Data cleaning

```
from google.colab import drive
drive.mount('/content/drive/')
→ Mounted at /content/drive/
import os
import pandas as pd
from matplotlib import pyplot as plt
\hbox{import numpy as np}\\
import seaborn as sns
from sklearn import linear_model, preprocessing
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn import feature_selection, metrics
from \ pandas.plotting \ import \ scatter\_matrix
from seaborn import pairplot
from sklearn import feature_selection, metrics
{\it from \ sklearn.preprocessing \ import \ StandardScaler}
os.chdir('/content/drive/MyDrive/big_data_pred/cw/')
dbdata = pd.read_csv('diabetic_data.csv')
#shape of the daibetes dataframe
dbdata.shape
#this implies that the data has 101,766 rows (observations) and 50 columns (attributes)
→ (101766, 50)
dbdata.head()
```

	encounter_id	patient_nbr	race	gender	age	weight	admission_type_id	discharge_disposition_id	admission_source_id	time
0	2278392	8222157	Caucasian	Female	[0- 10)	?	6	25	1	
1	149190	55629189	Caucasian	Female	[10- 20)	?	1	1	7	
2	64410	86047875	AfricanAmerican	Female	[20- 30)	?	1	1	7	
3	500364	82442376	Caucasian	Male	[30- 40)	?	1	1	7	
4	16680	42519267	Caucasian	Male	[40- 50)	?	1	1	7	
4										

dbdata.info()

 $\overrightarrow{\exists}$

#some of the columns which are meant to be categorical are numerical in the data #hence we need to convert them to categorical

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 101766 entries, 0 to 101765
Data columns (total 50 columns):

Data	COTAINIS (COLAT 20 COTAINIS) •	
#	Column	Non-Null Count	Dtype
0	encounter_id	101766 non-null	int64
1	patient_nbr	101766 non-null	int64
2	race	101766 non-null	object
3	gender	101766 non-null	object
4	age	101766 non-null	object
5	weight	101766 non-null	object
6	admission_type_id	101766 non-null	int64
7	discharge_disposition_id	101766 non-null	int64
8	admission_source_id	101766 non-null	int64
9	time_in_hospital	101766 non-null	int64
10	payer_code	101766 non-null	object
11	medical_specialty	101766 non-null	object

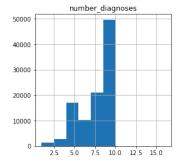
12	num_lab_procedures	101766	non-null	int64
13	num_procedures	101766	non-null	int64
14	num_medications	101766	non-null	int64
15	number_outpatient	101766	non-null	int64
16	number_emergency	101766	non-null	int64
17	number_inpatient	101766	non-null	int64
18	diag_1	101766	non-null	object
19	diag_2	101766	non-null	object
20	diag_3	101766	non-null	object
21	number_diagnoses	101766	non-null	int64
22	max_glu_serum	101766	non-null	object
23	A1Cresult	101766	non-null	object
24	metformin	101766	non-null	object
25	repaglinide	101766	non-null	object
26	nateglinide	101766	non-null	object
27	chlorpropamide	101766	non-null	object
28	glimepiride	101766	non-null	object
29	acetohexamide	101766	non-null	object
30	glipizide	101766	non-null	object
31	glyburide	101766	non-null	object
32	tolbutamide	101766	non-null	object
33	pioglitazone	101766	non-null	object
34	rosiglitazone	101766	non-null	object
35	acarbose	101766	non-null	object
36	miglitol	101766	non-null	object
37	troglitazone	101766	non-null	object
38	tolazamide	101766	non-null	object
39	examide	101766	non-null	object
40	citoglipton	101766	non-null	object
41	insulin	101766	non-null	object
42	glyburide-metformin	101766	non-null	object
43	glipizide-metformin	101766	non-null	object
44	glimepiride-pioglitazone	101766	non-null	object
45	metformin-rosiglitazone	101766	non-null	object
46	metformin-pioglitazone	101766	non-null	object
47	change	101766	non-null	object
48	diabetesMed	101766	non-null	object
49	readmitted	101766	non-null	object
d+vn	es: int64(13) ohiect(37)			

dtypes: int64(13), object(37) memory usage: 38.8+ MB

dbdata.describe()

→		encounter_id	patient_nbr	admission_type_id	discharge_disposition_id	admission_source_id	time_in_hospital	num_lab_procedures
	count	1.017660e+05	1.017660e+05	101766.000000	101766.000000	101766.000000	101766.000000	101766.000000
	mean	1.652016e+08	5.433040e+07	2.024006	3.715642	5.754437	4.395987	43.095641
	std	1.026403e+08	3.869636e+07	1.445403	5.280166	4.064081	2.985108	19.674362
	min	1.252200e+04	1.350000e+02	1.000000	1.000000	1.000000	1.000000	1.000000
	25%	8.496119e+07	2.341322e+07	1.000000	1.000000	1.000000	2.000000	31.000000
	50%	1.523890e+08	4.550514e+07	1.000000	1.000000	7.000000	4.000000	44.000000
	75%	2.302709e+08	8.754595e+07	3.000000	4.000000	7.000000	6.000000	57.000000
	may	4 4 38672 <u>e</u> +08	1 895026e+08	8 000000	28 000000	25 000000	14 000000	132 000000

dbdata.hist(figsize = (20,20))
plt.show()



dbdata.isnull().sum()

#while there are several missing data in the dataframe, isnull() function shows that there is non.
#this is because the null data are not represented by na or nan, but rather by space and ?
#to replace the null values with na so that we can track them i'll use the regex functions

} •	encounter_id	0
_	patient_nbr	0
	race	0
	gender	0
	age	0
	weight	0
	admission_type_id	0
	discharge_disposition_id	0
	admission_source_id	0
	time_in_hospital	0
	payer_code	0

```
medical_specialty
                                   0
     num_lab_procedures
                                   0
     num_procedures
                                   0
     num_medications
                                   0
     number_outpatient
                                   0
     number_emergency
                                   0
     number_inpatient
                                   0
                                   0
     {\sf diag\_1}
     diag_2
                                   0
     diag_3
                                   0
     number_diagnoses
                                   0
                                   0
     max_glu_serum
     A1Cresult
                                   0
     metformin
                                   0
     repaglinide
                                   0
     nateglinide
                                   0
     chlorpropamide
                                   0
     glimepiride
                                   0
     acetohexamide
                                   0
     glipizide
                                   0
     glyburide
                                   0
     tolbutamide
                                   0
     pioglitazone
                                   0
     rosiglitazone
                                   0
     acarbose
                                   0
     miglitol
                                   0
     troglitazone
                                   0
     tolazamide
     examide
                                   0
     citoglipton
                                   0
     insulin
     glyburide-metformin
                                   0
     glipizide-metformin
                                   0
     glimepiride-pioglitazone
                                   0
     metformin-rosiglitazone
                                   0
     metformin-pioglitazone
                                   0
                                   0
     change
     diabetesMed
                                   0
     readmitted
                                   0
     dtype: int64
dbdata = dbdata.replace('?', np.nan) # replace ? with nan
dbdata = dbdata.replace('^\s+', np.nan, regex=True) # replace empty spaces with nan
print(dbdata.isna().sum())
print(dbdata.shape[0])
                                       0
→ encounter_id
     patient_nbr
                                       0
                                    2273
     race
     gender
                                       0
                                       0
     age
     weight
                                   98569
     admission_type_id
                                       0
     {\tt discharge\_disposition\_id}
                                       0
     admission_source_id
                                       0
     time_in_hospital
                                       0
                                   40256
     payer_code
     medical_specialty
                                   49949
     num_lab_procedures
                                       0
                                       0
     {\tt num\_procedures}
     num_medications
                                       0
     number_outpatient
                                       0
     number_emergency
number_inpatient
                                       0
                                       0
     diag_1
                                      21
     diag_2
                                     358
                                    1423
     diag_3
     number_diagnoses
                                       0
     max_glu_serum
                                       0
     A1Cresult
                                       0
     metformin
                                       0
     repaglinide
                                       0
     nateglinide
                                       0
                                       0
     chlorpropamide
     glimepiride
                                       0
     ace to hexamide
                                       0
     glipizide
                                       0
     glyburide
                                       0
     tolbutamide
                                       0
     pioglitazone
                                       0
     {\tt rosiglitazone}
                                       0
     acarbose
                                       0
```

0

miglitol

```
tolazamide
                                     0
     examide
     citoglipton
     insulin
                                     0
     glyburide-metformin
                                     0
     glipizide-metformin
                                     0
     glimepiride-pioglitazone
                                     0
     metformin-rosiglitazone
                                     0
     metformin-pioglitazone
                                     0
     change
                                     0
     diabetesMed
                                     0
     readmitted
                                     0
     dtype: int64
     101766
#Drop column with more than 50% missing values
dbdata.dropna(thresh=len(dbdata.index)/2, axis=1, inplace=True)
dbdata.shape
→ (101766, 49)
# def remove(df):
   for x in df.columns:
      f = df[x].value_counts()/df.shape[0]
      if f.to_frame().iloc[:, 0].max() > 0.95:
        df.drop(x, axis = 1, inplace = True)
# remove(dbdata)
# dbdata.shape
for x in dbdata.columns:
    f = dbdata[x].value_counts()/dbdata.shape[0]
    if f.to_frame().iloc[:, 0].max() >= 0.95:
      dbdata.drop(x, axis = 1, inplace = True)
print(dbdata.shape)
→ (101766, 33)
dbdata['age'].value_counts()
def age(df):
  for i in range(df.shape[0]):
    if(df.loc[i,'age']=='[70-80)'):
        df.loc[i, 'age']=75
    elif(df.loc[i,'age']=='[60-70)'):
        df.loc[i,'age']=65
    elif(df.loc[i,'age']=='[50-60)'):
        df.loc[i,'age']=55
    elif(df.loc[i,'age']=='[80-90)'):
        df.loc[i,'age']=85
    elif(df.loc[i,'age']=='[40-50)'):
        df.loc[i,'age']=45
    elif(df.loc[i,'age']=='[30-40)'):
        df.loc[i,'age']=35
    elif(df.loc[i,'age']=='[90-100)'):
        df.loc[i,'age']=95
    elif(df.loc[i,'age']=='[20-30)'):
        df.loc[i,'age']=25
    elif(df.loc[i,'age']=='[10-20)'):
        df.loc[i,'age']=15
    elif(df.loc[i,'age']=='[0-10)'):
        df.loc[i,'age']=5
age(dbdata)
# Source: https://stackoverflow.com/questions/55159244/age-range-to-age-numerical-valuepython
dbdata['age'] = pd.to_numeric(dbdata['age'], errors='coerce')
```

troglitazone

0

```
#replacing missing values in the follwing column with
diagcols = ['diag_1','diag_2', 'diag_3']

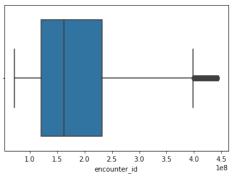
dbdata[diagcols] = dbdata[diagcols].fillna(0)

dbdata.dropna(inplace = True)

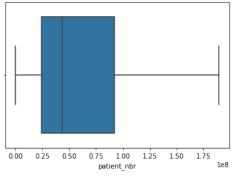
numeric_cols = dbdata.select_dtypes(include=['int64']).copy()
categorica_cols = dbdata.select_dtypes(include=['object']).copy()

for x in numeric_cols:
    sns.boxplot(dbdata.loc[:,x])
    plt.show()
```

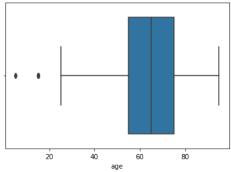
🚁 /usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro 🛦 FutureWarning



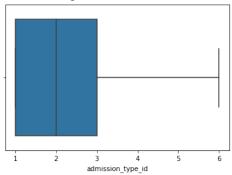
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



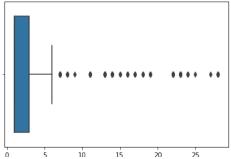
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/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

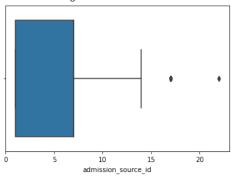


/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

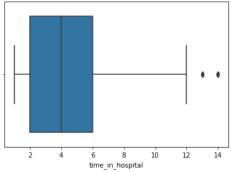


discharge_disposition_id

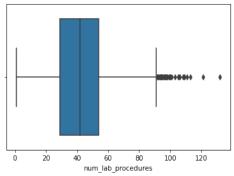
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



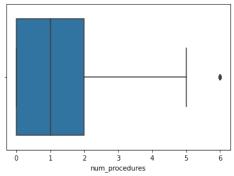
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



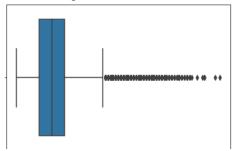
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

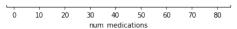


/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

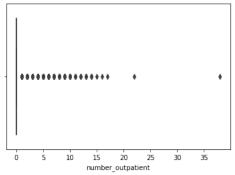


/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

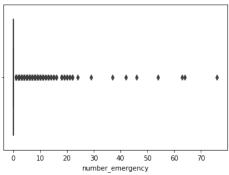




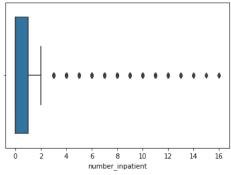
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



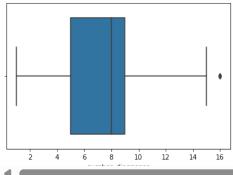
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/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



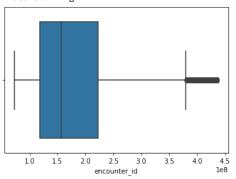
#removing outliers above 3 std from the mean of the numeric columns
upperlimit = numeric_cols.mean() + 3*numeric_cols.std()
rmoutlier = numeric_cols[numeric_cols < upperlimit]</pre>

for x in numeric_cols.columns:
 shell = dbdata[x] <= (dbdata[x].mean() + 3*dbdata[x].std())
 dbdata = dbdata[shell]</pre>

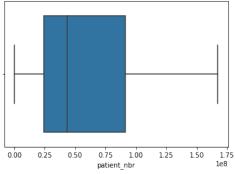
 ${\tt dbdata.shape}$



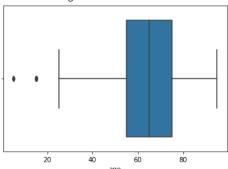
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



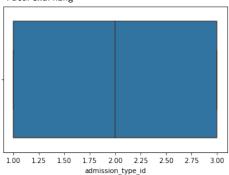
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



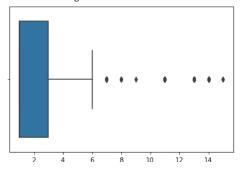
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

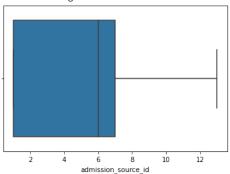


/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

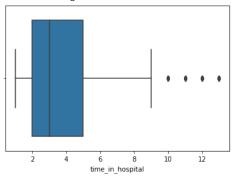


discharge_disposition_id

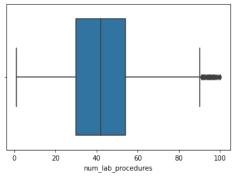
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



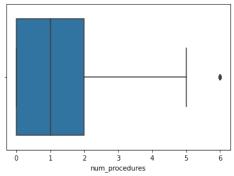
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



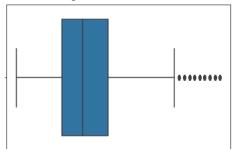
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

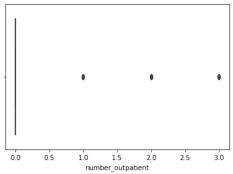


/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning

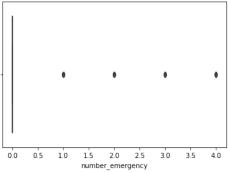




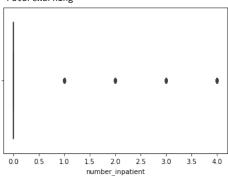
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



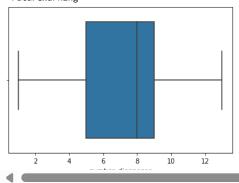
/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: x. Fro FutureWarning



	encounter_id	patient_nbr	race	gender	age	${\tt admission_type_id}$	${\tt discharge_disposition_id}$	admission_source_id	time_in_hosp	
20737	72848634	20377854	Caucasian	Female	65	2	1	1		
20824	73062156	20408121	Caucasian	Female	95	1	1	7		
21083	73731852	20542797	Caucasian	Male	75	1	2	7		
23879	81355914	7239654	Caucasian	Female	75	1	3	6		
23922	81458964	15466212	Caucasian	Male	75	1	3	6		
101249	436785812	87833862	Caucasian	Male	75	2	3	1		
101268	437270768	124367945	Caucasian	Male	65	1	1	7		
101278	437309498	52653654	Caucasian	Male	65	1	1	7		
101284	437331638	142026269	Caucasian	Male	85	2	1	4		
101286	437331728	46488123	Caucasian	Female	85	2	1	1		
17601 rowe x 22 columne										

readmitted_count = dbdata.groupby(['readmitted']).size().sort_values(ascending=False)
readmitted_count

Providing the information that 37.25% is the mean average rate of readmittence

readmitted
NO 11045
>30 5158
<30 1398
dtype: int64

₹

> Data Exploration

[] L, 37 cells hidden

> Question 2

Data exploration: Carry out a data exploration using appropriate plots to identify patterns or trends in the data. Bearing in mind our objective, we need to assess the impact of the predictors e.g. age, race, gender, or diagnosis type on the outcome (readmitted). Use graphs to prove or disprove the following hypotheses:

- Age has a higher impact on readmission.
- African Americans are more likely to be re-admitted than other ethnic groups.
- Momen patients are more likely to be re-admitted than men.
- Diagnose types have a higher impact on re-admission rates. For this purpose, you need to take into account the icd_codes and plot say diag_1 vs readmitted.
- Hint 1: You may want to join both datasets diabetic_data.csv and icd_codes.csv.
- Hint 2: Check for distinct values in categorical data and their frequencies. If there are too many distinct values (levels), then you may want to reduce the number of levels by grouping some of the detailed levels. This could be the case for race or diagnosis types.
- Hint 3: You may want to transform the readmitted column values to be 0 if the value is NO and 1 otherwise for a better exploration of the data.

[] L, 11 cells hidden

Part 3

```
subset_list = ['num_medications', 'number_outpatient', 'number_emergency', 'time_in_hospital',\
'number_inpatient', 'encounter_id', 'age', 'num_lab_procedures', 'number_diagnoses',\
'num_procedures', 'readmitted']

logsubset = dbdata[subset_list]
indp_col = dbdata[subset_list].select_dtypes(include=[np.number])
indp_col
#of all the independent variable only age is not numerical.
```

_		num_medications	number_outpatient	number_emergency	time_in_hospital	number_inpatient	encounter_id	age	num_lab_procedures
	20737	11	0	0	3	0	72848634	65	59
	20824	9	0	0	4	0	73062156	95	56
	21083	18	0	0	10	0	73731852	75	68
	23879	19	0	0	12	0	81355914	75	77
	23922	10	0	0	12	0	81458964	75	60
	101249	19	0	0	10	0	436785812	75	59
	101268	19	0	0	2	0	437270768	65	53
	101278	14	1	0	7	0	437309498	65	54
	101284	15	0	0	3	1	437331638	85	1
	101286	13	0	0	2	1	437331728	85	41
·	17601 rov	ve x 10 columne							•

logsubset

readmittion_test = logsubset.copy()

readmittion_test[readmittion_test['readmitted'] != 'NO']

→		num_medications	number_outpatient	number_emergency	time_in_hospital	number_inpatient	encounter_id	age	num_lab_procedures
	24028	5	0	0	6	0	81762780	55	33
	24247	13	0	0	9	0	82331772	75	64
	24262	2	0	0	1	0	82348062	65	68
	24304	7	0	0	4	0	82491186	55	33
	24310	12	3	0	3	0	82496730	75	60
	101148	29	0	0	8	3	435565568	75	28
	101167	13	0	0	1	0	436065734	75	1
	101214	33	0	0	3	0	436644764	45	37
	101233	13	0	0	3	0	436704890	85	30
	101240	13	0	0	5	0	436726154	95	20
F	5556 rows	e x 11 columne							•

readmittion_test.loc[:,'readmitted'][readmittion_test.loc[:,'readmitted'] != 'NO'] = 0
readmittion_test.loc[:,'readmitted'][readmittion_test.loc[:,'readmitted'] == 'NO'] = 1

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:1: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-cc """Entry point for launching an IPython kernel.

 $/usr/local/lib/python 3.7/dist-packages/ipykernel_launcher.py: 2: Setting With CopyWarning: \\$

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-cc