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In [224]: # NAME- ANKIT KUMAR PATHAK
          #ROLL NO -201046016
          # EXPLORATORY DATA ANALYSIS
          # This dataset was originally from the National Institute of Diabetes
          # and Digestive and Kidney Diseases.
          # The purpose of the dataset is to diagnostically predict whether a patient
          # has diabetes based on the specific diagnostic measures included in the data set.
          # Various restrictions have been imposed on the selection of these
          # samples from a larger database
In [199]: import numpy as np
          import pandas as pd
          import seaborn as sns
          import matplotlib.pyplot as plt
          import plotly.graph_objs as go
          import plotly.offline as py
In [200]: # Reading Data
          diabetes=pd.read csv('diabetes.csv')
```

df = diabetes.copy() df.head()

In [201]: df

Out[201]:

	Pregnancies	Glucose	BloodPressure	SkinThickness	Insulin	ВМІ	DiabetesPedigreeFunction	Age	Outcome
0	6	148.0	72.0	35.0	169.5	33.6	0.627	50	1
1	1	85.0	66.0	29.0	102.5	26.6	0.351	31	0
2	8	183.0	64.0	32.0	169.5	23.3	0.672	32	1
3	1	89.0	66.0	23.0	94.0	28.1	0.167	21	0
4	0	137.0	40.0	35.0	168.0	43.1	2.288	33	1
763	10	101.0	76.0	48.0	180.0	32.9	0.171	63	0
764	2	122.0	70.0	27.0	102.5	36.8	0.340	27	0
765	5	121.0	72.0	23.0	112.0	26.2	0.245	30	0
766	1	126.0	60.0	32.0	169.5	30.1	0.349	47	1
767	1	93.0	70.0	31.0	102.5	30.4	0.315	23	0

768 rows × 9 columns

```
In [202]: 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
                                                                 float64
             1
                 Glucose
                                               768 non-null
             2
                 BloodPressure
                                               768 non-null
                                                                 float64
                 SkinThickness
                                               768 non-null
                                                                 float64
             4
                 Insulin
                                                                 float64
                                               768 non-null
                 BMI
                                               768 non-null
                                                                 float64
             5
                 DiabetesPedigreeFunction
                                              768 non-null
                                                                 float64
             7
                 Age
                                               768 non-null
                                                                 int64
                                               768 non-null
                 Outcome
                                                                 int64
           dtypes: float64(6), int64(3)
           memory usage: 54.1 KB
  In [ ]:
In [203]:
           df.describe().T
Out[203]:
                                                                             25%
                                                                                      50%
                                                                                                75%
                                                            std
                                     count
                                                mean
                                                                   min
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                        Pregnancies
                                     768.0
                                             3.845052
                                                        3.369578
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                                                                                    3.0000
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                                                                                                       17.00
                            Glucose
                                     768.0
                                           121.677083
                                                      30.464161
                                                                44.000
                                                                         99.75000
                                                                                  117.0000
                                                                                           140.25000
                                                                                                     199.00
                      BloodPressure
                                     768.0
                                            72.389323
                                                      12.106039
                                                                24.000
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                       SkinThickness
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                                            29.089844
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                                           141.753906
                                                      89.100847
                                                                14.000
                                                                                            169.50000
                                                                                                     846.00
                             Insulin
                                     768.0
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                                                                                            36.60000
                                BMI
                                            32.434635
                                                        6.880498
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                                                                         27.50000
                                                                                                       67.10
             DiabetesPedigreeFunction
                                     768.0
                                             0.471876
                                                        0.331329
                                                                  0.078
                                                                          0.24375
                                                                                    0.3725
                                                                                             0.62625
                                                                                                       2.42
                                     768.0
                                            33.240885
                                                      11.760232 21.000
                                                                         24.00000
                                                                                   29.0000
                                                                                            41.00000
                                                                                                       81.00
                                Age
```

Outcome 768.0

0.348958

0.476951

0.000

0.00000

0.0000

1.00000

1.00

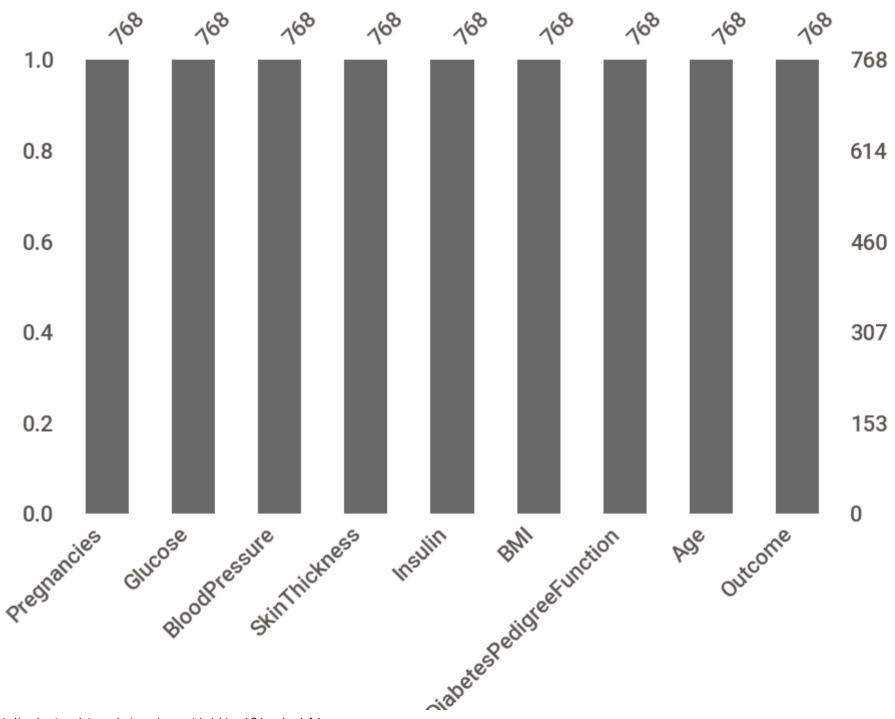
```
In [204]: # Handling with Missing Values
# In this dataset missing data are filled with 0. First, we are gonna change zeros with NaN

In [205]: df[['Glucose','BloodPressure','SkinThickness','Insulin','BMI']] = df[['Glucose','BloodPressure','SkinThickness',
'Insulin','BMI']].replace(0, np.NaN)

In [206]: # After filling the 0s with the value of NaN, the missing values
# will be visualized. We use the missingno library for this.
```

In [207]: msno.bar(df,figsize=(10,6))

Out[207]: <matplotlib.axes._subplots.AxesSubplot at 0x2058e788c10>





In [208]: msno.heatmap(df);

C:\anaconda\lib\site-packages\seaborn\matrix.py:305: UserWarning:

Attempting to set identical bottom == top == 0 results in singular transformations; automatically expanding.

C:\anaconda\lib\site-packages\seaborn\matrix.py:305: UserWarning:

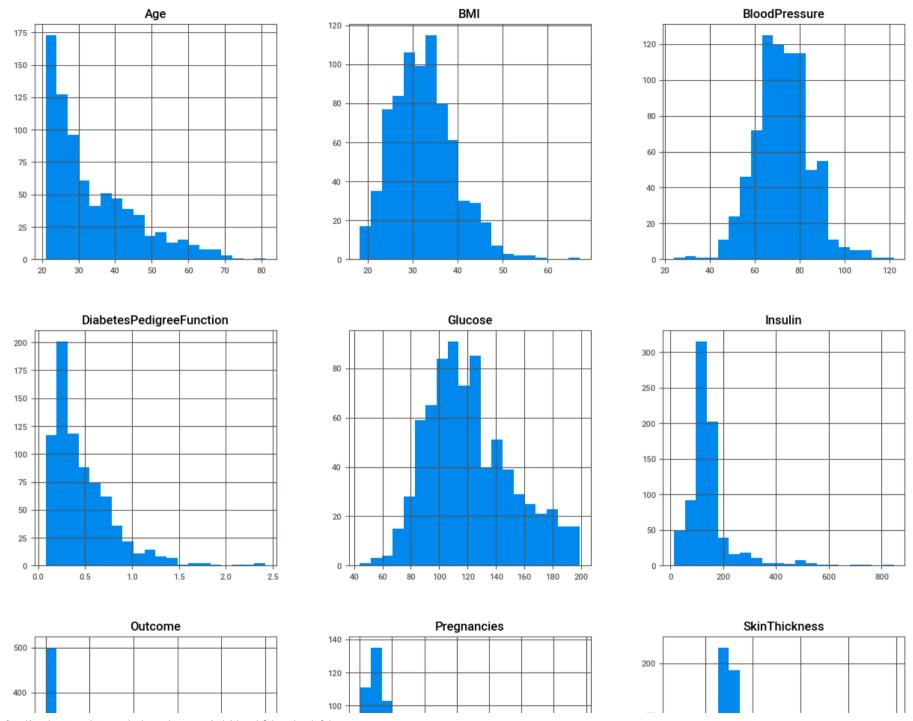
Attempting to set identical left == right == 0 results in singular transformations; automatically expanding.

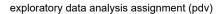


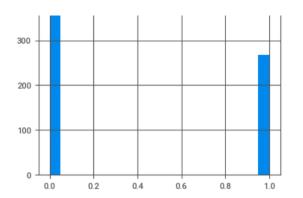
In [209]: # We will fill in each missing value with its median value.

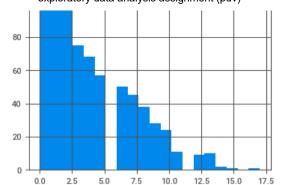
```
In [210]: def median target(var):
              temp = df[df[var].notnull()]
              temp = temp[[var, 'Outcome']].groupby(['Outcome'])[[var]].median().reset_index()
              return temp
In [211]: | columns = df.columns
          columns = columns.drop("Outcome")
          for i in columns:
              median target(i)
              df.loc[(df['Outcome'] == 0 ) & (df[i].isnull()), i] = median target(i)[i][0]
              df.loc[(df['Outcome'] == 1 ) & (df[i].isnull()), i] = median target(i)[i][1]
In [212]: # After filling if we examine null values in dataset, we will see there are not any missing values.
In [213]: | df.isnull().sum()
Out[213]: Pregnancies
                                      0
          Glucose
          BloodPressure
          SkinThickness
          Insulin
          BMI
          DiabetesPedigreeFunction
          Age
          Outcome
          dtype: int64
In [214]: # Data Visualization
In [215]: # Histogram-A histogram is a bar graph representation of a grouped data distribution
```

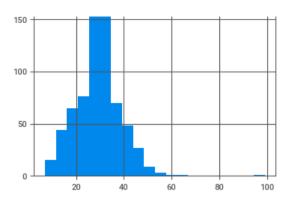
In [216]: df.hist(bins=20,figsize = (15,15));







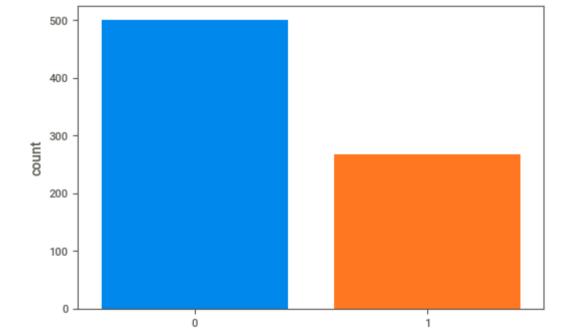




```
In [217]: # count plot
```

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In [218]: plt.title("Distribution of Outcome")
    sns.countplot(df["Outcome"], saturation=1)
```

Out[218]: <matplotlib.axes._subplots.AxesSubplot at 0x205975c0820>



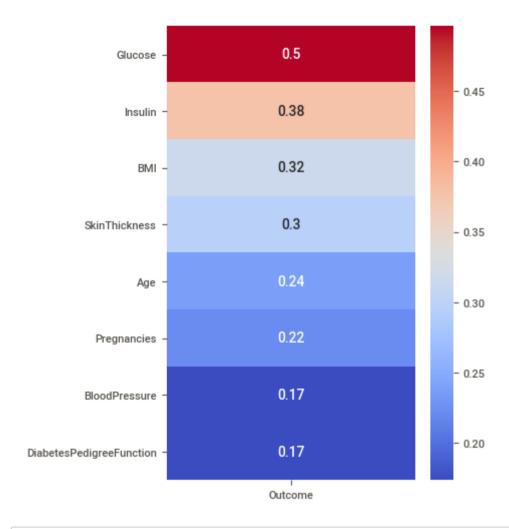
Outcome

Distribution of Outcome

In [219]: # pie plot

```
In [220]: def PlotPie(df, nameOfFeature):
    labels = [str(df[nameOfFeature].unique()[i]) for i in range(df[nameOfFeature].nunique())]
    values = [df[nameOfFeature].value_counts()[i] for i in range(df[nameOfFeature].nunique())]
    trace=go.Pie(labels=labels,values=values)
    py.iplot([trace])
PlotPie(df, "Outcome")
```

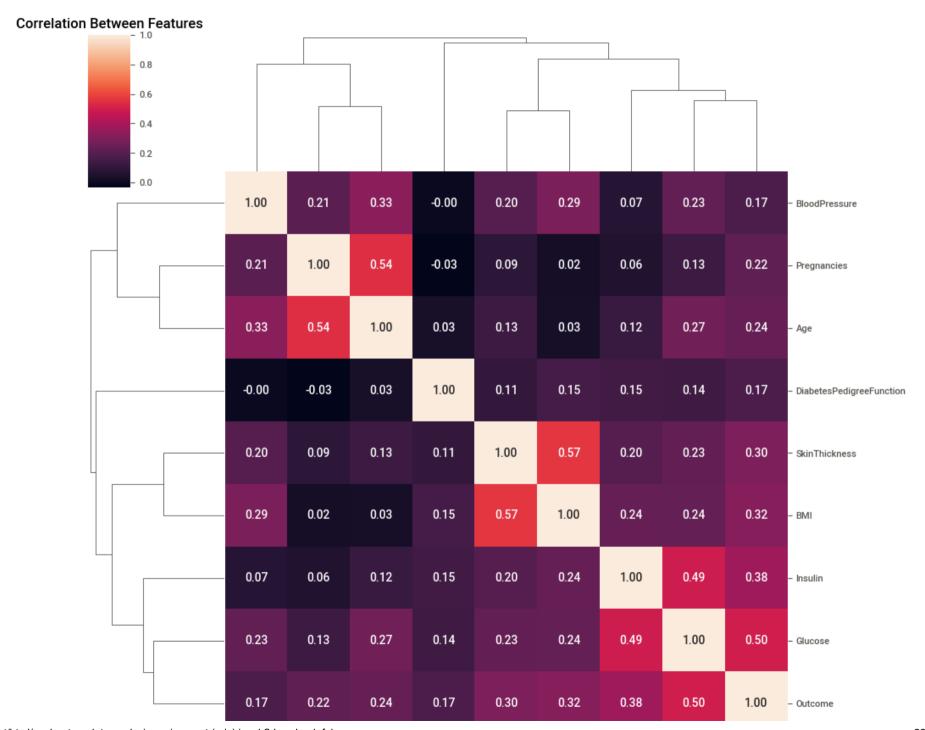
Outcome

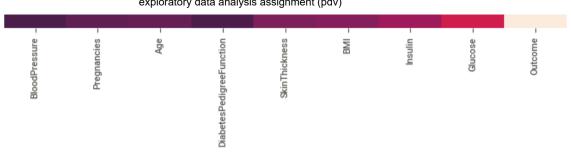


In [222]: # Correlation matrix of variables with each other.

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In [223]: corr_matrix = df.corr()
    sns.clustermap(corr_matrix, annot=True, fmt=".2f")
    plt.title("Correlation Between Features")
```

Out[223]: Text(0.5, 1.0, 'Correlation Between Features')





```
In [225]: import dtale
          dtale.show(df, ignore_duplicate=True)
 In [ ]:
          import sweetviz
          diabetes=pd.read_csv('diabetes.csv')
 In [ ]:
          # importing sweetviz
          import sweetviz as sv
          #analyzing the dataset
          advert_report = sv.analyze(df)
          #display the report
          advert_report.show_html('Advertising.html')
  In [ ]:
  In [ ]:
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