### **Classification Problem**

Inside this code, have

- 1. Random Forest Classifier
- 2. KN-Neighbors Classifier
- 3. Support vector machine
- 4. Decision Tree

## **Import Data and Libraries**

```
In [1]: import pandas as pd
        import numpy as np
        import seaborn as sns
        import matplotlib.pyplot as plt
        from sklearn.model selection import train test split, GridSearchCV
        from sklearn.ensemble import RandomForestClassifier
        from sklearn import preprocessing
        from sklearn.metrics import accuracy_score
        from sklearn.model selection import RandomizedSearchCV
        from scipy.stats import randint
        from IPython.display import Image
        from sklearn.preprocessing import StandardScaler
        data=pd.read_csv("data.csv")
In [3]:
        data.head()
In [4]:
```

Out[4]:	ba	aseline value	accelerations	fetal_movement	uterine_contractions	light_decelerations	severe_decelerations	prolongued_decelerations	abnormal_short_term
	0	120.0	0.000	0.0	0.000	0.000	0.0	0.0	
	1	132.0	0.006	0.0	0.006	0.003	0.0	0.0	
	2	133.0	0.003	0.0	0.008	0.003	0.0	0.0	
	3	134.0	0.003	0.0	0.008	0.003	0.0	0.0	
	4	132.0	0.007	0.0	0.008	0.000	0.0	0.0	

5 rows × 22 columns

<pre>data.nunique()</pre>		
baseline value	48	
accelerations	20	
fetal_movement	102	
uterine_contractions	16	
light_decelerations	16	
severe_decelerations	2	
<pre>prolongued_decelerations</pre>	6	
abnormal_short_term_variability	75	
<pre>mean_value_of_short_term_variability</pre>	57	
<pre>percentage_of_time_with_abnormal_long_term_variability</pre>	87	
<pre>mean_value_of_long_term_variability</pre>	249	
histogram_width	154	
histogram_min	109	
histogram_max	86	
histogram_number_of_peaks	18	
histogram_number_of_zeroes	9	
histogram_mode	88	
histogram_mean	103	
histogram_median	95	
histogram_variance	133	
histogram_tendency	3	
fetal_health	3	
dtype: int64		

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2126 entries, 0 to 2125
Data columns (total 22 columns):
     Column
                                                            Non-Null Count Dtype
    -----
    baseline value
                                                            2126 non-null float64
     accelerations
                                                            2126 non-null
                                                                            float64
    fetal movement
                                                            2126 non-null
                                                                            float64
    uterine contractions
                                                            2126 non-null
                                                                            float64
    light decelerations
                                                                            float64
                                                            2126 non-null
    severe decelerations
                                                            2126 non-null
                                                                            float64
    prolongued decelerations
                                                            2126 non-null
                                                                            float64
    abnormal short term variability
                                                            2126 non-null
                                                                            float64
    mean_value_of_short_term_variability
                                                            2126 non-null
                                                                            float64
    percentage of time with abnormal long term variability
                                                            2126 non-null
                                                                            float64
    mean value of long term variability
                                                                            float64
                                                            2126 non-null
    histogram width
                                                            2126 non-null
                                                                            float64
    histogram min
                                                            2126 non-null
                                                                            float64
13
    histogram max
                                                            2126 non-null
                                                                            float64
14 histogram number of peaks
                                                            2126 non-null
                                                                            float64
15 histogram number of zeroes
                                                            2126 non-null
                                                                            float64
16 histogram mode
                                                            2126 non-null
                                                                            float64
17
    histogram mean
                                                            2126 non-null
                                                                            float64
18 histogram median
                                                            2126 non-null
                                                                            float64
    histogram variance
                                                            2126 non-null
                                                                            float64
20 histogram_tendency
                                                            2126 non-null
                                                                            float64
21 fetal health
                                                            2126 non-null
                                                                            float64
dtypes: float64(22)
memory usage: 365.5 KB
```

In [7]: data.isnull().sum()

```
baseline value
                                                                   0
Out[7]:
        accelerations
                                                                   0
        fetal movement
        uterine contractions
        light decelerations
        severe decelerations
        prolongued decelerations
        abnormal short term variability
        mean value of short term variability
        percentage_of_time_with_abnormal_long_term_variability
        mean value of long term variability
        histogram width
        histogram min
        histogram_max
        histogram number of peaks
        histogram number of zeroes
        histogram mode
        histogram mean
        histogram median
        histogram variance
        histogram tendency
        fetal health
        dtype: int64
```

# **Exploratory Data Analysis**

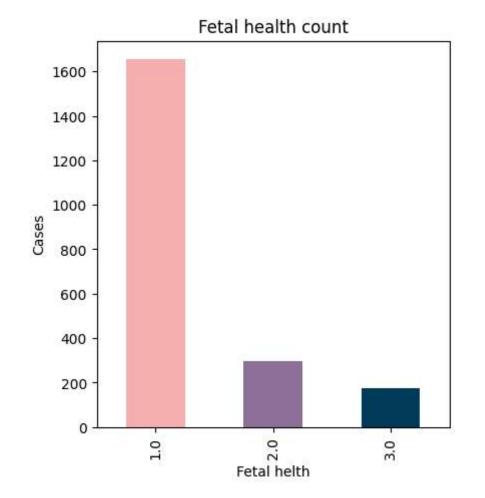
```
In [10]:
    total = data["fetal_health"].sum()
    normal = total - 471
    suspect = total - 1831
    pathological = total - 1950

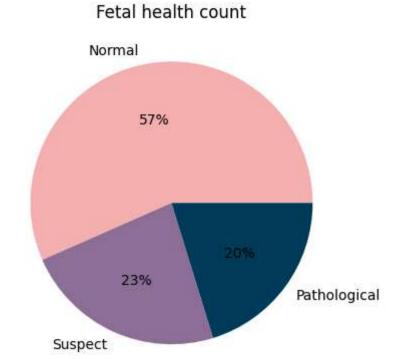
    plt.figure(figsize = (14,8))
    plt.subplot(121)
    vis_fetal_health = data.fetal_health.value_counts().plot(figsize=(10, 5), kind="bar", color = colours)
    plt.title("Fetal health count")
    plt.xlabel("Fetal health")
    plt.ylabel("Cases")

plt.subplot(122)
    plt.title("Fetal state")

vis_pie_fetal_health = plt.pie([normal, suspect, pathological], labels=["Normal", "Suspect", "Pathological"], colors = colours, a
    plt.title("Fetal health count")

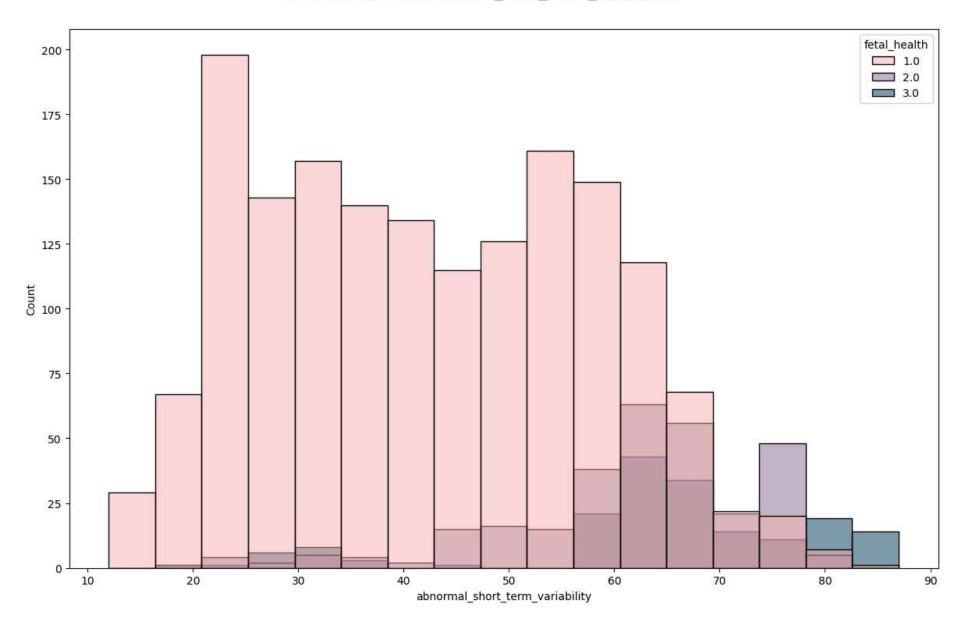
plt.show()
```





```
In [11]: plt.figure(figsize=(12,8))
    sns.histplot(data=data, x='abnormal_short_term_variability', hue='fetal_health', palette=colours)
    plt.suptitle('Distribution of Abnormal_hort_term_variability', size=16, y=1.01)
    plt.tight_layout()
    plt.show()
```

### Distribution of Abnormal\_hort\_term\_variability



# **Model Building**

```
In [12]: data.head()

Out[12]: baseline accolerations fotal movement utoring contractions light decelerations according approximate the property of the contractions and the contractions are contractions as a contraction of the contractions are contractions.
```

]:		baseline value	accelerations	fetal_movement	uterine_contractions	light_decelerations	severe_decelerations	prolongued_decelerations	abnormal_short_term
	0	120.0	0.000	0.0	0.000	0.000	0.0	0.0	
	1	132.0	0.006	0.0	0.006	0.003	0.0	0.0	
	2	133.0	0.003	0.0	0.008	0.003	0.0	0.0	
	3	134.0	0.003	0.0	0.008	0.003	0.0	0.0	
	4	132.0	0.007	0.0	0.008	0.000	0.0	0.0	

5 rows × 22 columns

```
In [13]: #assigning values to features as X and target as y
X=data.drop(["fetal_health"],axis=1)
y=data["fetal_health"]

#Set up a standard scaler for the features
col_names = list(X.columns)
s_scaler = preprocessing.StandardScaler()
X_df = s_scaler.fit_transform(X)
X_df = pd.DataFrame(X_df, columns=col_names)
In [14]: X
```

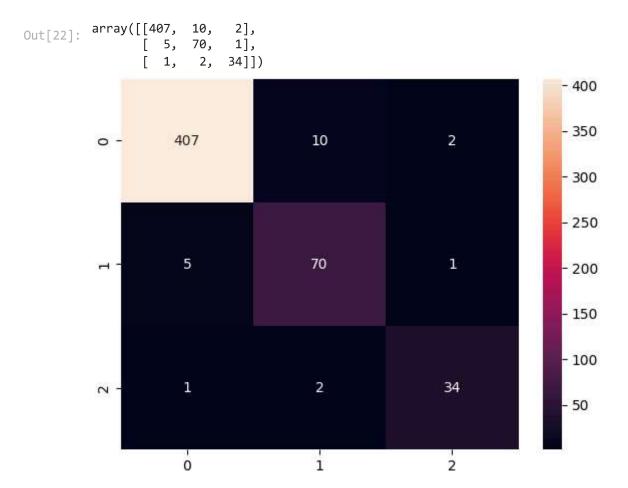
it[14]:		baseline value	accelerations	fetal_movement	uterine_contractions	light_decelerations	severe_decelerations	prolongued_decelerations	abnormal_short_te
	0	120.0	0.000	0.000	0.000	0.000	0.0	0.0	
	1	132.0	0.006	0.000	0.006	0.003	0.0	0.0	
	2	133.0	0.003	0.000	0.008	0.003	0.0	0.0	
	3	134.0	0.003	0.000	0.008	0.003	0.0	0.0	
	4	132.0	0.007	0.000	0.008	0.000	0.0	0.0	
	•••		•••						
	2121	140.0	0.000	0.000	0.007	0.000	0.0	0.0	
	2122	140.0	0.001	0.000	0.007	0.000	0.0	0.0	
	2123	140.0	0.001	0.000	0.007	0.000	0.0	0.0	
	2124	140.0	0.001	0.000	0.006	0.000	0.0	0.0	
	2125	142.0	0.002	0.002	0.008	0.000	0.0	0.0	

2126 rows × 21 columns

```
In [15]:
         print(y)
                 2.0
                 1.0
                 1.0
                 1.0
                 1.0
                 . . .
         2121
                 2.0
         2122
                 2.0
         2123
                 2.0
         2124
                 2.0
         2125
                 1.0
         Name: fetal_health, Length: 2126, dtype: float64
In [16]: X_train, X_test, y_train,y_test = train_test_split(X_df,y,test_size=0.25,random_state=42)
```

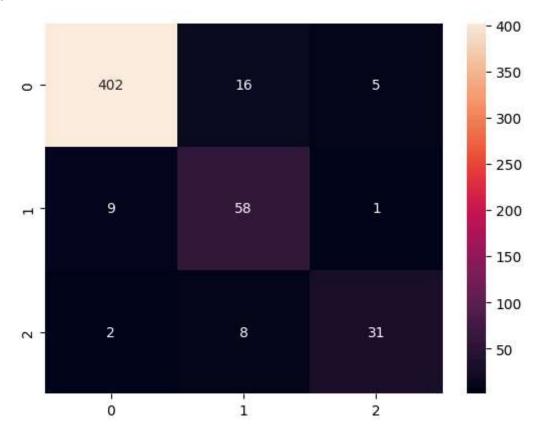
#### RandomForestClassifier

```
In [17]: from sklearn.ensemble import RandomForestClassifier
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.tree import DecisionTreeClassifier
         from sklearn import svm
         from sklearn.metrics import confusion matrix
         model = RandomForestClassifier()
In [18]:
         model.fit(X train, y train)
         ▼ RandomForestClassifier
Out[18]:
         RandomForestClassifier()
In [19]: model_train_prediction = model.predict(X train)
         training_data_accuracy = accuracy_score(y_train, model_train_prediction)
         print('Accuracy score of training data : ', training_data_accuracy)
         Accuracy score of training data : 1.0
In [20]: model_test_prediction = model.predict(X test)
         test data accuracy = accuracy score(y test, model test prediction)
         print('Accuracy score of test data : ', test data accuracy)
         Accuracy score of test data : 0.9605263157894737
In [40]: comp = pd.DataFrame(
             {"Y_pred":model_test_prediction,
              "Y_test":y_test}
         comp.to csv('file1.csv')
In [22]: cm=confusion matrix(model.predict(X test),y test)
         sns.heatmap(cm,annot=True, fmt='g')
          cm
```



### **KN-Neighbors Classifier**

Out[27]: <Axes: >



In [27]:

### Support vector machine

```
In [28]: model 3 = svm.SVC(kernel='rbf')
In [29]: # training the SVM model with training data
         model 3.fit(X train,y train)
Out[29]: ▼ SVC
         SVC()
         model 3 train score=model 3.score(X train,y train)
         print('Accuracy score of training data : ',model 3 train score)
         Accuracy score of training data: 0.9316185696361355
         modul 3 pred=model 3.predict(X test)
In [31]:
         svm pred score=accuracy score(y test,modul 3 pred)
In [32]:
         print('Accuracy score of test data : ',svm pred score)
         Accuracy score of test data: 0.9172932330827067
         Decision Tree
In [33]: model_4 = DecisionTreeClassifier()
In [34]: model_4.fit(X_train,y_train)
Out[34]: ▼ DecisionTreeClassifier
         DecisionTreeClassifier()
In [35]: model 4 train score=model 4.score(X train,y train)
         print('Accuracy score of training data : ',model 4 train score)
         Accuracy score of training data: 1.0
```

```
In [36]: modul_4_pred=model_4.predict(X_test)
          dt_pred_score=accuracy_score(y_test,modul_4_pred)
In [37]:
          print('Accuracy score of test data : ',dt pred score)
          Accuracy score of test data : 0.9323308270676691
          Results
          result=pd.DataFrame(columns=['Algorithm','Training-Result','Predictive-Result'])
In [38]:
In [39]: m=['KN-Neighbors Classifier', model 2 train score, knn pred score]
          result.loc[0]=m
          m = ['Random Forest', training_data_accuracy, test_data_accuracy]
          result.loc[1]=m
          m = ['Support Vector Machine', model_3_train_score, svm_pred_score]
          result.loc[2]=m
          m = ['Decision Tree Classifier', model 4 train score, dt pred score]
          result.loc[3]=m
          result
Out[39]:
                       Algorithm Training-Result Predictive-Result
          0 KN-Neighbors Classifier
                                      0.948557
                                                      0.922932
                   Random Forest
                                      1.000000
                                                      0.960526
          2 Support Vector Machine
                                      0.931619
                                                      0.917293
             Decision Tree Classifier
                                      1.000000
                                                      0.932331
In [39]:
```