```
In [1]:
        @Authour:
                              Adharsh.S
        Date created:
                              20.10.22
        language written:
                              python
        #importing the necessary package for the usage of
        import matplotlib.pyplot as plt # for ploting the file if we needed for visualization
        import seaborn as sns
        from sklearn.model_selection import train_test_split
        from sklearn import metrics
        from sklearn.preprocessing import StandardScaler
        sns.set(color_codes=True) # adds a nice background to the graphs
        %matplotlib inline
        import pandas as pd #pandas package
        import numpy as np #numpy package
        from sklearn.discriminant_analysis import LinearDiscriminantAnalysis # for analysing the LDA
In [2]: #header_list = ["baseline value "," accelerations", "fetal_health "]
        df = pd.read_csv("train.csv")#names=header_List)
        df.head() #head of the the data
        df.shape # shape of the data
        df.info() # info of the data
        df.describe() # describtion of the data
        <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 1700 entries, 0 to 1699
        Data columns (total 22 columns):
                                                                    Non-Null Count Dtype
         # Column
                                                                     -----
            baseline value
                                                                    1700 non-null int64
         0
                                                                    1700 non-null float64
            accelerations
         1
         2 fetal movement
                                                                    1700 non-null float64
            uterine_contractions
                                                                    1700 non-null float64
                                                                    1700 non-null float64
            light_decelerations
                                                                    1700 non-null
         5
                                                                                    float64
             severe_decelerations
                                                                    1700 non-null
             prolongued_decelerations
                                                                                    float64
             abnormal_short_term_variability
                                                                    1700 non-null
                                                                                    int64
             mean_value_of_short_term_variability
                                                                    1700 non-null
                                                                                    float64
```

percentage_of_time_with_abnormal_long_term_variability 1700 non-null

21 fetal_health dtypes: float64(8), int64(14) memory usage: 292.3 KB

14 histogram_number_of_peaks

15 histogram_number_of_zeroes

11 histogram_width

12 histogram_min

13 histogram_max

16 histogram_mode

17 histogram_mean

18 histogram_median

19 histogram_variance

20 histogram_tendency

10 mean_value_of_long_term_variability

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	baseline value	accelerations	fetal_movement	uterine_contractions	light_decelerations	severe_decelerations	prolongued_decelerations	abnorma
count	1700.000000	1700.000000	1700.000000	1700.000000	1700.000000	1700.000000	1700.000000	
mean	133.213529	0.003212	0.010211	0.004356	0.001899	0.000004	0.000158	
std	9.873344	0.003888	0.050124	0.002943	0.002976	0.000059	0.000587	
min	106.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	
25%	126.000000	0.000000	0.000000	0.002000	0.000000	0.000000	0.000000	
50%	133.000000	0.002000	0.000000	0.004000	0.000000	0.000000	0.000000	
75%	140.000000	0.006000	0.003000	0.006000	0.003000	0.000000	0.000000	
max	159.000000	0.019000	0.481000	0.015000	0.015000	0.001000	0.005000	

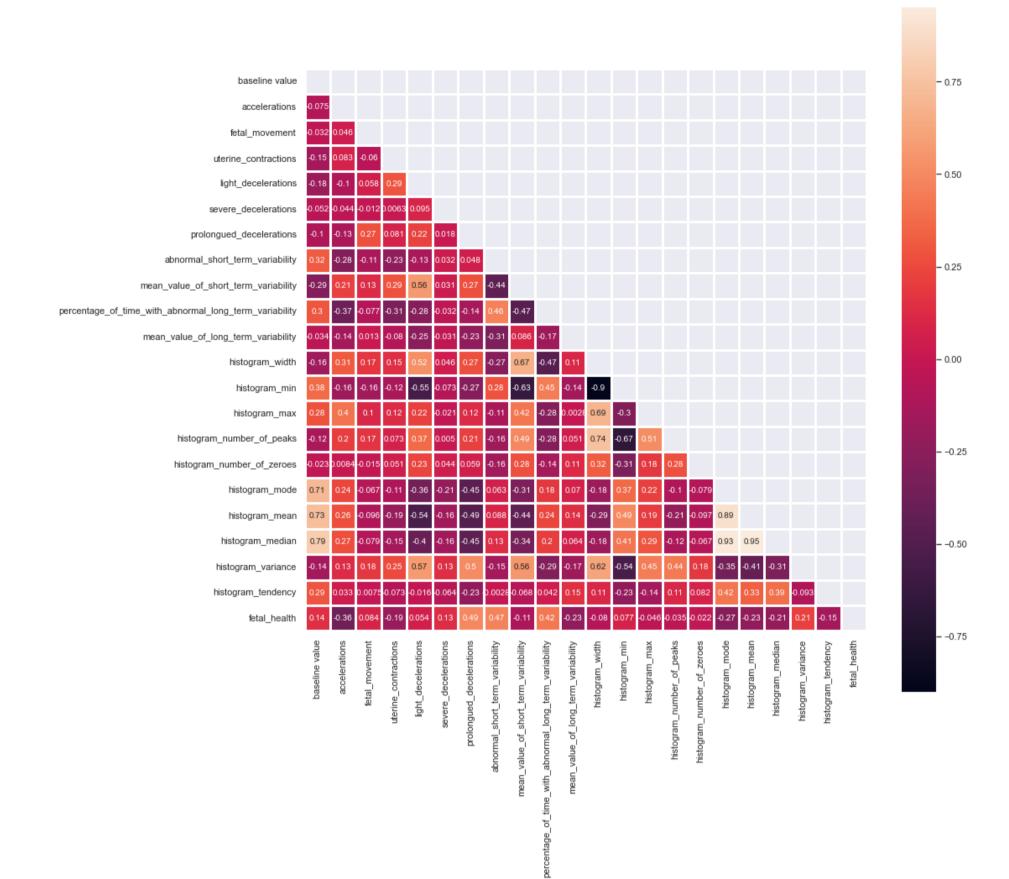
int64

float64

1700 non-null

8 rows × 22 columns

```
In [3]: # for the visualization of the data for our understanding
Target = df["fetal_health"]
corr = df.corr()
mask = np.zeros_like(corr)
mask[np.triu_indices_from(mask)] = True
with sns.axes_style("dark"):
    f, ax = plt.subplots(figsize=(15, 15))
    ax = sns.heatmap(corr,mask=mask,square=True,linewidths=2.5,cmap="rocket",annot=True)
```



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In [4]: #for avoiding the duplicates of the data
df_dup = df.drop_duplicates(subset = None , keep = 'first', inplace = False)

In [5]: print("Count of type 1.0 fetal health in the dataset ",len(df.loc[df["fetal_health"]==1.0])) #for counting the data
print("Count of type 2.0 fetal health in the dataset ",len(df.loc[df["fetal_health"]==2.0]))
print("Count of type 3.0 fetal health in the dataset ",len(df.loc[df["fetal_health"]==3.0]))

Count of type 1.0 fetal health in the dataset 1323
Count of type 2.0 fetal health in the dataset 236
```

Count of type 3.0 fetal health in the dataset 141

```
In [6]: # for Locating the data
         X = df_{dup.iloc[:,:-1]}
         y = df_dup.iloc[:,-1]
 In [7]: #for scaling the data for the usage
         scale = StandardScaler()
         X = scale.fit_transform(X)
         X = pd.DataFrame(X,columns=df_dup.iloc[:,:-1].columns)
 In [8]: #for resampling the data over a random state
         from imblearn.over_sampling import RandomOverSampler
         ROS = RandomOverSampler(random_state=42)
         X_ros, y_ros = ROS.fit_resample(X,y)
         from collections import Counter
         print('Resampled dataset shape %s' % Counter(y_ros))
         Resampled dataset shape Counter({1: 1317, 3: 1317, 2: 1317})
 In [9]: # for allocating the test and train data for the predicting
         import statsmodels.api as sm
         X = sm.add_constant(X)
         X_train, X_test, y_train, y_test = train_test_split(X, y, random_state = 10, test_size = 0.2)
         print('X_train', X_train.shape)
         print('y_train', y_train.shape)
         print('X_test', X_test.shape)
         print('y_test', y_test.shape)
         C:\Users\DELL\anaconda3\lib\site-packages\statsmodels\tsa\base\tsa_model.py:7: FutureWarning: pandas.Int64Index is depr
         ecated and will be removed from pandas in a future version. Use pandas. Index with the appropriate dtype instead.
           from pandas import (to_datetime, Int64Index, DatetimeIndex, Period,
         C:\Users\DELL\anaconda3\lib\site-packages\statsmodels\tsa\base\tsa_model.py:7: FutureWarning: pandas.Float64Index is de
         precated and will be removed from pandas in a future version. Use pandas. Index with the appropriate dtype instead.
           from pandas import (to_datetime, Int64Index, DatetimeIndex, Period,
         X_train (1354, 22)
         y_train (1354,)
         X_test (339, 22)
         y_test (339,)
         C:\Users\DELL\anaconda3\lib\site-packages\statsmodels\tsa\tsatools.py:142: FutureWarning: In a future version of pandas
         all arguments of concat except for the argument 'objs' will be keyword-only.
           x = pd.concat(x[::order], 1)
In [10]: # for the preprocessing method
         from sklearn.preprocessing import StandardScaler
         sc = StandardScaler()
         X_train = sc.fit_transform(X_train)
         X_test = sc.transform(X_test)
In [11]: |# for discriminant analysis method
         from sklearn.discriminant_analysis import LinearDiscriminantAnalysis as LDA
         lda = LDA(n_components=1)
         X train = lda.fit transform(X train, y train)
         X test = lda.transform(X test)
In [12]: from sklearn.ensemble import RandomForestClassifier
         classifier = RandomForestClassifier(max_depth=2, random_state=0)
         classifier.fit(X_train, y_train)
         y_pred = classifier.predict(X_test)
In [13]: | #for the confusion matrix and the accuracy score too
         from sklearn.metrics import confusion_matrix
         from sklearn.metrics import accuracy_score
         cm = confusion_matrix(y_test, y_pred)
         print('Accuracy' + str(accuracy_score(y_test, y_pred)))
         [[243 22
                     1]
          [ 14 34
                     0]
          [ 1 10 14]]
         Accuracy0.8584070796460177
```

```
In [14]: y_pred
1, 3, 2, 2, 1, 2, 2, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 3, 1,
               1, 2, 1, 1, 1, 2, 1, 1, 1, 2, 2, 1, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
               1, 2, 3, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 2, 2, 1, 1, 1, 1, 1, 1,
               3, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 2, 1, 2, 1, 3, 1, 3, 2, 1, 1, 2,
               1, 1, 1, 1, 2, 1, 3, 2, 1, 3, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 2,
               1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1, 1, 2, 2, 1, 1, 1,
               1, 1, 2, 1, 1, 1, 2, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
               1, 1, 2, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 3, 1, 1, 2, 1, 1, 1, 1,
               1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 1, 1, 1, 2, 2, 1,
               1, 1, 1, 1, 2, 1, 1, 3, 2, 1, 1, 1, 1, 2, 1, 1, 1, 1, 1, 1, 1, 1,
               1, 1, 1, 1, 3, 2, 1, 1, 1, 1, 1, 2, 1, 2, 2, 2, 2, 1, 3, 3, 1, 2,
               1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 3, 1, 2, 1, 1, 2, 1, 1, 1, 1,
               1, 2, 1, 1, 1, 1, 2, 2, 1, 1, 1, 2, 2, 2, 1, 1, 1, 2, 1, 2, 1, 1,
               2, 1, 1, 1, 1, 1, 1, 1], dtype=int64)
In [26]: |import csv
        data = [y_pred]
        file = open('/output.csv', 'w+', newline ='')
        with file:
            write = csv.writer(file)
            write.writerow(data)
In [36]: import csv
        data = [y_pred]
        # open the file in the write mode
        with open("/output.csv", 'w', encoding='UTF8') as f:
            # create the csv writer
            writer = csv.writer(f)
            # write a row to the csv file
            writer.writerow(data)
In [55]: from pandas import DataFrame
        data = [y_pred]
        df = DataFrame(data)
        export_csv = df.to_csv (r'output.csv', index = None, header=True)# here you have to write path, where result file will be
        print (df)
                                                    9
                1
                    2
                         3
                                  5
                                                8
                                                         ... 329 330 331 332 \
                                       6
                      1
                           1
                               1
                                    1
                                        1
                                            2 1
                                                      1 ...
           333 334 335 336 337 338
             1
                      1
                 1
                           1
                               1
        [1 rows x 339 columns]
```