```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
df = pd.read_csv('Datasets/kag_risk_factors_cervical_cancer.csv')
pd.set option('display.max columns', 500)
df.replace('?',np.nan,inplace = True)
df.shape[0]
858
# null percentage
for i in df.columns:
    print( 'Null values in ', i , 'are', df[i].isnull().sum())
Null values in Age are 0
Null values in Number of sexual partners are 26
Null values in First sexual intercourse are 7
Null values in Num of pregnancies are 56
Null values in Smokes are 13
Null values in Smokes (years) are 13
Null values in
               Smokes (packs/year) are 13
Null values in Hormonal Contraceptives are 108
Null values in
               Hormonal Contraceptives (years) are 108
Null values in IUD are 117
Null values in IUD (years) are 117
Null values in
                STDs are 105
Null values in STDs (number) are 105
Null values in STDs:condylomatosis are 105
Null values in
                STDs:cervical condylomatosis are 105
Null values in
                STDs:vaginal condylomatosis are 105
Null values in
                STDs:vulvo-perineal condylomatosis are 105
Null values in
                STDs:syphilis are 105
Null values in
                STDs:pelvic inflammatory disease are 105
Null values in STDs:genital herpes are 105
Null values in
                STDs:molluscum contagiosum are 105
Null values in
                STDs:AIDS are 105
Null values in STDs:HIV are 105
Null values in STDs: Hepatitis B are 105
Null values in STDs: HPV are 105
Null values in STDs: Number of diagnosis are 0
Null values in STDs: Time since first diagnosis are 787
Null values in STDs: Time since last diagnosis are 787
```

```
Null values in
               Dx:Cancer are 0
Null values in
               Dx:CIN are 0
Null values in
               Dx:HPV are 0
Null values in
               Dx are 0
Null values in Hinselmann are 0
Null values in Schiller are 0
Null values in Citology are 0
Null values in Biopsy are 0
for i in df.columns:
   print( i ,'Null Percent = ',int(df[i].isnull().sum()/df.shape[0]
*100))
Age Null Percent = 0
Number of sexual partners Null Percent = 3
First sexual intercourse Null Percent = 0
Num of pregnancies Null Percent = 6
Smokes Null Percent = 1
Smokes (years) Null Percent = 1
Smokes (packs/year) Null Percent = 1
Hormonal Contraceptives Null Percent = 12
Hormonal Contraceptives (years) Null Percent = 12
IUD Null Percent = 13
IUD (years) Null Percent = 13
STDs Null Percent = 12
STDs (number) Null Percent = 12
STDs:condylomatosis Null Percent = 12
STDs:cervical condylomatosis Null Percent = 12
STDs:vaginal condylomatosis Null Percent = 12
STDs:vulvo-perineal condylomatosis Null Percent = 12
STDs:syphilis Null Percent = 12
STDs:pelvic inflammatory disease Null Percent = 12
STDs:genital herpes Null Percent = 12
STDs:molluscum contagiosum Null Percent = 12
STDs:AIDS Null Percent = 12
STDs:HIV Null Percent = 12
STDs:Hepatitis B Null Percent = 12
STDs:HPV Null Percent = 12
STDs: Number of diagnosis Null Percent = 0
STDs: Time since first diagnosis Null Percent = 91
STDs: Time since last diagnosis Null Percent = 91
Dx:Cancer Null Percent = 0
Dx:CIN Null Percent = 0
Dx:HPV Null Percent = 0
Dx Null Percent = 0
Hinselmann Null Percent = 0
Schiller Null Percent = 0
Citology Null Percent = 0
Biopsy Null Percent = 0
```

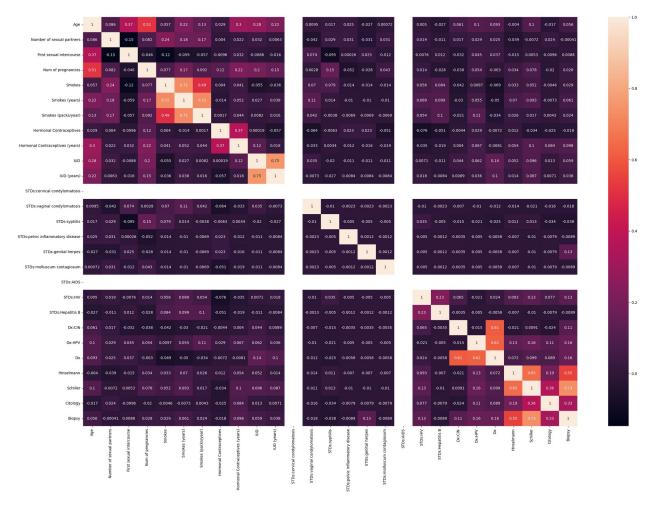
```
# columns to be dropped
# STDs: Time since first diagnosis Null Percent = 91
# STDs: Time since last diagnosis Null Percent = 91
df.drop(columns= ['STDs: Time since first diagnosis','STDs: Time since
last diagnosis'],inplace = True)
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 858 entries, 0 to 857
Data columns (total 34 columns):
     Column
                                          Non-Null Count
                                                          Dtype
- - -
     _ _ _ _ _
                                                          ----
0
     Age
                                          858 non-null
                                                          int64
 1
     Number of sexual partners
                                          832 non-null
                                                          object
 2
     First sexual intercourse
                                          851 non-null
                                                          object
 3
     Num of pregnancies
                                          802 non-null
                                                          object
4
     Smokes
                                          845 non-null
                                                          object
 5
     Smokes (years)
                                          845 non-null
                                                          object
 6
     Smokes (packs/year)
                                          845 non-null
                                                          object
 7
     Hormonal Contraceptives
                                          750 non-null
                                                          object
 8
     Hormonal Contraceptives (years)
                                          750 non-null
                                                          object
9
                                          741 non-null
                                                          object
10 IUD (years)
                                          741 non-null
                                                          object
 11
    STDs
                                          753 non-null
                                                          object
 12 STDs (number)
                                          753 non-null
                                                          object
 13 STDs:condylomatosis
                                          753 non-null
                                                          object
                                          753 non-null
 14 STDs:cervical condylomatosis
                                                          object
 15
    STDs:vaginal condylomatosis
                                          753 non-null
                                                          object
                                          753 non-null
 16 STDs:vulvo-perineal condylomatosis
                                                          object
 17
     STDs:syphilis
                                          753 non-null
                                                          object
 18 STDs:pelvic inflammatory disease
                                          753 non-null
                                                          object
 19 STDs:genital herpes
                                          753 non-null
                                                          object
 20 STDs:molluscum contagiosum
                                          753 non-null
                                                          object
 21 STDs:AIDS
                                          753 non-null
                                                          object
 22 STDs:HIV
                                          753 non-null
                                                          object
 23 STDs:Hepatitis B
                                          753 non-null
                                                          object
 24 STDs:HPV
                                          753 non-null
                                                          object
 25 STDs: Number of diagnosis
                                          858 non-null
                                                          int64
 26 Dx:Cancer
                                          858 non-null
                                                          int64
27
     Dx:CIN
                                          858 non-null
                                                          int64
 28
    Dx:HPV
                                          858 non-null
                                                          int64
 29
     Dx
                                          858 non-null
                                                          int64
                                          858 non-null
 30 Hinselmann
                                                          int64
 31
    Schiller
                                          858 non-null
                                                          int64
 32
    Citology
                                          858 non-null
                                                          int64
33
     Biopsy
                                          858 non-null
                                                          int64
dtypes: int64(10), object(24)
memory usage: 228.0+ KB
```

```
df['Age'].mean()
26.82051282051282

for i in dfl.columns:
    dfl[i] = dfl[i].astype(float)

plt.figure(figsize=(30,20))
sns.heatmap(dfl.corr(),annot= True)

<Axes: >
```



```
df1.drop(columns= ['STDs:condylomatosis','STDs:vulvo-perineal
condylomatosis','STDs:condylomatosis'],inplace = True)

df1.shape
(858, 27)

df1= df.copy()
```

```
X = df1.drop('Biopsy',axis = 1)
y = df1['Biopsy']
from sklearn.model_selection import train test split
X train, X test, y train, y test = train test split(X, y,
test size=0.33, random state=42)
from sklearn.metrics import accuracy score
from sklearn.linear_model import LogisticRegression
log reg = LogisticRegression(max iter= 1000)
log reg.fit(X train scaled,y train)
log reg pred = log reg.predict(X test scaled)
print(accuracy score(y test,log reg pred))
0.9577464788732394
from sklearn.tree import DecisionTreeClassifier
clf = DecisionTreeClassifier()
clf.fit(X train,y train)
clf pred = log reg.predict(X test)
print(accuracy score(y test,clf pred))
0.9577464788732394
# Scaling
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
X train scaled = sc.fit transform(X train)
X test scaled = sc.transform(X test)
from sklearn.svm import SVC
svc = SVC()
svc.fit(X train scaled,y train)
svc pred = svc.predict(X test scaled)
print(accuracy score(y test,svc pred))
final pred = []
def voting(models, X_test):
    for i in X test.index:
        ones=0
        zeroes = 0
        input1 = X test[X test.index == i].values
        for mod in models:
            out = mod.predict(input1)
            if out == 0:
                zeroes = zeroes+1
            else:
                ones = ones +1
        if(ones > zeroes):
```

```
final pred.append(1)
        else:
            final pred.append(0)
    return final_pred
voting out = voting([svc,log reg,knn,naive],X test)
len(voting out)
284
X train.shape, y train.shape, X test.shape, y test.shape, dfl.shape
((574, 26), (574,), (284, 26), (284,), (858, 27))
print(accuracy_score(y_test,voting_out))
0.9366197183098591
X test[X test.index == 713].values.shape
(1, 26)
X test
           Number of sexual partners First sexual intercourse \
     Age
713
     16.0
                                  1.0
                                                            16.0
604
    23.0
                                  3.0
                                                            17.0
120 33.0
                                                            16.0
                                  1.0
208 27.0
                                  4.0
                                                            16.0
380 18.0
                                  3.0
                                                            15.0
. .
     . . .
                                                             . . .
422
    18.0
                                  2.0
                                                            15.0
                                  1.0
764
     23.0
                                                            15.0
477
     38.0
                                  2.0
                                                            19.0
41
     37.0
                                  2.0
                                                            18.0
530 21.0
                                  4.0
                                                            15.0
     Num of pregnancies Smokes Smokes (years)
                                                  Smokes
(packs/year) \
713
                    1.0
                             0.0
                                              0.0
                                                                 0.000
604
                    2.0
                             0.0
                                             0.0
                                                                 0.000
120
                    4.0
                             0.0
                                             0.0
                                                                 0.000
208
                    1.0
                             0.0
                                             0.0
                                                                 0.000
380
                    1.0
                             1.0
                                              2.0
                                                                 0.003
```

422		2.0	1.0	0.5		0.050
764		3.0	0.0	0.0		0.000
477		2.0	0.0	0.0		0.000
41		1.0	0.0	0.0		0.000
530		1.0	0.0	0.0		0.000
713 604 120 208 380 422 764 477 41 530	Hormonal Cont	raceptives 0.0 0.0 1.0 1.0 1.0 0.0 0.0	9 9 9 9 9 9 9	ntraceptives	(years) 0.00 0.00 0.67 0.58 0.33 0.25 0.00 0.00 0.00	IUD \ 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.
713 0.0 604 0.0	<pre>IUD (years) ! ylomatosis \ 0.0 0.0</pre>	SIDS: CEIV.	ical condyloma [.]	tosis STDs:\ 0.0	/aginat	
120 0.0 208 0.0 380 0.0 422 0.0 764 0.0 477 0.0 41 0.0 530	0.0 0.0 0.0 0.0 7.0 0.0 0.0			0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0		

	STDs:syphi	lis	STDs:pelvic	inflammatory	/ disease	STDs:genital
herpe	es \		3.23.pct12c			5.55.gc
713	(0.0			0.0	
0.0 604	(0.0			0.0	
0.0	,	0.0			0.0	
120	(0.0			0.0	
0.0	,	0.0			0.0	
0.0	,	0.0			0.0	
380	(0.0			0.0	
0.0						
422	(0.0			0.0	
0.0 764	(0.0			0.0	
0.0	,	0.0			0.0	
477	(0.0			0.0	
0.0 41	(0.0			0.0	
0.0	,	0.0			0.0	
530	(0.0			0.0	
0.0						
	STDs:mollus	scum	contagiosum	STDs:AIDS	STDs:HIV	STDs:Hepatitis B
\						
713			0.0	0.0	0.0	0.0
604			0.0	0.0	0.0	0.0
120			0.0	0.0	0.0	0.0
120			0.0	0.0	0.0	0.0
208			0.0	0.0	0.0	0.0
380			0.0	0.0	0.0	0.0
500			0.0	0.0	0.0	0.0
422			0.0	0.0	0.0	0.0
764			0.0	0.0	0.0	0.0
477			0.0	0.0	0.0	0.0
477 41			0.0 0.0	0.0	0.0 1.0	0.0 0.0

0.0						
3		1.0			3.0	0.0
0.0						
4		1.0			15.0	0.0
0.0						
• •						
853		0.0			0.0	0.0
0.0		0.0			010	0.0
854		1.0			8.0	0.0
0.0						
855		1.0			0.08	0.0
0.0		1 0			0 00	0 0
856 0.0		1.0			0.08	0.0
857		1.0			0.5	0.0
0.0		210				
					_	
	STDs STDs (n	umber) STDs	:condylomatos:	s STDs:cervica	al.	
	dylomatosis	0.0	0	٥		
0 0.0	0.0	0.0	9	U		
1	0.0	0.0	0	0		
0.0				_		
2	0.0	0.0	0	0		
0.0						
3	0.0	0.0	Θ	0		
0.0	0.0	0.0	0	A		
0.0	0.0	0.0	0	U		
853	0.0	0.0	0	0		
0.0				•		
854	0.0	0.0	Θ	Θ		
0.0 855	0.0	0.0	0	Θ		
0.0	0.0	0.0	U ,	U		
856	0.0	0.0	0	0		
0.0						
857	0.0	0.0	0	0		
0.0						
	STDs:vaginal	condylomat		o-perineal con	ndylom	
0 1 2 3			0.0 0.0			0.0 0.0
2			0.0			0.0
3			0.0			0.0
4			0.0			0.0

853 854 855		0.6 0.6 0.6	9 9		0.0 0.0 0.0	
856 857		0.0 0.0			0.0 0.0	
\	STDs:syphilis	STDs:pelvic i	inflammator	y disease	STDs:genital he	rpes
ò	0.0			0.0		0.0
1	0.0			0.0		0.0
2	0.0			0.0		0.0
3	0.0			0.0		0.0
4	0.0			0.0		0.0
853	0.0			0.0		0.0
854	0.0			0.0		0.0
855	0.0			0.0		0.0
856	0.0			0.0		0.0
857	0.0			0.0		0.0
	CTD 11		CTD - ATDC	CTD - UTV (TD. H	
	s:HPV \				STDs:Hepatitis B	
0 0.0		0.0	0.0	0.0	0.0	
1 0.0		0.0	0.0	0.0	0.0	
2 0.0		0.0	0.0	0.0	0.0	
3		0.0	0.0	0.0	0.0	
4		0.0	0.0	0.0	0.0	
0.0						
853		0.0	0.0	0.0	0.0	
0.0 854		0.0	0.0	0.0	0.0	
0.0						

855			0.0	0.0	0.0			0.0
0.0			0 0	0 0	0.0			0 0
856 0.0			0.0	0.0	0.0			0.0
857			0.0	0.0	0.0			0.0
0.0								
CTD -	a . Niconala a .a	-£ 43			D CTN	D LIDV	D	
Sids Hinselmar	s: Number	от dlagn	OSIS DX	(:Cancer	Dx:CIN	Dx:HPV	Dx	
0	111 \		0	0	0	0	0	
			· ·	J		J	J	
1			0	0	0	0	0	
0								
2			0	0	0	0	0	
0			Θ	1	Θ	1	0	
0			U	1	U	1	0	
0 1 0 2 0 3 0 4			0	0	0	0	0	
0			-	-	_	_		
853			0	0	0	0	0	
0 854			0	0	Θ	0	0	
0			U	U	U	U	U	
855			0	0	0	0	0	
0								
856			0	0	0	0	0	
0			0	0	0	0	0	
857 0			0	0	0	0	0	
O								
Schi	iller Cit	cology B	iopsy					
0	0	0	0					
1	0	0	0					
2 3 4	0 0	0 0	0 0					
4	Θ	0	0					
853	Θ	0	0					
854	0	0	0					
855	0	1	0					
856 857	0 0	0 0	0 0					
037	U	U	U					
	x 34 col							
df[df.ind	dex == 9].	values						

```
array([[44, '3.0', '15.0', nan, '1.0', '1.266972909', '2.8', '0.0',
       '0.0', nan, nan, '0.0', '0.0', '0.0', '0.0', '0.0', '0.0',
'0.0',
        '0.0', '0.0', '0.0', '0.0', '0.0', '0.0', '0.0', '0.0', 0, 0, 0,
0,
        0, 0, 0, 0]], dtype=object)
from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier()
knn.fit(X_train_scaled,y_train)
knn pred = knn.predict(X test scaled)
print(accuracy_score(y_test,knn_pred))
# with scaled accurecy 0.943661\overline{9}718309859
0.9401408450704225
from sklearn.naive bayes import GaussianNB
naive = GaussianNB()
naive.fit(X train,y train)
naive pred = naive.predict(X test scaled)
print(accuracy score(y test, naive pred))
0.9366197183098591
C:\ProgramData\anaconda3\lib\site-packages\sklearn\base.py:420:
UserWarning: X does not have valid feature names, but GaussianNB was
fitted with feature names
  warnings.warn(
# voting out = voting([svc,log reg,knn,naive],X test)
from sklearn.ensemble import RandomForestClassifier
rand clf = RandomForestClassifier()
from sklearn.ensemble import RandomForestClassifier, VotingClassifier
eclf1 = VotingClassifier(estimators=[('svc', svc), ('log reg',
log reg), ('knn', knn),('naive', naive),('random', rand clf)],
voting='hard')
eclf1 = eclf1.fit(X train scaled, y train)
from sklearn.linear model import LogisticRegression
log reg = LogisticRegression()
voting pred = eclf1.predict(X test scaled)
print(accuracy score(y_test,voting_pred))
0.9577464788732394
```