DM Project

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```
CO37 Section B batch b2
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import warnings
warnings.filterwarnings('ignore')
```

1) Choosing a data set

```
df = pd.read_csv("/Users/devansh/Documents/ML
Projects/project4/diabetes.csv")
df.head()
```

D	Pregnancies MI \	Glucose	BloodPressure	SkinThickness	Insulin	
0	•	148	72	35	0	33.6
1	1	85	66	29	0	26.6
2	8	183	64	0	0	23.3
3	1	89	66	23	94	28.1
4	0	137	40	35	168	43.1

	DiabetesPedigreeFunction	Age	Outcome
0	0.627	50	1
1	0.351	31	0
2	0.672	32	1
3	0.167	21	0
4	2.288	33	1

2) Reading the data and understanding the dataset

df.dtypes

```
Pregnancies
                               int64
Glucose
                               int64
BloodPressure
                               int64
SkinThickness
                               int64
Insulin
                               int64
BMI
                             float64
DiabetesPedigreeFunction
                             float64
                               int64
Age
Outcome
                               int64
dtype: object
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 9 columns):
#
     Column
                                Non-Null Count
                                                 Dtype
- - -
     -----
                                                 ----
 0
     Pregnancies
                                768 non-null
                                                 int64
     Glucose
                                768 non-null
 1
                                                 int64
 2
     BloodPressure
                                768 non-null
                                                 int64
 3
     SkinThickness
                                768 non-null
                                                 int64
 4
     Insulin
                                768 non-null
                                                 int64
 5
     BMI
                                768 non-null
                                                 float64
 6
     DiabetesPedigreeFunction
                                768 non-null
                                                 float64
 7
                                768 non-null
                                                 int64
 8
     Outcome
                                768 non-null
                                                 int64
dtypes: float64(2), int64(7)
memory usage: 54.1 KB
df.describe().round(2).style.background gradient()
<pandas.io.formats.style.Styler at 0x7ff324d584c0>
df.isnull().sum()
Pregnancies
                             0
Glucose
                             0
BloodPressure
                             0
                             0
SkinThickness
                             0
Insulin
                             0
DiabetesPedigreeFunction
                             0
                             0
Aae
Outcome
                             0
dtype: int64
```

3) Cleaning the dataset

df[['Glucose','BloodPressure','SkinThickness','Insulin','BMI','Diabete
sPedigreeFunction','Age']] =

```
df[['Glucose','BloodPressure','SkinThickness','Insulin','BMI','Diabete
sPedigreeFunction','Age']].replace(0,np.NaN)
df.fillna(df.mean(), inplace = True) #Filled Mising values with Mean
df.isnull().sum()
```

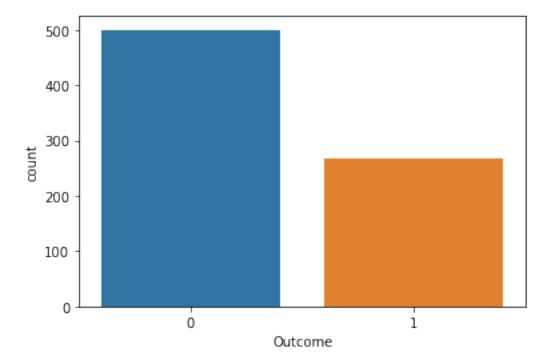
#all we did here is replaced 0 will nul and replaced all the null by the mean of the column

Pregnancies	0
Glucose	0
BloodPressure	0
SkinThickness	0
Insulin	0
BMI	0
DiabetesPedigreeFunction	0
Age	0
Outcome	0
dtvpe: int64	

4) Visulaztion of dataset

sns.countplot(df.Outcome)

<AxesSubplot:xlabel='Outcome', ylabel='count'>



```
lis=["don't have diabetes", "have diabetes"]
have_or_not = df["Outcome"].value_counts().tolist()
values = [have_or_not[0], have_or_not[1]]
fig = px.pie(values=df['Outcome'].value_counts(), names=lis ,
```

```
width=800, height=400, color discrete sequence=["skyblue","black"]
              ,title="percentage between have diabetes or not")
fig.show()
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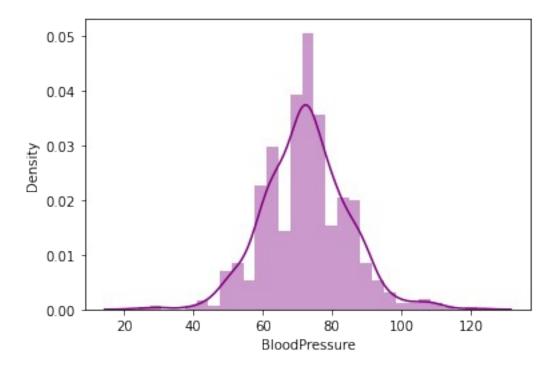
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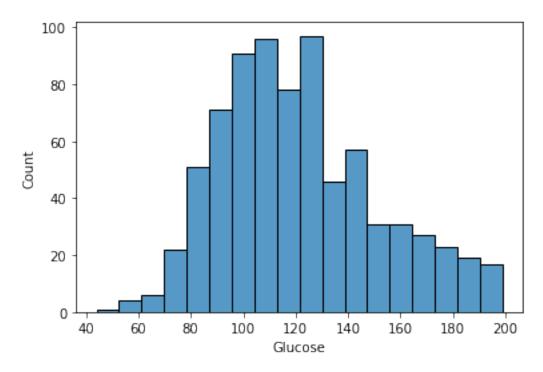
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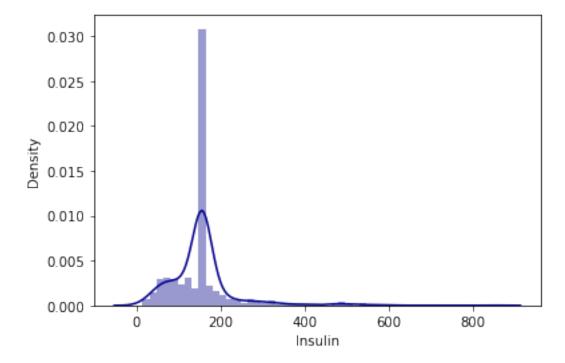
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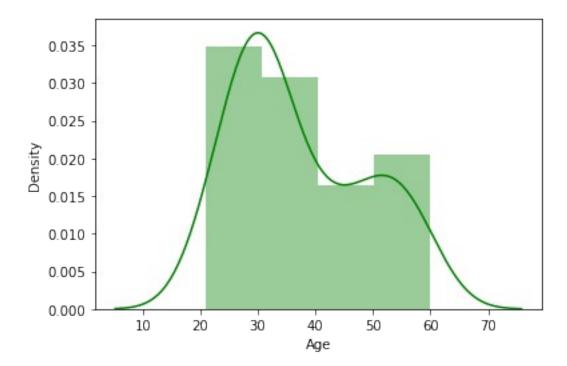
sns.histplot(df["Glucose"]);



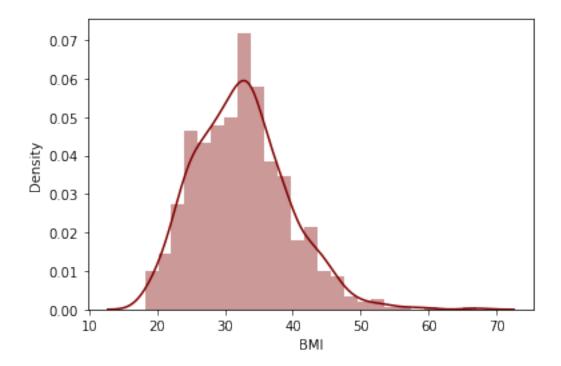
sns.distplot(df["Insulin"], color = "DarkBlue");



sns.distplot(df["Age"][:50], color = "Green");

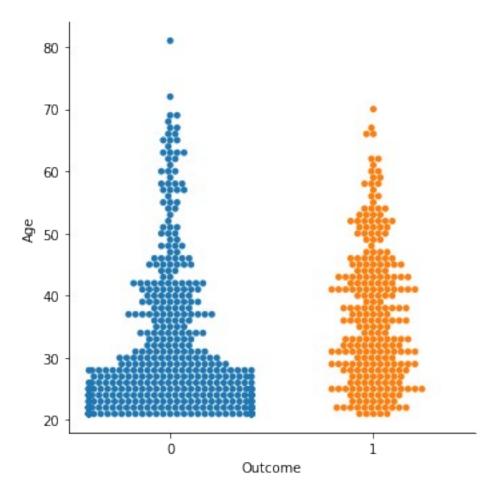


sns.distplot(df["BMI"], color = "Maroon");

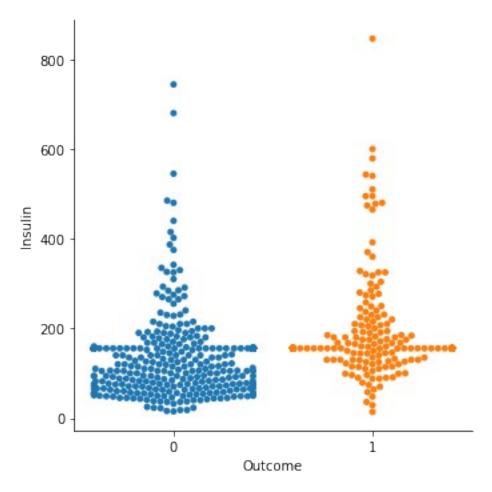


Advance cat plot

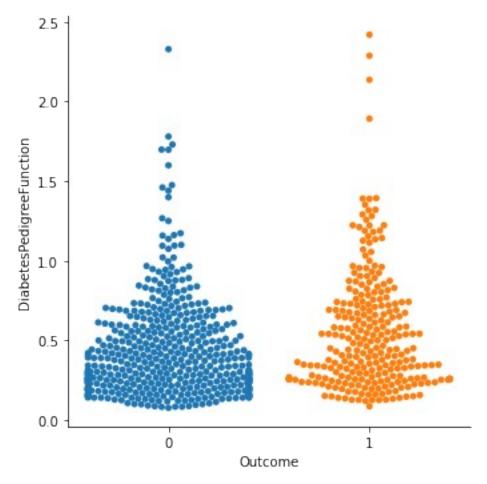
```
# cat plot
sns.catplot(x = "Outcome", y = "Age", hue = "Outcome", kind = "swarm",
data = df)
plt.show()
```



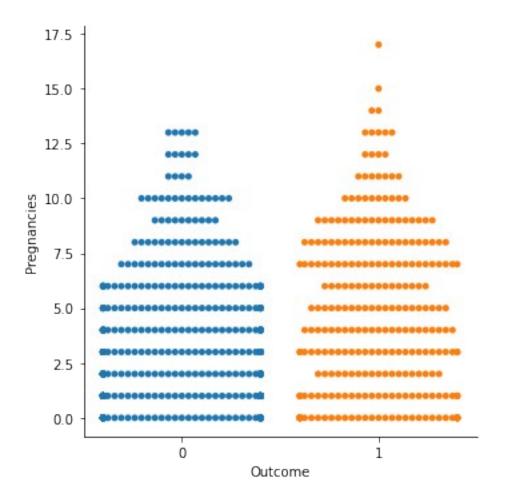
```
sns.catplot(x = "Outcome", y = "Insulin", hue = "Outcome", kind =
"swarm", data = df)
plt.show()
```



```
sns.catplot(x = "Outcome", y = "DiabetesPedigreeFunction", hue =
"Outcome", kind = "swarm", data = df)
plt.show()
```



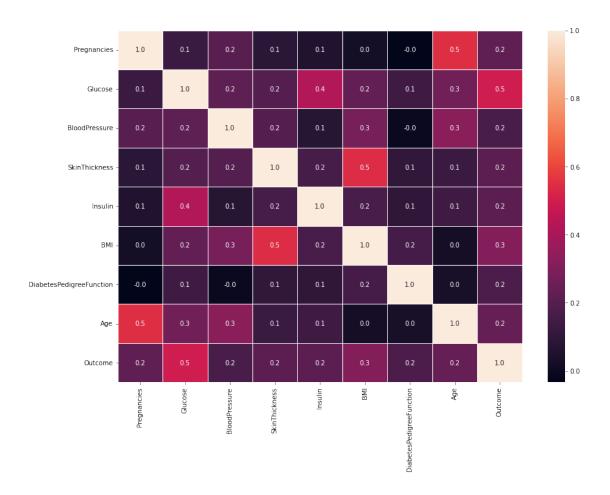
```
sns.catplot(x = "Outcome", y = "Pregnancies", hue = "Outcome", kind =
"swarm", data = df)
plt.show()
```



5)Feature Selection for making the output more and more effective df.corr()

	Pregnancies	Glucose	BloodPressure
SkinThickness \ Pregnancies 0.082989	1.000000	0.127911	0.208522
Glucose 0.192991	0.127911	1.000000	0.218367
BloodPressure 0.192816	0.208522	0.218367	1.000000
SkinThickness 1.000000	0.082989	0.192991	0.192816
Insulin 0.158139	0.056027	0.420157	0.072517
BMI 0.542398	0.021565	0.230941	0.281268
DiabetesPedigreeFunct 0.100966	ion -0.033523	0.137060	-0.002763
Age	0.544341	0.266534	0.324595

```
0.127872
                             0.221898 0.492928
                                                      0.166074
Outcome
0.215299
                           Insulin
                                              DiabetesPedigreeFunction
                                         BMI
Pregnancies
                          0.056027
                                    0.021565
                                                             -0.033523
Glucose
                          0.420157
                                    0.230941
                                                              0.137060
BloodPressure
                          0.072517 0.281268
                                                             -0.002763
SkinThickness
                          0.158139
                                    0.542398
                                                              0.100966
Insulin
                          1.000000
                                    0.166586
                                                              0.098634
BMI
                          0.166586
                                    1.000000
                                                              0.153400
DiabetesPedigreeFunction 0.098634
                                    0.153400
                                                              1.000000
                          0.136734
                                                              0.033561
Age
                                    0.025519
                          0.214411 0.311924
Outcome
                                                              0.173844
                                     Outcome
                               Age
Pregnancies
                          0.544341
                                    0.221898
Glucose
                          0.266534
                                    0.492928
BloodPressure
                          0.324595
                                    0.166074
SkinThickness
                          0.127872
                                    0.215299
Insulin
                          0.136734
                                    0.214411
BMI
                          0.025519
                                    0.311924
                          0.033561
DiabetesPedigreeFunction
                                    0.173844
                          1.000000
                                    0.238356
Outcome
                          0.238356
                                    1.000000
plt.figure(figsize = (14, 10))
sns.heatmap(df.corr(), annot = True, fmt = ".1f", linewidths = .7)
plt.show()
```



```
from sklearn.ensemble import RandomForestClassifier
clf = RandomForestClassifier()
x=df[df.columns[:8]]
y=df.Outcome
clf.fit(x,y)
feature_imp = pd.DataFrame(clf.feature_importances_,index=x.columns)
feature imp.sort values(by = 0 , ascending = False)
Glucose
                          0.261190
BMI
                          0.154962
Age
                          0.133530
DiabetesPedigreeFunction
                          0.127818
Insulin
                          0.085612
BloodPressure
                          0.081630
                          0.079389
Pregnancies
SkinThickness
                          0.075869
```

first 4 features displayed maybe important for us!!We might neglect the rest from sklearn.model selection import train test split

```
features = df[["Glucose", 'BMI', 'Age', 'DiabetesPedigreeFunction']]
labels = df.Outcome
features.head()
           BMI Age DiabetesPedigreeFunction
   Glucose
0
    148.0 33.6 50
     85.0 26.6
                  31
                                          0.351
1
2
     183.0 23.3
                   32
                                          0.672
3
     89.0 28.1
                  21
                                          0.167
     137.0 43.1
                  33
                                          2.288
features train, features test, labels train, labels test =
train test split(features, labels, stratify=df.Outcome, test size=0.4)
```

Decision Trees

```
from sklearn.tree import DecisionTreeClassifier
dtclf = DecisionTreeClassifier()
dtclf.fit(features_train,labels_train)
dtclf.score(features_test,labels_test)
```

0.7045454545454546

SVM

```
from sklearn import svm
clf = svm.SVC(kernel="linear")
clf.fit(features_train, labels_train)
clf.score(features_test, labels_test)
```

0.7954545454545454

Naive Bayes

```
from sklearn import naive_bayes
nbclf = naive_bayes.GaussianNB()
nbclf.fit(features_train,labels_train)
nbclf.score(features_test,labels_test)
0.8051948051948052
```

K Neighbor

```
from sklearn.neighbors import KNeighborsClassifier
knnclf = KNeighborsClassifier(n_neighbors=2)
knnclf.fit(features_train,labels_train)
print(knnclf.score(features_test,labels_test))
```

0.72727272727273

```
Logistic Regression
```

```
from sklearn.linear_model import LogisticRegression
clf1 = LogisticRegression()
clf1.fit(features_train,labels_train)
clf1.score(features_test,labels_test)
0.8051948051948052
```

Accurary Table

```
algos = ["Support Vector Machine", "Decision Tree", "Logistic
Regression", "K Nearest Neighbor", "Naive Bayes"]
clfs =
[svm.SVC(kernel="linear"), DecisionTreeClassifier(), LogisticRegression(
),KNeighborsClassifier(n neighbors=2),naive bayes.GaussianNB()]
result = []
for clff in clfs:
    clff.fit(features train, labels train)
    acc = clff.score(features test, labels test)
    result.append(acc)
result df = pd.DataFrame(result,index=algos)
result df.columns=["Accuracy"]
result_df.sort_values(by="Accuracy",ascending=False)
                        Accuracy
Logistic Regression
                        0.805195
Naive Bayes
                        0.805195
Support Vector Machine 0.795455
K Nearest Neighbor 0.727273
Decision Tree
                        0.711039
```

Cross Validation

```
#Cross Validation
```

```
from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
kfold =KFold(n_splits=10)

algos = ["Support Vector Machine", "Decision Tree", "Logistic
Regression", "K Nearest Neighbor", "Naive Bayes"]
clfs =
[svm.SVC(kernel="linear"), DecisionTreeClassifier(), LogisticRegression(), KNeighborsClassifier(n_neighbors=2), naive_bayes.GaussianNB()]
cv_results=[]
for classifiers in clfs:
    cv_score =
cross_val_score(classifiers, features, labels, cv=kfold, scoring="accuracy")
```

Observations- We can see the accuracy changed a bit this time. It is because we have done cross validation and trained and tested the algorithms on diffrent combinations of data. From the above output, it is clear that for this dataset, SVM, Logistic Regression and Naive Bayes works better