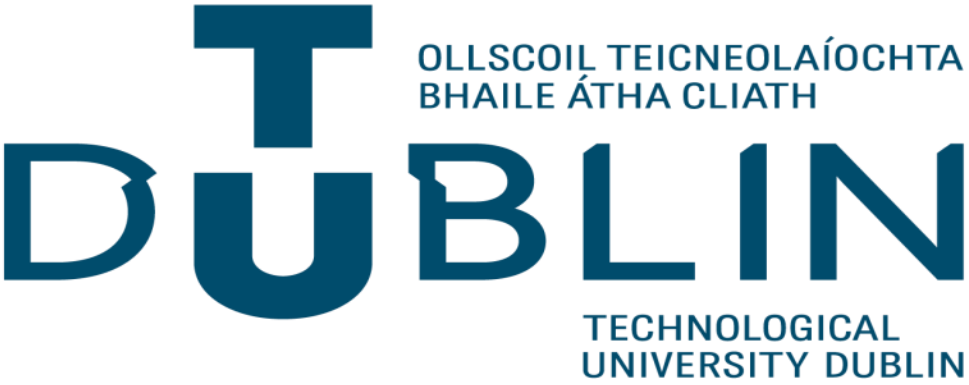

DATA MINING ASSIGNMENT 2

ANALYSIS AND MODELLING OF PHARMACEUTICAL DATA FOR OPINION MINING AND DRUG CHARACTERISTICS

PYTHON CODE APPENDIX

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Subject Code	DATA 9900
Course	TU059
Stream	Data Science (DS)
Year	First year
Assignment	Assignment 2
Deadline	8th Jan 2023

This document contains only python code snippets. For rest, please refer report.

Python code related to importing and cleaning the training dataset:

[1] Reading the excel file

```
# Reading the raw data from the file
raw_med_data = pd.read_csv('drugLib_raw/drugLibTrain_raw.tsv',sep='\t')
display(raw_med_data.head(5))
```

	Unnamed: 0	urlDrugName	rating	effectiveness	sideEffects	condition	benefitsReview	sideEffectsReview	commentsReview
0	2202	enalapril	4	Highly Effective	Mild Side Effects	management of congestive heart failure	slowed the progression of left ventricular dys...	cough, hypotension , proteinuria, impotence , ...	monitor blood pressure , weight and asses for ...
1	3117	ortho-tri-cyclen	1	Highly Effective	Severe Side Effects	birth prevention	Although this type of birth control has more c...	Heavy Cycle, Cramps, Hot Flashes, Fatigue, Lon...	I Hate This Birth Control, I Would Not Suggest...
2	1146	ponstel	10	Highly Effective	No Side Effects	menstrual cramps	I was used to having cramps so badly that they...	Heavier bleeding and clotting than normal.	I took 2 pills at the onset of my menstrual cr...
3	3947	prilosec	3	Marginally Effective	Mild Side Effects	acid reflux	The acid reflux went away for a few months aft...	Constipation, dry mouth and some mild dizzines...	I was given Prilosec prescription at a dose of...
4	1951	lyrica	2	Marginally Effective	Severe Side Effects	fibromyalgia	I think that the Lyrica was starting to help w...	I felt extremely drugged and dopey. Could not...	See above

[2] Checking for null values

```
# Check for null values in the dataset
raw_med_data.isna().sum()
#raw_med_data.isna().sum().sum()
```

```
Unnamed: 0            0
urlDrugName           0
rating                0
effectiveness         0
sideEffects           0
condition             1
benefitsReview        0
sideEffectsReview     2
commentsReview        8
dtype: int64
```

[3] Remove the reviews without alphanumeric characters

```
# Remove the reviews and conditions that don't have alphanumeric characters in them
med_cleaned = raw_med_data.drop(raw_med_data[~raw_med_data.commentsReview.str.contains('[a-zA-Z]',na=True)].index)
med_cleaned = med_cleaned.drop(med_cleaned[~med_cleaned.sideEffectsReview.str.contains('[a-zA-Z]',na=True)].index)
med_cleaned = med_cleaned.drop(med_cleaned[~med_cleaned.condition.str.contains('[a-zA-Z]',na=True)].index)

display(med_cleaned)
med_cleaned.shape
```

	Unnamed: 0	urlDrugName	rating	effectiveness	sideEffects	condition	benefitsReview	sideEffectsReview	commentsReview
0	2202	enalapril	4	Highly Effective	Mild Side Effects	management of congestive heart failure	slowed the progression of left ventricular dys...	cough, hypotension , proteinuria, impotence , ...	monitor blood pressure , weight and asses for ...
1	3117	ortho-tri-cyclen	1	Highly Effective	Severe Side Effects	birth prevention	Although this type of birth control has more c...	Heavy Cycle, Cramps, Hot Flashes, Fatigue, Lon...	I Hate This Birth Control, I Would Not Suggest...

[4] Rename columns and drop unnecessary columns

```
# Rename the columns
med_cleaned.rename(columns={'Unnamed: 0': 'Unnamed', 'urlDrugName': 'DrugName','rating':'Rating','effectiveness':'Effectiveness',
                            'sideEffects':'SideEffects','condition':'Condition','benefitsReview':'BenefitsReview',
                            'sideEffectsReview':'SideEffectsReview','commentsReview':'CommentsReview'}, inplace=True)
# Drop the column that is not described
med_cleaned = med_cleaned.drop('Unnamed', axis=1)
display(med_cleaned)
```

	DrugName	Rating	Effectiveness	SideEffects	Condition	BenefitsReview	SideEffectsReview	CommentsReview
0	enalapril	4	Highly Effective	Mild Side Effects	management of congestive heart failure	slowed the progression of left ventricular dys...	cough, hypotension , proteinuria, impotence , ...	monitor blood pressure , weight and asses for ...
1	ortho-tri-cyclen	1	Highly Effective	Severe Side Effects	birth prevention	Although this type of birth control has more c...	Heavy Cycle, Cramps, Hot Flashes, Fatigue, Lon...	I Hate This Birth Control, I Would Not Suggest...

Analysis screenshots

[1] Check unique number of drugs and conditions

```
# How many medicines are produced by the pharmaceutical company
med_cleaned['DrugName'].nunique()
```

```
502
```

```
med_cleaned['Condition'].nunique()
```

```
1422
```

[2] Distribution of effectiveness with pie chart

```
effectiveness_counts = med_cleaned.Effectiveness.value_counts()
effectiveness_counts_labels = effectiveness_counts.index.to_list()
print(effectiveness_counts)

fig = px.pie(med_cleaned, values= effectiveness_counts, names=effectiveness_counts_labels)
fig.show()
```

[3] Distribution of side effects with pie chart

```
side_effect_serious_counts = med_cleaned.SideEffects.value_counts()
side_effect_serious_counts_labels = side_effect_serious_counts.index.to_list()
print(side_effect_serious_counts)

fig = px.pie(med_cleaned, values = side_effect_serious_counts, names = side_effect_serious_counts_labels)
fig.show()
```

[4] How many drugs have an average rating of 10

```
# Average rating of all the drugs
avg_rating = med_cleaned.groupby(['DrugName']).mean()
display(avg_rating.sort_values('Rating', ascending = False).head(5))
# How many have an average rating of 10
avg_rating[avg_rating.Rating == 10].count()
```

[5] Name the medicines with 10 rating

```
rating_10 = avg_rating[avg_rating.Rating == 10]
rating_10 = rating_10.sort_values('DrugName')
rating_10_names = rating_10.index.to_list()

fig = go.Figure(data=[go.Table(header=dict(values=['Medicine Name', 'Medicine Name', 'Medicine Name', 'Medicine Name', 'Medicine
                cells=dict(values=[rating_10_names[0:10], rating_10_names[10:20], rating_10_names[20:30],
                                   rating_10_names[30:40], rating_10_names[40:50]]))
                      ])
fig.update_layout(
    title = "Names of medicines with an average rating of 10"
)
fig.show()
```

[6] Correlation between rating and effectiveness

Contingency table:

```
# Correlation between rating and effectiveness
# Does the rating depend on how effetive the medicine is
rat_effect_cont = pd.crosstab(index=med_cleaned['Effectiveness'], columns=med_cleaned['Rating'])
display(rat_effect_cont)
```

Chi square:

```
# Chi-sq test for rating vs
stat, p, dof, expected = chi2_contingency(rat_effect_cont)
print(stat)
print(p)
print(dof)
# Since more than 2 categories for the variable, cannot do fisher exact
```

Cramer v:

```
#Calculate cramer's v
# V = √(X2/n) / min(c-1, r-1)

cramer1_N = rat_effect_cont.sum().sum()
print(cramer1_N)
minimum_dimension = min(rat_effect_cont.shape)-1
print(minimum_dimension)

result = np.sqrt((stat/cramer1_N) / minimum_dimension)
print(result)

#There are 36 degrees of freedom
#For them, 0.5 shows a strong effect for cramer's v
```

[7] Which condition do we have a lot of reviews for:

```
#Which condition do we have a lot of reviews for

print(med_cleaned['Condition'].nunique())

condi_count = med_cleaned.groupby(['Condition']).count()

condi_count = condi_count.drop('Rating', axis=1)
condi_count = condi_count.drop('Effectiveness', axis=1)
condi_count = condi_count.drop('SideEffects', axis=1)
condi_count = condi_count.drop('BenefitsReview', axis=1)
condi_count = condi_count.drop('SideEffectsReview', axis=1)
condi_count = condi_count.drop('CommentsReview', axis=1)
condi_count.rename(columns={'DrugName': 'Count'}, inplace=True)

display(condi_count.sort_values('Count',ascending = False).head(10))
```

```
1422
```

Displaying top 10 as a bar chart:

```
# Displaying top 10 as a bar chart
top_10_condi = condi_count.sort_values('Count',ascending = False).head(10)
top_10_condi_names = top_10_condi.index.to_list()

fig = px.bar(top_10_condi, x=top_10_condi_names, y='Count', color = top_10_condi_names,
             title="Top 10 Conditions we have reviews for",
             labels=dict(x="Condition Name", Count="Count of Reviews", color="Condition"))

fig.show()
```

[8] Finding the best drugs for a condition

```
#Best rated drugs for a condition

best_med = med_cleaned

best_med = best_med.loc[best_med['Rating'].eq(best_med.groupby('Condition')['Rating'].transform('max')), :]
display(best_med.sort_values('Condition').head(10))
```

	DrugName	Rating	Effectiveness	SideEffects	Condition	BenefitsReview	SideEffectsReview	CommentsReview
703	tramadol	9	Highly Effective	Mild Side Effects	2 compressed discs in neck	Completely eliminated neck pain. Interestingly...	Slight euphoria, slight tiredness, but cannot ...	I've had 2 compressed discs in neck for over 5...
2792	chantix	10	Highly Effective	Mild Side Effects	20 year pack a day smoker	Definately helps control the urge to smoke. W...	Disrupted sleep is the only side effect so far...	2 pills a day as prescribed. I recommend it, ...

Studying the rating distribution of best drugs:

```
# Studying the rating distribution
best_meds_overall_dist = best_med.groupby(['Rating']).count()

#best_meds_overall_dist = best_meds_overall_dist.drop('Rating', axis=1)
best_meds_overall_dist = best_meds_overall_dist.drop('Effectiveness', axis=1)
best_meds_overall_dist = best_meds_overall_dist.drop('SideEffects', axis=1)
best_meds_overall_dist = best_meds_overall_dist.drop('BenefitsReview', axis=1)
best_meds_overall_dist = best_meds_overall_dist.drop('SideEffectsReview', axis=1)
best_meds_overall_dist = best_meds_overall_dist.drop('CommentsReview', axis=1)
best_meds_overall_dist = best_meds_overall_dist.drop('Condition', axis=1)
best_meds_overall_dist.rename(columns={'DrugName': 'Count'}, inplace=True)

print(best_meds_overall_dist)
best_meds_overall = best_meds_overall_dist.Count.to_list()
best_meds_overall_names = best_meds_overall_dist.index.to_list()

print(best_meds_overall_names)

fig = px.bar(best_med, x=best_meds_overall_names, y=best_meds_overall,
                title="Rating distribution for the best rated medicines",
                labels=dict(x="Rating", y="Count of Reviews", color=best_meds_overall_names))

fig.show()
```

Plot best medicine vs. severity:

```
best_med_top_10 = best_med.loc[best_med.Condition.isin(top_10_condi_names)]
#display(best_med_top_10)

fig = px.scatter(best_med_top_10, x = 'DrugName', y = 'SideEffects', size = 'Rating',
                color = 'Condition', title = "Best-rated Medicine vs. Side effect severity")
fig.show()

#From chart, all top conditions seem to have a 10 rated medicine
```

Plot best medicine vs. Effectiveness:

```
best_med_top_10 = best_med.loc[best_med.Condition.isin(top_10_condi_names)]
#display(best_med_top_10)

fig = px.scatter(best_med_top_10, x = 'DrugName', y = 'Effectiveness', size = 'Rating', color = 'Condition',
                title='Best-rated medicine vs. Effectiveness')
fig.show()

#From chart, all top conditions seem to have a 10 rated medicine
```

[9] Correlation: rating and side effects

Contingency table:

```
# Correlation between rating and side effects
# Does the rating depend on how bad the side effects are
rat_se_cont = pd.crosstab(index=med_cleaned['SideEffects'], columns=med_cleaned['Rating'])
display(rat_se_cont)
```

Chisq:

```
# Chi-sq test for rating vs side-effects
stat2, p2, dof2, expected2 = chi2_contingency(rat_se_cont)
print(stat2)
print(p2)
print(dof2)
# Since more than 2 categories for the variable, cannot do fisher exact
```

Cramer V:

```
#Calculate cramer's v
# V = √(X2/n) / min(c-1, r-1)

cramer2_N = rat_se_cont.sum().sum()
print(cramer2_N)
minimum_dimension2 = min(rat_se_cont.shape)-1
print(minimum_dimension2)

result2 = np.sqrt((stat2/cramer2_N) / minimum_dimension2)
print(result2)

#There are 36 degrees of freedom
#For them, 0.4 shows a strong effect for cramer's v
```

```
3099
4
0.4442284721296169
```

[10] Correlation: Effectiveness and side effects

Contingency table:

```
# Correlation between rating and side effects
# Does the rating depend on how bad the side effects are
eff_se_cont = pd.crosstab(index=med_cleaned['Effectiveness'], columns=med_cleaned['SideEffects'])
display(eff_se_cont)
```

Chisq:

```
# Chi-sq test for effectiveness vs sideeffects
stat3, p3, dof3, expected3 = chi2_contingency(eff_se_cont)
print(stat3)
print(p3)
print(dof3)
# Since more than 2 categories for the variable, cannot do fisher exact
```

Cramer V:

```
#Calculate cramer's v
# V = √(X2/n) / min(c-1, r-1)

cramer3_N = eff_se_cont.sum().sum()
print(cramer3_N)
minimum_dimension3 = min(eff_se_cont.shape)-1
print(minimum_dimension3)

result3 = np.sqrt((stat3/cramer3_N) / minimum_dimension3)
print(result3)

#There are 16 degrees of freedom
#For them, 0.2 shows a strong effect for cramer's v
```

```
3099
4
0.2330083640182357
```

[11] Analyse highly effective but extremely severe side effects

Filter medicines:

```
### What is people's opinion on highly effective medicine with extremely severe side effects

effect_and_severe = med_cleaned.loc[(med_cleaned.Effectiveness == "Highly Effective") &
                                    (med_cleaned.SideEffects == "Extremely Severe Side Effects")]
display(effect_and_severe.head(10))
```

Plot ratings as bar graph:

```
# Plot the reviews of people for the above records
# Using rating counts as a measure

e_and_s_counts = effect_and_severe.Rating.value_counts()
e_and_s_count_labels = e_and_s_counts.index.to_list()
print(e_and_s_counts)

fig = px.bar(med_cleaned, x=e_and_s_count_labels, y=e_and_s_counts,
             title="Opinions of people on Medicine which is Highly Effective but has Extremely Severe Side effects",
             labels=dict(x="Rating", y="No of reviews"))

fig.show()
```

Count of conditions in this category:

```
effect_and_severe.Condition.value_counts()
```

```
birth control                                    2
sore throat                                      1
contraception                                    1
crohns disease, psoriatic arthritis              1
seizures                                         1
acne                                             1
grand mal seizures                               1
skin wound/infection                             1
add                                              1
adhd                                             1
weightloss                                       1
cornea transplant rejection                      1
i kept getting pregnant. no more stairs.... please  1
hysterectomy                                     1
u/i                                              1
acid reflux, gerd                                1
cystic acne                                      1
controceptive; help with pmsd                    1
headaches                                        1
```

[12] Ineffective and extremely severe medicines:

Filter medicines:

```
ineffect_and_severe = med_cleaned.loc[(med_cleaned.Effectiveness == "Ineffective") &
                                      (med_cleaned.SideEffects == "Extremely Severe Side Effects")]
display(ineffect_and_severe)
```

Plot ratings as bar graphs:

```
# Plot the reviews of people for the above records
# Using rating counts as a measure

ie_and_s_counts = ineffect_and_severe.Rating.value_counts()
ie_and_s_count_labels = ie_and_s_counts.index.to_list()
print(ie_and_s_counts)

fig = px.bar(med_cleaned, x=ie_and_s_count_labels, y=ie_and_s_counts,
             title="Opinions of people on Medicine which is Ineffective and has Extremely Severe Side effects",
             labels=dict(x="Rating", y="No of reviews"))

fig.show()
```

Look for alternatives for such medicines:

```
# Look for alternatives for the above cases and then check the ratings on those.

#First take a list of the conditions from above
ie_and_s_conditions = ineffect_and_severe.Condition.to_list()
#print(ie_and_s_conditions)

alternatives = med_cleaned.loc[med_cleaned.Condition.isin(ie_and_s_conditions)]
alternatives = alternatives.loc[(alternatives.Effectiveness != 'Ineffective')
                                &(alternatives.SideEffects != 'Extremely Severe Side Effects')]
display(alternatives)
```

Plot ratings as bar graph:

```
#Plot the graph to study the ratings on the alternatives
alternative_counts = alternatives.Rating.value_counts()
alternative_counts_labels = alternative_counts.index.to_list()
print(alternative_counts)

fig = px.bar(med_cleaned, x=alternative_counts_labels, y=alternative_counts,
             title="Opinions of people on alternatives to medicine which is Ineffective and has Extremely Severe side effects
             labels=dict(x="Rating", y="No of reviews"))

fig.show()
```

[13] Adding sentiment column for ML:

```
med_cleaned['Sentiment'] = np.where(med_cleaned['Rating'] > 5, 'Positive', 'Negative')
display(med_cleaned.head(5))
```

Exporting the training data after cleaning:

```
# Exporting the excel
med_cleaned.to_csv("MedCleanedTrainingSet.csv", index = False)
```

Cleaning and exporting the test set:

Reading the data:

```
# Reading the raw data from the file
raw_med_test_data = pd.read_csv('drugLib_raw/drugLibTest_raw.tsv',sep='\t')
display(raw_med_test_data.head(5))
```

Null checks on data:

```
# Check for null values in the dataset
raw_med_test_data.isna().sum()
```

Remove non alpha numeric reviews:

```
# Remove the reviews and conditions that don't have alphanumeric characters in them
raw_med_test_data = raw_med_test_data.drop(raw_med_test_data[~raw_med_test_data.commentsReview.str.contains('[a-zA-Z]',na=True)].
raw_med_test_data = raw_med_test_data.drop(raw_med_test_data[~raw_med_test_data.sideEffectsReview.str.contains('[a-zA-Z]',na=True
raw_med_test_data = raw_med_test_data.drop(raw_med_test_data[~raw_med_test_data.condition.str.contains('[a-zA-Z]',na=True)].index

display(raw_med_test_data)
raw_med_test_data.shape
```

Rename columns/ Drop unnecessary:

```
# Rename the columns
raw_med_test_data.rename(columns={'Unnamed: 0': 'Unnamed', 'urlDrugName': 'DrugName','rating':'Rating','effectiveness':'Effective
                          'sideEffects':'SideEffects','condition':'Condition','benefitsReview':'BenefitsReview',
                          'sideEffectsReview':'SideEffectsReview','commentsReview':'CommentsReview'}, inplace=True)
# Drop the column that is not described
raw_med_test_data = raw_med_test_data.drop('Unnamed', axis=1)
display(raw_med_test_data)
```

Add sentiment column:

```
#Adding column for ML
raw_med_test_data['Sentiment'] = np.where(raw_med_test_data['Rating'] > 5, 'Positive', 'Negative')
display(raw_med_test_data.head(5))
```

Export:

```
#Export into excel
raw_med_test_data.to_csv("MedCleanedTestSet.csv", index = False)
```

Additional snippets for conclusions:

[1] Highly effective medicines with severe side effects rating distribution

Get such medicines:

```
### What is people's opinion on highly effective medicine with severe side effects

effect_and_severe_side = med_cleaned.loc[(med_cleaned.Effectiveness == "Highly Effective") &
                                         (med_cleaned.SideEffects == "Severe Side Effects")]
display(effect_and_severe_side.head(10))
```

Plot rating distribution as bar graph:

```
e_and_ss_counts = effect_and_severe_side.Rating.value_counts()
e_and_ss_count_labels = e_and_ss_counts.index.to_list()
print(e_and_ss_counts)

fig = px.bar(med_cleaned, x=e_and_ss_count_labels, y=e_and_ss_counts,
             title="Opinions of people on Medicine which is Highly Effective but has Severe Side effects",
             labels=dict(x="Rating", y="No of reviews"))

fig.show()
```

Conclusion:

This concludes the code appendix where the snippets of the python code are pasted. For full interpretation, plots and results please refer to the report in the respective sections.