```
# Importing Libraries
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import seaborn as sns
import matplotlib.pyplot as plt
```

import warnings
warnings.filterwarnings("ignore")

Importing Dataset
df_train=pd.read_csv("/content/train.csv")
df_test=pd.read_csv("/content/test.csv")
df_train.head()

	Patient Id	Patient Age	Genes in mother's side	Inherited from father	Maternal gene	Paternal gene	Blood cell count (mcL)	Patient First Name	Family Name	Fath∈ r		
0	PID0x6418	2.0	Yes	No	Yes	No	4.760603	Richard	NaN	L		
1	PID0x25d5	4.0	Yes	Yes	No	No	4.910669	Mike	NaN	Bry		
2	PID0x4a82	6.0	Yes	No	No	No	4.893297	Kimberly	NaN	Nas		
3	PID0x4ac8	12.0	Yes	No	Yes	No	4.705280	Jeffery	Hoelscher	Aay		
4	PID0x1bf7	11.0	Yes	No	NaN	Yes	4.720703	Johanna	Stutzman	Sı		
5 rows × 45 columns												

df_train.tail()

		Patient Id	Patient Age	Genes in mother's side	Inherited from father	Maternal gene	Paternal gene	Blood cell count (mcL)	Patient First Name	Family Name	Fat	
	22078 PID0x559		4.0	Yes	Yes	Yes	No	5.258298	Lynn	NaN	Alh	
	22079	PID0x19cb	8.0	No	Yes	No	Yes	4.974220	Matthew	Farley	Da	
	22080	PID0x3c4f	8.0	Yes	No	Yes	No	5.186470	John	NaN		
	22084 al Colu ain.col		7 0	Vac	No	Vac	Vac	<i>ለ</i>	Sharon	NeN		
	<pre>Index(['Patient Id', 'Patient Age', 'Genes in mother's side',</pre>											
df_tr df_tr df_tr df_tr df_tr df_tr	ain.dro ain.dro ain.dro ain.dro ain.dro ain.dro	pp("Father's pp("Institut	Id",axis Name",axi First Na s name",a te Name", n of Inst	s=1,inplac me",axis=1 xis=1,inpl axis=1,inp itute",axi	e=True) ,inplace=Tr ace=True) lace=True) s=1,inplace	·						
	_	otal null v a().sum()	/alues									
		: Age .n mother's ed from fat				1427 0 306						

```
Maternal gene
                                                      2810
Paternal gene
                                                         0
                                                         0
Blood cell count (mcL)
Mother's age
                                                      6036
Father's age
                                                      5986
Status
                                                         0
                                                      2149
Respiratory Rate (breaths/min)
Heart Rate (rates/min
                                                      2113
Test 1
                                                      2127
Test 2
                                                      2152
Test 3
                                                      2147
Test 4
                                                      2140
Test 5
                                                      2170
Parental consent
                                                      2125
                                                      2166
Follow-up
Gender
                                                      2173
Birth asphyxia
                                                      2139
Autopsy shows birth defect (if applicable)
                                                      1026
Folic acid details (peri-conceptional)
                                                      2117
H/O serious maternal illness
                                                      2152
H/O radiation exposure (x-ray)
                                                      2153
H/O substance abuse
                                                      2195
Assisted conception IVF/ART
                                                      2122
History of anomalies in previous pregnancies
                                                      2172
No. of previous abortion
                                                      2162
Birth defects
                                                      2154
White Blood cell count (thousand per microliter)
                                                      2148
Blood test result
                                                      2145
Symptom 1
                                                      2155
Symptom 2
                                                      2222
Symptom 3
                                                      2101
Symptom 4
                                                      2113
Symptom 5
                                                      2153
Genetic Disorder
                                                      2146
Disorder Subclass
                                                      2168
dtype: int64
```

df_train["Patient Age"]

```
0
           2.0
1
          4.0
2
          6.0
3
         12.0
4
         11.0
22078
          4.0
22079
          8.0
22080
          8.0
22081
          7.0
22082
         11.0
```

Name: Patient Age, Length: 22083, dtype: float64

```
# Filling Null values with mode
df_train["Patient Age"].fillna(str(df_train["Patient Age"].mode().values[0]),inplace=True)
df train["Inherited from father"].fillna(str(df train["Inherited from father"].mode().values[0]),inpl
df train["Maternal gene"].fillna(str(df train["Maternal gene"].mode().values[0]),inplace=True)
df_train["Mother's age"].fillna(str(df_train["Mother's age"].mode().values[0]),inplace=True)
df_train["Father's age"].fillna(str(df_train["Father's age"].mode().values[0]),inplace=True)
df train["Respiratory Rate (breaths/min)"].fillna(str(df train["Respiratory Rate (breaths/min)"].mode
df_train["Heart Rate (rates/min"].fillna(str(df_train["Heart Rate (rates/min"].mode().values[0]),inpl
df_train["Test 1"].fillna(str(df_train["Test 1"].mode().values[0]),inplace=True)
df train["Test 2"].fillna(str(df train["Test 2"].mode().values[0]),inplace=True)
df_train["Test 3"].fillna(str(df_train["Test 3"].mode().values[0]),inplace=True)
df_train["Test 4"].fillna(str(df_train["Test 4"].mode().values[0]),inplace=True)
df_train["Test 5"].fillna(str(df_train["Test 5"].mode().values[0]),inplace=True)
df_train["Parental consent"].fillna(str(df_train["Parental consent"].mode().values[0]),inplace=True)
df_train["Follow-up"].fillna(str(df_train["Follow-up"].mode().values[0]),inplace=True)
df_train["Gender"].fillna(str(df_train["Gender"].mode().values[0]),inplace=True)
df_train["Birth asphyxia"].fillna(str(df_train["Birth asphyxia"].mode().values[0]),inplace=True)
df_train["Autopsy shows birth defect (if applicable)"].fillna(str(df_train["Autopsy shows birth defec
df_train["Folic acid details (peri-conceptional)"].fillna(str(df_train["Folic acid details (peri-conc
df_train["H/O serious maternal illness"].fillna(str(df_train["H/O serious maternal illness"].mode().v
df_train["H/O radiation exposure (x-ray)"].fillna(str(df_train["H/O radiation exposure (x-ray)"].mode
df_train["H/O substance abuse"].fillna(str(df_train["H/O substance abuse"].mode().values[0]),inplace=
df train["Assisted conception IVF/ART"].fillna(str(df train["Assisted conception IVF/ART"].mode().val
df_train["History of anomalies in previous pregnancies"].fillna(str(df_train["History of anomalies ir
df_train["No. of previous abortion"].fillna(str(df_train["No. of previous abortion"].mode().values[0]
df_train["Birth defects"].fillna(str(df_train["Birth defects"].mode().values[0]),inplace=True)
df train["White Blood cell count (thousand per microliter)"].fillna(str(df train["White Blood cell cc
df_train["Blood test result"].fillna(str(df_train["Blood test result"].mode().values[0]),inplace=True
df_train["Symptom 1"].fillna(str(df_train["Symptom 1"].mode().values[0]),inplace=True)
df_train["Symptom 2"].fillna(str(df_train["Symptom 2"].mode().values[0]),inplace=True)
df_train["Symptom 3"].fillna(str(df_train["Symptom 3"].mode().values[0]),inplace=True)
df_train["Symptom 4"].fillna(str(df_train["Symptom 4"].mode().values[0]),inplace=True)
df_train["Symptom 5"].fillna(str(df_train["Symptom 5"].mode().values[0]),inplace=True)
df_train["Genetic Disorder"].fillna(str(df_train["Genetic Disorder"].mode().values[0]),inplace=True)
df_train["Disorder Subclass"].fillna(str(df_train["Disorder Subclass"].mode().values[0]),inplace=True
```

Checking if any null value is present df train.isna().sum()

Patient Age	0
Genes in mother's side	0
Inherited from father	0
Maternal gene	0
Paternal gene	0
Blood cell count (mcL)	0
Mother's age	0
Father's age	0
Status	0
Respiratory Rate (breaths/min)	0
Heart Rate (rates/min	0
Test 1	0
Test 2	0
Test 3	0
Test 4	0
Test 5	0
Parental consent	0
Follow-up	0

Gender	0				
Birth asphyxia	0				
Autopsy shows birth defect (if applicable)	0				
Folic acid details (peri-conceptional)	0				
H/O serious maternal illness					
H/O radiation exposure (x-ray)	0				
H/O substance abuse	0				
Assisted conception IVF/ART	0				
History of anomalies in previous pregnancies					
No. of previous abortion	0				
Birth defects	0				
White Blood cell count (thousand per microliter)	0				
Blood test result	0				
Symptom 1	0				
Symptom 2	0				
Symptom 3	0				
Symptom 4	0				
Symptom 5	0				
Genetic Disorder	0				
Disorder Subclass	0				
dtype: int64					

df_train.head()

	Patient Age	Genes in mother's side	Inherited from father	Maternal gene	Paternal gene	Blood cell count (mcL)	Mother's age	Father's age	Sta
0	2.0	Yes	No	Yes	No	4.760603	23.0	20.0	A
1	4.0	Yes	Yes	No	No	4.910669	23.0	23.0	Decea
2	6.0	Yes	No	No	No	4.893297	41.0	22.0	А
3	12.0	Yes	No	Yes	No	4.705280	21.0	20.0	Decea
4	11.0	Yes	No	Yes	Yes	4.720703	32.0	20.0	A
5 ro	ws × 38 cc	olumns							

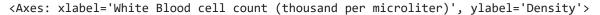
df_train.info()

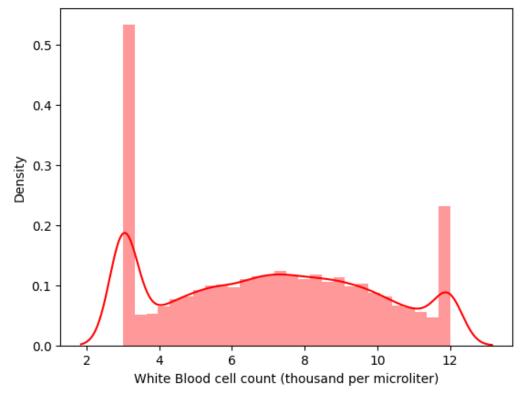
#

```
<class 'pandas.core.frame.DataFrame'>
    RangeIndex: 22083 entries, 0 to 22082
    Data columns (total 38 columns):
     # Column
                                                         Non-Null Count Dtype
                                                         -----
     0
         Patient Age
                                                         22083 non-null object
         Genes in mother's side
                                                         22083 non-null object
     1
                                                         22083 non-null object
         Inherited from father
                                                         22083 non-null object
     3
         Maternal gene
     4
         Paternal gene
                                                         22083 non-null object
     5
         Blood cell count (mcL)
                                                         22083 non-null float64
     6
                                                         22083 non-null object
         Mother's age
     7
         Father's age
                                                         22083 non-null object
     8
         Status
                                                         22083 non-null object
     9
         Respiratory Rate (breaths/min)
                                                         22083 non-null object
                                                         22083 non-null object
     10 Heart Rate (rates/min
                                                         22083 non-null object
     11 Test 1
     12 Test 2
                                                         22083 non-null object
     13 Test 3
                                                         22083 non-null object
     14 Test 4
                                                         22083 non-null object
     15 Test 5
                                                         22083 non-null object
     16 Parental consent
                                                         22083 non-null object
     17 Follow-up
                                                         22083 non-null object
     18 Gender
                                                         22083 non-null object
     19 Birth asphyxia
                                                         22083 non-null object
     20 Autopsy shows birth defect (if applicable)
                                                         22083 non-null object
     21 Folic acid details (peri-conceptional)
                                                         22083 non-null object
     22 H/O serious maternal illness
                                                         22083 non-null object
     23 H/O radiation exposure (x-ray)
                                                         22083 non-null object
     24 H/O substance abuse
                                                         22083 non-null object
     25 Assisted conception IVF/ART
                                                         22083 non-null object
     26 History of anomalies in previous pregnancies
                                                         22083 non-null object
     27 No. of previous abortion
                                                         22083 non-null object
     28 Birth defects
                                                         22083 non-null object
     29 White Blood cell count (thousand per microliter) 22083 non-null object
     30 Blood test result
                                                         22083 non-null object
     31 Symptom 1
                                                         22083 non-null object
     32 Symptom 2
                                                         22083 non-null object
     33 Symptom 3
                                                         22083 non-null object
     34 Symptom 4
                                                         22083 non-null object
     35 Symptom 5
                                                         22083 non-null object
     36 Genetic Disorder
                                                         22083 non-null object
     37 Disorder Subclass
                                                         22083 non-null object
    dtypes: float64(1), object(37)
    memory usage: 6.4+ MB
# Optional Column name change
# for column in df train:
     columnSeriesObj = df train[column]
     print('Column Name : ', column)
     print('Column Contents : ', columnSeriesObj.values)
     print("----")
df_train.columns
    Index(['Patient Age', 'Genes in mother's side', 'Inherited from father',
           'Maternal gene', 'Paternal gene', 'Blood cell count (mcL)',
           'Mother's age', 'Father's age', 'Status',
           'Respiratory Rate (breaths/min)', 'Heart Rate (rates/min', 'Test 1',
```

```
'Test 2', 'Test 3', 'Test 4', 'Test 5', 'Parental consent', 'Follow-up', 'Gender', 'Birth asphyxia',
'Autopsy shows birth defect (if applicable)',
'Folic acid details (peri-conceptional)',
'H/O serious maternal illness', 'H/O radiation exposure (x-ray)',
'H/O substance abuse', 'Assisted conception IVF/ART',
'History of anomalies in previous pregnancies',
'No. of previous abortion', 'Birth defects',
'White Blood cell count (thousand per microliter)', 'Blood test result',
'Symptom 1', 'Symptom 2', 'Symptom 3', 'Symptom 4', 'Symptom 5',
'Genetic Disorder', 'Disorder Subclass'],
dtype='object')
```

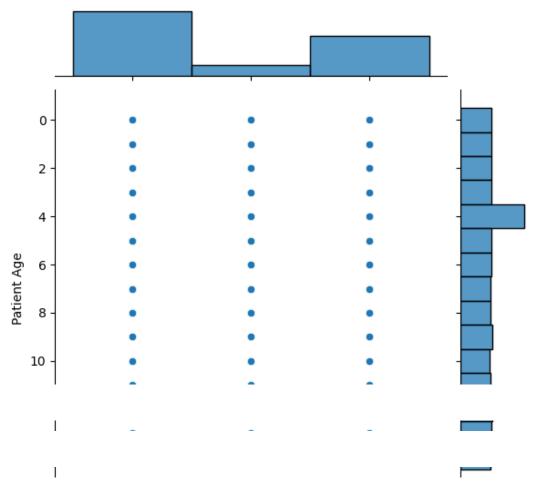
sns.distplot(df_train["White Blood cell count (thousand per microliter)"],color = "red")



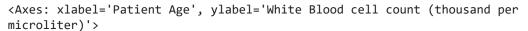


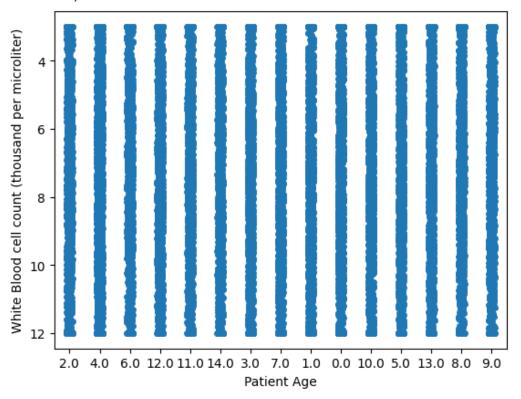
sns.jointplot(x="Genetic Disorder",y="Patient Age",data=df_train)

<seaborn.axisgrid.JointGrid at 0x7e15b4cf4b20>



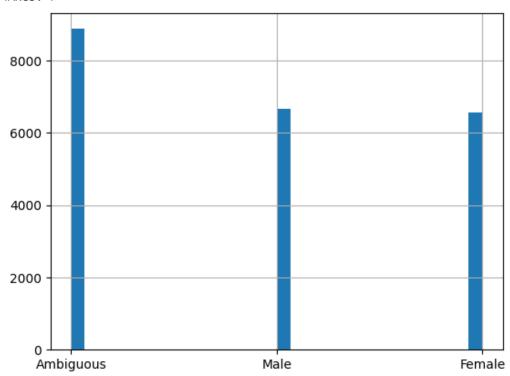
sns.stripplot(x="Patient Age",y="White Blood cell count (thousand per microliter)",data=df_train,jitt





df_train["Gender"].hist(bins=30)





Changing from yes or no to numerical(1 or 0)

df_train["Genes in mother's side"]=[1 if i.strip()== "Yes" else 0 for i in df_train["Genes in mother'

df_train["Inherited from father"]=[1 if i.strip()== "Yes" else 0 for i in df_train["Inherited from fa

df_train["Maternal gene"]=[1 if i.strip()== "Yes" else 0 for i in df_train["Maternal gene"]]

df_train["Paternal gene"]=[1 if i.strip()== "Yes" else 0 for i in df_train["Paternal gene"]]

df_train["Parental consent"]=[1 if i.strip()== "Yes" else 0 for i in df_train["Parental consent"]]

df_train["Birth asphyxia"]=[1 if i.strip()== "Yes" else 0 for i in df_train["Birth asphyxia"]]

df_train["Folic acid details (peri-conceptional)"]=[1 if i.strip()== "Yes" else 0 for i in df_train["H/O radi

df_train["H/O substance abuse"]=[1 if i.strip()== "Yes" else 0 for i in df_train["H/O substance abuse

df_train["Assisted conception IVF/ART"]=[1 if i.strip()== "Yes" else 0 for i in df_train["Assisted co

df_train["History of anomalies in previous pregnancies"]=[1 if i.strip()== "Yes" else 0 for i in df_train["H/O serious

df_train.head()

```
Blood
                 Genes in Inherited
        Patient
                                      Maternal Paternal
                                                             cell Mother's Father's
                 mother's
                                from
                                                                                        Sta
            Age
                                                            count
                                          gene
                                                   gene
                                                                       age
                                                                                 age
                              father
                     side
                                                            (mcL)
     0
             2.0
                                   0
                                             1
                                                      0 4.760603
                                                                       23.0
                                                                                 20.0
                                                                                          Α
                                            0
     1
             4.0
                        1
                                   1
                                                      0 4.910669
                                                                       23.0
                                                                                 23.0 Decea
     2
             6.0
                        1
                                   0
                                            0
                                                      0 4.893297
                                                                       41.0
                                                                                 22.0
                                                                                          Α
# Check if you changed the column name
# for column in df train:
     columnSeriesObj = df_train[column]
     print('Column Name : ', column)
     print('Column Contents : ', columnSeriesObj.values)
#
     print("-----")
#
# Checking the unique elements in Categorical Columns
print("Status: ",df train["Status"].unique())
print("Respiratory Rate (breaths/min): ",df_train["Respiratory Rate (breaths/min)"].unique())
print("Heart Rate (rates/min: ",df train["Heart Rate (rates/min"].unique())
print("Follow-up: ",df_train["Follow-up"].unique())
print("Gender: ",df_train["Gender"].unique())
print("Autopsy shows birth defect (if applicable): ",df_train["Autopsy shows birth defect (if applica
print("Birth defects: ",df_train["Birth defects"].unique())
print("Blood test result: ",df_train["Blood test result"].unique())
print("Genetic Disorder: ",df_train["Genetic Disorder"].unique())
print("Disorder Subclass: ",df train["Disorder Subclass"].unique())
     Status: ['Alive' 'Deceased']
     Respiratory Rate (breaths/min): ['Normal (30-60)' 'Tachypnea']
     Heart Rate (rates/min: ['Normal' 'Tachycardia']
     Follow-up: ['High' 'Low']
     Gender: ['Ambiguous' 'Male' 'Female']
     Autopsy shows birth defect (if applicable): ['Not applicable' 'None' 'No' 'Yes']
     Birth defects: ['Singular' 'Multiple']
     Blood test result: ['slightly abnormal' 'normal' 'inconclusive' 'abnormal']
     Genetic Disorder: ['Mitochondrial genetic inheritance disorders'
      'Multifactorial genetic inheritance disorders'
      'Single-gene inheritance diseases']
     Disorder Subclass: ["Leber's hereditary optic neuropathy" 'Cystic fibrosis' 'Diabetes'
      'Leigh syndrome' 'Cancer' 'Tay-Sachs' 'Hemochromatosis'
      'Mitochondrial myopathy' "Alzheimer's"]
# plots
```

df train.head()

	Patient Age	Genes in mother's side	Inherited from father	Maternal gene	Paternal gene	Blood cell count (mcL)	Mother's age	Father's age	Status	Res (brea	
0	2.0	1	0	1	0	4.760603	23.0	20.0	Alive	Norm	
1	4.0	1	1	0	0	4.910669	23.0	23.0	Deceased	٦	
2	6.0	1	0	0	0	4.893297	41.0	22.0	Alive	Norm	
3	12.0	1	0	1	0	4.705280	21.0	20.0	Deceased	٦	
4	11.0	1	0	1	1	4.720703	32.0	20.0	Alive	7	
5 rows × 38 columns											

```
# Changing Categorical Values to Numerical Values
#Alive':1 'Deceased:0'
df train["Status"]=[1 if i.strip()== "Alive" else 0 for i in df train["Status"]]
#Normal (30-60):1' 'Tachypnea:0
df_train["Respiratory Rate (breaths/min)"]=[1 if i.strip()== "Normal (30-60)" else 0 for i in df_trai
#Normal:1' 'Tachycardia:0
df train["Heart Rate (rates/min"]=[1 if i.strip()== "Normal" else 0 for i in df train["Heart Rate (ra
#High:1, Low:0
df_train["Follow-up"]=[1 if i.strip()== "High" else 0 for i in df_train["Follow-up"]]
#['Singular' 'Multiple']
df_train["Birth defects"]=[1 if i.strip()== "Singular" else 0 for i in df_train["Birth defects"]]
#1: male 0: female 2: ambiguous
df train["Gender"]=[1 if i.strip()== "Male" else 0 if i.strip() == "Female" else 2 for i in df_train[
#Not applicable:3' 'None:2' 'No:0' 'Yes:1'
df train["Autopsy shows birth defect (if applicable)"]=[1 if i.strip()== "Yes" else 0 if i.strip() ==
#'slightly abnormal':1, 'normal':0, 'inconclusive':2 'abnormal:3']
df_train["Blood test result"]=[1 if i.strip()== "slightly abnormal" else 0 if i.strip() == "normal" ε
#'Mitochondrial genetic inheritance disorders':1,'Multifactorial genetic inheritance disorders':0'Sir
df_train["Genetic Disorder"]=[1 if i.strip()== "Mitochondrial genetic inheritance disorders" else 0 i
#Leber's hereditary optic neuropathy:1
#Cvstic fibrosis:0
#Diabetes:2
#Leigh syndrome:3
#Cancer:4
#Tay-Sachs:5
#Hemochromatosis:6
#Mitochondrial myopathy:7
#Alzheimer's:8
df train["Disorder Subclass"]=[1 if i.strip()== "Leber's hereditary optic neuropathy"
                                else 0 if i.strip() == "Cystic fibrosis"
                               else 2 if i.strip()=="Diabetes"
                               else 3 if i.strip()=="Leigh syndrome"
                               else 4 if i.strip()=="Cancer"
                               else 5 if i.strip()=="Tay-Sachs"
                               else 6 if i.strip()=="Hemochromatosis"
                               else 7 if i.strip()=="Mitochondrial myopathy"
                               else 8 for i in df train["Disorder Subclass"]]
```

df train.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 22083 entries, 0 to 22082
Data columns (total 38 columns):
 # Column

#	Column	Non-Null Count	Dtype
0	Patient Age	22083 non-null	object
1	Genes in mother's side	22083 non-null	int64
2	Inherited from father	22083 non-null	int64
3	Maternal gene	22083 non-null	int64
4	Paternal gene	22083 non-null	int64
5	Blood cell count (mcL)	22083 non-null	float64
6	Mother's age	22083 non-null	object
7	Father's age	22083 non-null	object
8	Status	22083 non-null	int64
9	Respiratory Rate (breaths/min)	22083 non-null	int64

```
10 Heart Rate (rates/min
                                                           22083 non-null int64
     11 Test 1
                                                           22083 non-null object
     12 Test 2
                                                           22083 non-null object
                                                           22083 non-null object
     13 Test 3
     14 Test 4
                                                           22083 non-null object
                                                           22083 non-null object
     15 Test 5
     16 Parental consent
                                                           22083 non-null int64
                                                           22083 non-null int64
     17 Follow-up
                                                           22083 non-null int64
     18 Gender
     19 Birth asphyxia
                                                           22083 non-null int64
                                                           22083 non-null int64
     20 Autopsy shows birth defect (if applicable)
     21 Folic acid details (peri-conceptional)
                                                           22083 non-null int64
     22 H/O serious maternal illness
                                                           22083 non-null int64
     23 H/O radiation exposure (x-ray)
                                                           22083 non-null int64
     24 H/O substance abuse
                                                           22083 non-null int64
                                                           22083 non-null int64
     25 Assisted conception IVF/ART
     26 History of anomalies in previous pregnancies
                                                           22083 non-null int64
     27 No. of previous abortion
                                                           22083 non-null object
     28 Birth defects
                                                           22083 non-null int64
         White Blood cell count (thousand per microliter) 22083 non-null object
                                                           22083 non-null int64
     30 Blood test result
     31 Symptom 1
                                                           22083 non-null object
     32 Symptom 2
                                                           22083 non-null object
     33 Symptom 3
                                                           22083 non-null object
     34 Symptom 4
                                                           22083 non-null object
     35 Symptom 5
                                                           22083 non-null object
     36 Genetic Disorder
                                                           22083 non-null int64
     37 Disorder Subclass
                                                           22083 non-null int64
     dtypes: float64(1), int64(22), object(15)
     memory usage: 6.4+ MB
# Changing the datatype to float
df_train = df_train.apply(pd.to_numeric,downcast="float")
#total symptom
df_train["total symptom"]=(df_train["Symptom 1"]+df_train["Symptom 2"]+df_train["Symptom 3"]+df_trair
df train["sum of Mother's and fathers age avg"]=(df train["Mother's age"]+df train["Father's age"]) /
#Dropping Symptom Columns
df_train.drop(["Symptom 1", "Symptom 2", "Symptom 3", "Symptom 4", "Symptom 5"], axis=1, inplace=True)
df train.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 22083 entries, 0 to 22082
     Data columns (total 35 columns):
     # Column
                                                           Non-Null Count Dtype
         -----
                                                           -----
         Patient Age
                                                           22083 non-null float32
         Genes in mother's side
                                                           22083 non-null float32
     1
     2
         Inherited from father
                                                           22083 non-null float32
     3
         Maternal gene
                                                           22083 non-null float32
                                                           22083 non-null float32
     4
         Paternal gene
     5
         Blood cell count (mcL)
                                                           22083 non-null float32
                                                           22083 non-null float32
     6
         Mother's age
     7
         Father's age
                                                           22083 non-null float32
                                                           22083 non-null float32
     8
         Status
```

9	Respiratory Rate (breaths/min)	22083	non-null	float32
10	Heart Rate (rates/min	22083	non-null	float32
11	Test 1	22083	non-null	float32
12	Test 2	22083	non-null	float32
13	Test 3	22083	non-null	float32
14	Test 4	22083	non-null	float32
15	Test 5	22083	non-null	float32
16	Parental consent	22083	non-null	float32
17	Follow-up	22083	non-null	float32
18	Gender	22083	non-null	float32
19	Birth asphyxia	22083	non-null	float32
20	Autopsy shows birth defect (if applicable)	22083	non-null	float32
21	Folic acid details (peri-conceptional)	22083	non-null	float32
22	H/O serious maternal illness	22083	non-null	float32
23	H/O radiation exposure (x-ray)	22083	non-null	float32
24	H/O substance abuse	22083	non-null	float32
25	Assisted conception IVF/ART	22083	non-null	float32
26	History of anomalies in previous pregnancies	22083	non-null	float32
27	No. of previous abortion	22083	non-null	float32
28	Birth defects	22083	non-null	float32
29	White Blood cell count (thousand per microliter)	22083	non-null	float32
30	Blood test result	22083	non-null	float32
31	Genetic Disorder	22083	non-null	float32
32	Disorder Subclass	22083	non-null	float32
33	total symptom	22083	non-null	float32
34	sum of Mother's and fathers age avg	22083	non-null	float32
dt vn	es: float32(35)			

dtypes: float32(35)
memory usage: 2.9 MB

df_train.head()

	Patient Age	Genes in mother's side	Inherited from father	Maternal gene	Paternal gene	Blood cell count (mcL)	Mother's age	Father's age	Statu
0	2.0	1.0	0.0	1.0	0.0	4.760603	23.0	20.0	1.
1	4.0	1.0	1.0	0.0	0.0	4.910669	23.0	23.0	0.
2	6.0	1.0	0.0	0.0	0.0	4.893297	41.0	22.0	1.
3	12.0	1.0	0.0	1.0	0.0	4.705280	21.0	20.0	0.
4	11.0	1.0	0.0	1.0	1.0	4.720703	32.0	20.0	1.

5 rows × 35 columns

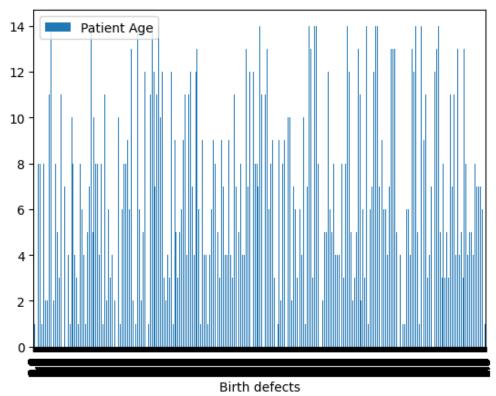
df_train.Status.value_counts()

1.0 11083 0.0 11000

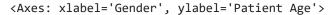
Name: Status, dtype: int64

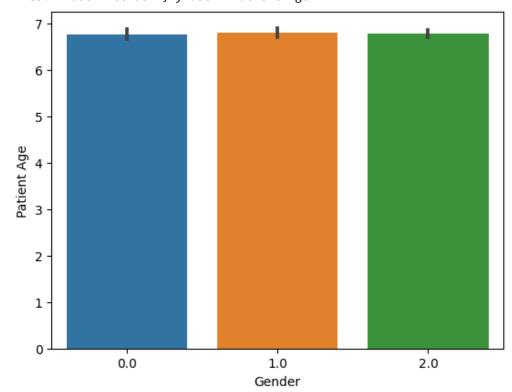
df_train.plot.bar(y="Patient Age",x="Birth defects")

<Axes: xlabel='Birth defects'>



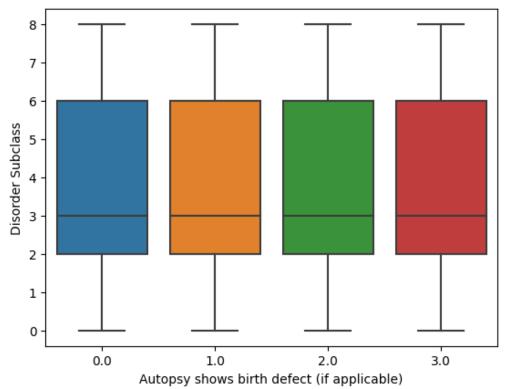
sns.barplot(x="Gender",y="Patient Age",data=df_train)





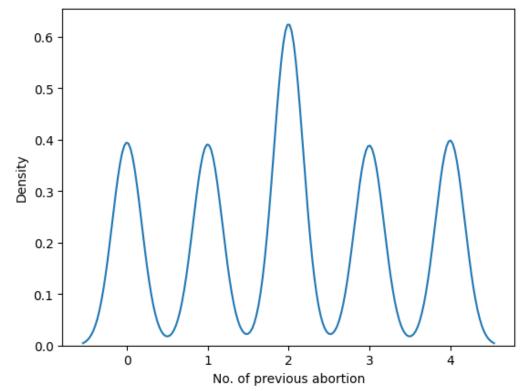
sns.boxplot(x="Autopsy shows birth defect (if applicable)",y="Disorder Subclass",data=df_train)

<Axes: xlabel='Autopsy shows birth defect (if applicable)', ylabel='Disorder Subclass'>



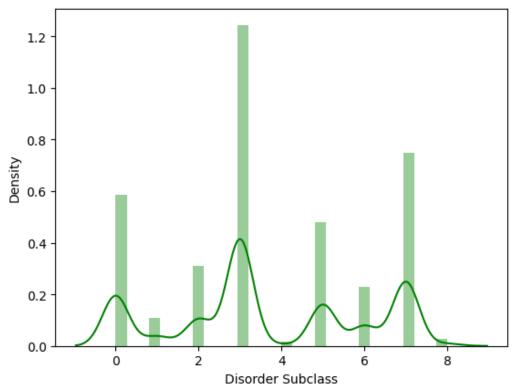
sns.kdeplot(df_train["No. of previous abortion"],palette="dark")





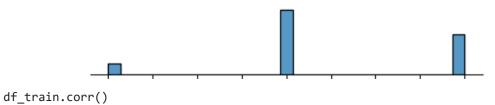
Distplot
sns.distplot(df_train['Disorder Subclass'],color="green",bins=30)

<Axes: xlabel='Disorder Subclass', ylabel='Density'>



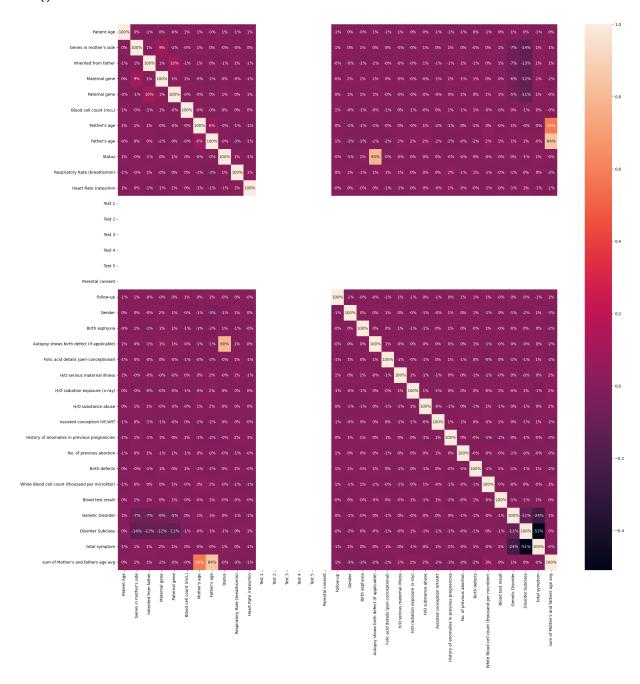
#JointPlot
plt.figure(figsize=(12,6))
sns.jointplot(x=df_train["Genetic Disorder"],y=df_train['Patient Age'],kind="hex")

<seaborn.axisgrid.JointGrid at 0x7e1592780ac0>
<Figure size 1200x600 with 0 Axes>



	Patient Age	Genes in mother's side	Inherited from father	Maternal gene	Paternal gene	Blood cell count (mcL)	Mother's age	Fi
Patient Age	1.000000	0.003452	-0.008275	0.001492	-0.004422	0.010155	0.008203	-0
Genes in mother's side	0.003452	1.000000	0.008960	0.089605	-0.007389	-0.002403	0.010247	0
Inherited from father	-0.008275	0.008960	1.000000	0.009222	0.095115	-0.007677	0.006349	0
Maternal gene	0.001492	0.089605	0.009222	1.000000	0.011829	0.008119	-0.004247	-0
Paternal gene	-0.004422	-0.007389	0.095115	0.011829	1.000000	-0.003494	-0.001070	0
Blood cell count (mcL)	0.010155	-0.002403	-0.007677	0.008119	-0.003494	1.000000	-0.001129	-0
Mother's age	0.008203	0.010247	0.006349	-0.004247	-0.001070	-0.001129	1.000000	0
Father's age	-0.000949	0.000520	0.003769	-0.022747	0.000636	-0.003498	0.059002	1
Status	0.007764	-0.000221	-0.012293	0.003566	0.013799	0.003149	-0.001519	-0
Respiratory Rate (breaths/min)	-0.011186	-0.001917	0.011765	-0.003921	0.000572	0.002200	-0.008097	-0
Heart Rate (rates/min	0.008489	0.001019	-0.010575	-0.005682	0.005119	0.000185	-0.008515	-0
Test 1	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
Test 2	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
Test 3	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
Test 4	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
Test 5	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
Parental consent	NaN	NaN	NaN	NaN	NaN	NaN	NaN	
Follow-up	-0.005463	0.009939	-0.003019	-0.003337	0.003752	0.005366	0.002463	0
Gender	0.000485	0.004320	-0.002209	0.020249	0.006780	-0.004604	-0.011823	-0
Birth asphyxia	-0.001758	0.005706	-0.009471	0.010482	0.006934	-0.007196	-0.014797	-0
Autopsy shows birth defect (if applicable)	0.007851	0.000360	-0.008126	0.009456	0.013202	-0.000833	-0.013375	-0
Folic acid details (peri-conceptional)	-0.008526	0.000115	-0.001314	0.003564	-0.000652	-0.005324	-0.002391	-0

```
plt.figure(figsize=(25,25))
sns.heatmap(df_train.iloc[:,0:39].corr(),annot=True,fmt=".0%")
plt.show()
```



```
df_train.columns
```

```
'Test 2', 'Test 3', 'Test 4', 'Test 5', 'Parental consent', 'Follow-up', 'Gender', 'Birth asphyxia',
'Autopsy shows birth defect (if applicable)',
'Folic acid details (peri-conceptional)',
'H/O serious maternal illness', 'H/O radiation exposure (x-ray)',
'H/O substance abuse', 'Assisted conception IVF/ART',
'History of anomalies in previous pregnancies',
'No. of previous abortion', 'Birth defects',
'White Blood cell count (thousand per microliter)', 'Blood test result',
'Genetic Disorder', 'Disorder Subclass', 'total symptom',
'sum of Mother's and fathers age avg'],
dtype='object')
```

df_test.head()

	Patient Id	Patient Age	Genes in mother's side	Inherited from father	Maternal gene	Paternal gene	Blood cell count (mcL)	Patient First Name	Family Name	Father na
0	PID0x4175	6	No	Yes	No	No	4.981655	Charles	NaN	Kc
1	PID0x21f5	10	Yes	No	NaN	Yes	5.118890	Catherine	NaN	Home
2	PID0x49b8	5	No	NaN	No	No	4.876204	James	NaN	Danie
3	PID0x2d97	13	No	Yes	Yes	No	4.687767	Brian	NaN	Orvi
4	PID0x58da	5	No	NaN	NaN	Yes	5.152362	Gary	NaN	Issi
5 r	ows × 43 colun	nns								

```
df_train['Genetic Disorder'].head()
```

prediction = knn.predict(x)

```
0   1.0
1   1.0
2   0.0
3   1.0
4   0.0
Name: Genetic Disorder, dtype: float32

from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors = 3)
x,y = df_train.loc[:,df_train.columns != 'Genetic Disorder'], df_train.loc[:,'Genetic Disorder']
knn.fit(x,y)
```

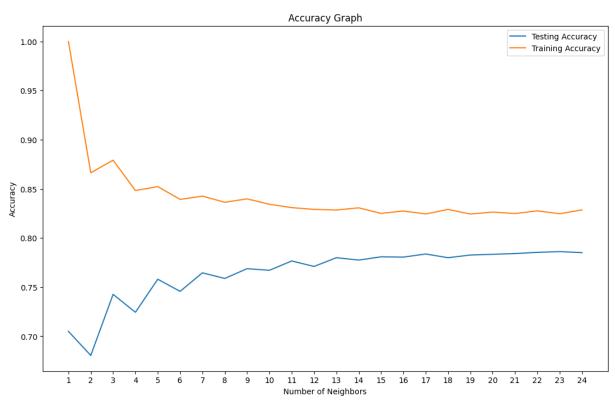
print('Prediction: {}'.format(prediction))

Prediction: [2. 1. 0. ... 1. 1. 1.]

x.columns

```
'Mother's age', 'Father's age', 'Status',
            'Respiratory Rate (breaths/min)', 'Heart Rate (rates/min', 'Test 1',
            'Test 2', 'Test 3', 'Test 4', 'Test 5', 'Parental consent', 'Follow-up',
            'Gender', 'Birth asphyxia',
            'Autopsy shows birth defect (if applicable)',
            'Folic acid details (peri-conceptional)',
            'H/O serious maternal illness', 'H/O radiation exposure (x-ray)',
            'H/O substance abuse', 'Assisted conception IVF/ART',
            'History of anomalies in previous pregnancies',
            'No. of previous abortion', 'Birth defects',
            'White Blood cell count (thousand per microliter)', 'Blood test result',
            'Disorder Subclass', 'total symptom',
            'sum of Mother's and fathers age avg'],
           dtype='object')
У
              1.0
     0
     1
              1.0
     2
              0.0
     3
              1.0
              0.0
             . . .
              1.0
     22078
     22079
              0.0
     22080
              1.0
     22081
              1.0
     22082
              0.0
     Name: Genetic Disorder, Length: 22083, dtype: float32
KNN
# train test split
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size = 0.2,random_state = 1)
knn = KNeighborsClassifier(n neighbors = 3)
x,y = df train.loc[:,df train.columns != 'Status'], df train.loc[:,'Status']
knn.fit(x train,y train)
prediction = knn.predict(x_test)
#print('Prediction: {}'.format(prediction))
print('With KNN (K=3) accuracy is: ',knn.score(x test,y test)) # accuracy
     With KNN (K=3) accuracy is: 0.7502829975096219
# train test split
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size = 0.3,random_state = 1)
knn = KNeighborsClassifier(n neighbors = 3)
x,y = df_train.loc[:,df_train.columns != 'Genetic Disorder'], df_train.loc[:,'Genetic Disorder']
knn.fit(x_train,y_train)
prediction = knn.predict(x test)
#print('Prediction: {}'.format(prediction))
print('With KNN (K=3) accuracy is: ',knn.score(x_test,y_test)) # accuracy
     With KNN (K=3) accuracy is: 0.7426415094339622
```

```
neig = np.arange(1, 25)
train_accuracy = []
test_accuracy = []
# Loop over different values of k
for i, k in enumerate(neig):
    # k from 1 to 25(exclude)
    knn = KNeighborsClassifier(n neighbors=k)
    # Fit with knn
    knn.fit(x_train,y_train)
    #train accuracy
    train_accuracy.append(knn.score(x_train, y_train))
    # test accuracy
    test accuracy.append(knn.score(x test, y test))
# Plot
plt.figure(figsize=[13,8])
plt.plot(neig, test_accuracy, label = 'Testing Accuracy')
plt.plot(neig, train_accuracy, label = 'Training Accuracy')
plt.legend()
plt.title('Accuracy Graph')
plt.xlabel('Number of Neighbors')
plt.ylabel('Accuracy')
plt.xticks(neig)
plt.savefig('graph.png')
plt.show()
print("Best accuracy is {} with K = {}".format(np.max(test_accuracy),1+test_accuracy.index(np.max(test_accuracy))
```



Best accuracy is 0.7861132075471698 with K = 23

Random Forest

```
from sklearn.model selection import train test split
from sklearn.ensemble import RandomForestClassifier
x, y = df train.loc[:, df train.columns != 'Genetic Disorder'], df train.loc[:, 'Genetic Disorder']
x train, x test, y train, y test = train test split(x, y, test size=0.8, random state=1)
# Create a Random Forest classifier
rf = RandomForestClassifier(n_estimators=120, random_state=3) # You can adjust n_estimators as neede
rf.fit(x_train, y_train)
prediction = rf.predict(x test)
accuracy = rf.score(x_test, y_test)
print('Random Forest accuracy for predicting Genetic Disorder is:', accuracy)
     Random Forest accuracy for predicting Genetic Disorder is: 0.8540782249391521
x, y = df train.loc[:, df train.columns != 'Genetic Disorder'], df train.loc[:, 'Genetic Disorder']
x train, x test, y train, y test = train test split(x, y, test size=0.8, random state=1)
# Create a Random Forest classifier
rf = RandomForestClassifier(n estimators=120, random state=1) # You can adjust n estimators as neede
rf.fit(x_train, y_train)
prediction = rf.predict(x_test)
accuracy = rf.score(x_test, y_test)
print('Random Forest accuracy for predicting Disorder Subclass is:', accuracy)
     Random Forest accuracy for predicting Disorder Subclass is: 0.8486443652006566
from sklearn.ensemble import RandomForestClassifier
import matplotlib.pyplot as plt
import numpy as np
n estimators = np.arange(1, 25)
train accuracy = []
test accuracy = []
for n in n estimators:
    # Create a Random Forest classifier with 'n' estimators
    rf = RandomForestClassifier(n_estimators=n, random_state=1)
    rf.fit(x_train, y_train)
    # Train accuracy
    train_accuracy.append(rf.score(x_train, y_train))
    # Test accuracy
    test_accuracy.append(rf.score(x_test, y_test))
```

```
# Plot
plt.figure(figsize=[13, 8])
plt.plot(n_estimators, test_accuracy, label='Testing Accuracy')
plt.plot(n_estimators, train_accuracy, label='Training Accuracy')
plt.legend()
plt.title('Number of Estimators VS Accuracy')
plt.xlabel('Number of Estimators')
plt.ylabel('Accuracy')
plt.xticks(n_estimators)
plt.savefig('rf_graph.png')
plt.show()

best_accuracy = max(test_accuracy)
best_n_estimators = n_estimators[test_accuracy.index(best_accuracy, best_n_estimators))

print("Best accuracy is {} with n_estimators = {}".format(best_accuracy, best_n_estimators))
```

Number of Estimators VS Acc

Extra Tree Classifier

```
# Import necessary libraries
from sklearn.ensemble import ExtraTreesClassifier
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy score
# Split the data into training and testing sets
# X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
x, y = df_train.loc[:, df_train.columns != 'Genetic Disorder'], df_train.loc[:, 'Genetic Disorder']
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=1)
extra trees classifier = ExtraTreesClassifier(n estimators=100, random state=42)
extra trees classifier.fit(x train,y train)
y pred = extra trees classifier.predict(x test)
accuracy = accuracy_score(y_test, y_pred)
print(f"Accuracy: {accuracy}")
     Accuracy: 0.8750282997509622
```

```
n_estimators = np.arange(1, 25)
train_accuracy = []

for n in n_estimators:
    # Create a Extra Trees Classifier with 'n' estimators
    rf = ExtraTreesClassifier(n_estimators=n, random_state=1)
    rf.fit(x_train, y_train)

# Train accuracy
    train_accuracy.append(rf.score(x_train, y_train))

# Test accuracy
    test_accuracy.append(rf.score(x_test, y_test))

# Plot
plt.figure(figsize=[13, 8])
plt.plot(n_estimators, test_accuracy, label='Testing Accuracy')
plt.plot(n_estimators, train_accuracy, label='Training Accuracy')
```