

rinex-majorproject1

August 26, 2024

1 MAJOR PROJECT 1

```
[ ]: Diagnosis - KNN Classifier
```

```
[1]: import pandas as pd
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
```

```
[9]: df=pd.read_csv('C:
↪\\Users\\moham\\OneDrive\\Desktop\\Documents\\KNNAlgorithmDataset.csv')
```

```
[10]: df.head()
```

```
[10]:
```

	id	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	\
0	842302	M	17.99	10.38	122.80	1001.0	
1	842517	M	20.57	17.77	132.90	1326.0	
2	84300903	M	19.69	21.25	130.00	1203.0	
3	84348301	M	11.42	20.38	77.58	386.1	
4	84358402	M	20.29	14.34	135.10	1297.0	

	smoothness_mean	compactness_mean	concavity_mean	concave	points_mean	\
0	0.11840	0.27760	0.3001		0.14710	
1	0.08474	0.07864	0.0869		0.07017	
2	0.10960	0.15990	0.1974		0.12790	
3	0.14250	0.28390	0.2414		0.10520	
4	0.10030	0.13280	0.1980		0.10430	

...	radius_worst	texture_worst	perimeter_worst	area_worst	\
0	25.38	17.33	184.60	2019.0	
1	24.99	23.41	158.80	1956.0	
2	23.57	25.53	152.50	1709.0	
3	14.91	26.50	98.87	567.7	
4	22.54	16.67	152.20	1575.0	

	smoothness_worst	compactness_worst	concavity_worst	concave	points_worst	\
0	0.1622	0.6656	0.7119		0.2654	

1	0.1238	0.1866	0.2416	0.1860
2	0.1444	0.4245	0.4504	0.2430
3	0.2098	0.8663	0.6869	0.2575
4	0.1374	0.2050	0.4000	0.1625

	symmetry_worst	fractal_dimension_worst
0	0.4601	0.11890
1	0.2750	0.08902
2	0.3613	0.08758
3	0.6638	0.17300
4	0.2364	0.07678

[5 rows x 32 columns]

```
[11]: print(df.columns)
```

```
Index(['id', 'diagnosis', 'radius_mean', 'texture_mean', 'perimeter_mean',
      'area_mean', 'smoothness_mean', 'compactness_mean', 'concavity_mean',
      'concave points_mean', 'symmetry_mean', 'fractal_dimension_mean',
      'radius_se', 'texture_se', 'perimeter_se', 'area_se', 'smoothness_se',
      'compactness_se', 'concavity_se', 'concave points_se', 'symmetry_se',
      'fractal_dimension_se', 'radius_worst', 'texture_worst',
      'perimeter_worst', 'area_worst', 'smoothness_worst',
      'compactness_worst', 'concavity_worst', 'concave points_worst',
      'symmetry_worst', 'fractal_dimension_worst'],
      dtype='object')
```

```
[13]: # Strip any whitespace from column names
df.columns = df.columns.str.strip()

# Safely drop the 'Unnamed: 32' column if it exists
df = df.drop('Unnamed: 32', axis=1, errors='ignore')

# Safely drop the 'id' column if it exists
df = df.drop('id', axis=1, errors='ignore')

# Display the first five rows of the DataFrame
df.head()
```

```
[13]:  diagnosis  radius_mean  texture_mean  perimeter_mean  area_mean  \
0         M         17.99         10.38         122.80        1001.0
1         M         20.57         17.77         132.90        1326.0
2         M         19.69         21.25         130.00        1203.0
3         M         11.42         20.38          77.58         386.1
4         M         20.29         14.34         135.10        1297.0

      smoothness_mean  compactness_mean  concavity_mean  concave points_mean  \
```

0	0.11840	0.27760	0.3001	0.14710
1	0.08474	0.07864	0.0869	0.07017
2	0.10960	0.15990	0.1974	0.12790
3	0.14250	0.28390	0.2414	0.10520
4	0.10030	0.13280	0.1980	0.10430

	symmetry_mean	...	radius_worst	texture_worst	perimeter_worst	\
0	0.2419	...	25.38	17.33	184.60	
1	0.1812	...	24.99	23.41	158.80	
2	0.2069	...	23.57	25.53	152.50	
3	0.2597	...	14.91	26.50	98.87	
4	0.1809	...	22.54	16.67	152.20	

	area_worst	smoothness_worst	compactness_worst	concavity_worst	\
0	2019.0	0.1622	0.6656	0.7119	
1	1956.0	0.1238	0.1866	0.2416	
2	1709.0	0.1444	0.4245	0.4504	
3	567.7	0.2098	0.8663	0.6869	
4	1575.0	0.1374	0.2050	0.4000	

	concave points_worst	symmetry_worst	fractal_dimension_worst
0	0.2654	0.4601	0.11890
1	0.1860	0.2750	0.08902
2	0.2430	0.3613	0.08758
3	0.2575	0.6638	0.17300
4	0.1625	0.2364	0.07678

[5 rows x 31 columns]

```
[14]: df.columns
```

```
[14]: Index(['diagnosis', 'radius_mean', 'texture_mean', 'perimeter_mean',
          'area_mean', 'smoothness_mean', 'compactness_mean', 'concavity_mean',
          'concave points_mean', 'symmetry_mean', 'fractal_dimension_mean',
          'radius_se', 'texture_se', 'perimeter_se', 'area_se', 'smoothness_se',
          'compactness_se', 'concavity_se', 'concave points_se', 'symmetry_se',
          'fractal_dimension_se', 'radius_worst', 'texture_worst',
          'perimeter_worst', 'area_worst', 'smoothness_worst',
          'compactness_worst', 'concavity_worst', 'concave points_worst',
          'symmetry_worst', 'fractal_dimension_worst'],
          dtype='object')
```

```
[15]: unique_classes = df['diagnosis'].unique()
       print(f"Unique classes in 'diagnosis': {unique_classes}")
```

Unique classes in 'diagnosis': ['M' 'B']

```
[16]: df['diagnosis'] = df['diagnosis'].replace({'M': 1, 'B': 0})
print(df.head())
```

	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean	\
0	1	17.99	10.38	122.80	1001.0	
1	1	20.57	17.77	132.90	1326.0	
2	1	19.69	21.25	130.00	1203.0	
3	1	11.42	20.38	77.58	386.1	
4	1	20.29	14.34	135.10	1297.0	

	smoothness_mean	compactness_mean	concavity_mean	concave points_mean	\
0	0.11840	0.27760	0.3001	0.14710	
1	0.08474	0.07864	0.0869	0.07017	
2	0.10960	0.15990	0.1974	0.12790	
3	0.14250	0.28390	0.2414	0.10520	
4	0.10030	0.13280	0.1980	0.10430	

	symmetry_mean	...	radius_worst	texture_worst	perimeter_worst	\
0	0.2419	...	25.38	17.33	184.60	
1	0.1812	...	24.99	23.41	158.80	
2	0.2069	...	23.57	25.53	152.50	
3	0.2597	...	14.91	26.50	98.87	
4	0.1809	...	22.54	16.67	152.20	

	area_worst	smoothness_worst	compactness_worst	concavity_worst	\
0	2019.0	0.1622	0.6656	0.7119	
1	1956.0	0.1238	0.1866	0.2416	
2	1709.0	0.1444	0.4245	0.4504	
3	567.7	0.2098	0.8663	0.6869	
4	1575.0	0.1374	0.2050	0.4000	

	concave points_worst	symmetry_worst	fractal_dimension_worst
0	0.2654	0.4601	0.11890
1	0.1860	0.2750	0.08902
2	0.2430	0.3613	0.08758
3	0.2575	0.6638	0.17300
4	0.1625	0.2364	0.07678

[5 rows x 31 columns]

```
[17]: # Preparing feature set
X = df[['radius_mean', 'texture_mean', 'perimeter_mean',
        'area_mean', 'smoothness_mean', 'compactness_mean', 'concavity_mean',
        'concave points_mean', 'symmetry_mean', 'fractal_dimension_mean',
        'radius_se', 'texture_se', 'perimeter_se', 'area_se', 'smoothness_se',
        'compactness_se', 'concavity_se', 'concave points_se', 'symmetry_se',
        'fractal_dimension_se', 'radius_worst', 'texture_worst',
```

```

        'perimeter_worst', 'area_worst', 'smoothness_worst',
        'compactness_worst', 'concavity_worst', 'concave points_worst',
        'symmetry_worst', 'fractal_dimension_worst']]
Y = df[['diagnosis']]
print(X[0:5])
print(Y[0:5])

```

	radius_mean	texture_mean	perimeter_mean	area_mean	smoothness_mean	\
0	17.99	10.38	122.80	1001.0	0.11840	
1	20.57	17.77	132.90	1326.0	0.08474	
2	19.69	21.25	130.00	1203.0	0.10960	
3	11.42	20.38	77.58	386.1	0.14250	
4	20.29	14.34	135.10	1297.0	0.10030	

	compactness_mean	concavity_mean	concave points_mean	symmetry_mean	\
0	0.27760	0.3001	0.14710	0.2419	
1	0.07864	0.0869	0.07017	0.1812	
2	0.15990	0.1974	0.12790	0.2069	
3	0.28390	0.2414	0.10520	0.2597	
4	0.13280	0.1980	0.10430	0.1809	

	fractal_dimension_mean	...	radius_worst	texture_worst	perimeter_worst	\
0	0.07871	...	25.38	17.33	184.60	
1	0.05667	...	24.99	23.41	158.80	
2	0.05999	...	23.57	25.53	152.50	
3	0.09744	...	14.91	26.50	98.87	
4	0.05883	...	22.54	16.67	152.20	

	area_worst	smoothness_worst	compactness_worst	concavity_worst	\
0	2019.0	0.1622	0.6656	0.7119	
1	1956.0	0.1238	0.1866	0.2416	
2	1709.0	0.1444	0.4245	0.4504	
3	567.7	0.2098	0.8663	0.6869	
4	1575.0	0.1374	0.2050	0.4000	

	concave points_worst	symmetry_worst	fractal_dimension_worst
0	0.2654	0.4601	0.11890
1	0.1860	0.2750	0.08902
2	0.2430	0.3613	0.08758
3	0.2575	0.6638	0.17300
4	0.1625	0.2364	0.07678

[5 rows x 30 columns]

	diagnosis
0	1
1	1
2	1

```
3      1
4      1
```

```
[18]: print(df.describe())
```

	diagnosis	radius_mean	texture_mean	perimeter_mean	area_mean \
count	569.000000	569.000000	569.000000	569.000000	569.000000
mean	0.372583	14.127292	19.289649	91.969033	654.889104
std	0.483918	3.524049	4.301036	24.298981	351.914129
min	0.000000	6.981000	9.710000	43.790000	143.500000
25%	0.000000	11.700000	16.170000	75.170000	420.300000
50%	0.000000	13.370000	18.840000	86.240000	551.100000
75%	1.000000	15.780000	21.800000	104.100000	782.700000
max	1.000000	28.110000	39.280000	188.500000	2501.000000

	smoothness_mean	compactness_mean	concavity_mean	concave points_mean \
count	569.000000	569.000000	569.000000	569.000000
mean	0.096360	0.104341	0.088799	0.048919
std	0.014064	0.052813	0.079720	0.038803
min	0.052630	0.019380	0.000000	0.000000
25%	0.086370	0.064920	0.029560	0.020310
50%	0.095870	0.092630	0.061540	0.033500
75%	0.105300	0.130400	0.130700	0.074000
max	0.163400	0.345400	0.426800	0.201200

	symmetry_mean ...	radius_worst	texture_worst	perimeter_worst \
count	569.000000 ...	569.000000	569.000000	569.000000
mean	0.181162 ...	16.269190	25.677223	107.261213
std	0.027414 ...	4.833242	6.146258	33.602542
min	0.106000 ...	7.930000	12.020000	50.410000
25%	0.161900 ...	13.010000	21.080000	84.110000
50%	0.179200 ...	14.970000	25.410000	97.660000
75%	0.195700 ...	18.790000	29.720000	125.400000
max	0.304000 ...	36.040000	49.540000	251.200000

	area_worst	smoothness_worst	compactness_worst	concavity_worst \
count	569.000000	569.000000	569.000000	569.000000
mean	880.583128	0.132369	0.254265	0.272188
std	569.356993	0.022832	0.157336	0.208624
min	185.200000	0.071170	0.027290	0.000000
25%	515.300000	0.116600	0.147200	0.114500
50%	686.500000	0.131300	0.211900	0.226700
75%	1084.000000	0.146000	0.339100	0.382900
max	4254.000000	0.222600	1.058000	1.252000

	concave points_worst	symmetry_worst	fractal_dimension_worst
count	569.000000	569.000000	569.000000
mean	0.114606	0.290076	0.083946

std	0.065732	0.061867	0.018061
min	0.000000	0.156500	0.055040
25%	0.064930	0.250400	0.071460
50%	0.099930	0.282200	0.080040
75%	0.161400	0.317900	0.092080
max	0.291000	0.663800	0.207500

[8 rows x 31 columns]

```
[19]: import matplotlib.pyplot as plt
      from sklearn import preprocessing
      # Preprocessing
      X = preprocessing.StandardScaler().fit(X).transform(X.astype(float))
      X[0:5]
```

```
[19]: array([[ 1.09706398e+00, -2.07333501e+00,  1.26993369e+00,
                9.84374905e-01,  1.56846633e+00,  3.28351467e+00,
                2.65287398e+00,  2.53247522e+00,  2.21751501e+00,
                2.25574689e+00,  2.48973393e+00, -5.65265059e-01,
                2.83303087e+00,  2.48757756e+00, -2.14001647e-01,
                1.31686157e+00,  7.24026158e-01,  6.60819941e-01,
                1.14875667e+00,  9.07083081e-01,  1.88668963e+00,
               -1.35929347e+00,  2.30360062e+00,  2.00123749e+00,
                1.30768627e+00,  2.61666502e+00,  2.10952635e+00,
                2.29607613e+00,  2.75062224e+00,  1.93701461e+00],
              [ 1.82982061e+00, -3.53632408e-01,  1.68595471e+00,
                1.90870825e+00, -8.26962447e-01, -4.87071673e-01,
               -2.38458552e-02,  5.48144156e-01,  1.39236330e-03,
               -8.68652457e-01,  4.99254601e-01, -8.76243603e-01,
                2.63326966e-01,  7.42401948e-01, -6.05350847e-01,
               -6.92926270e-01, -4.40780058e-01,  2.60162067e-01,
               -8.05450380e-01, -9.94437403e-02,  1.80592744e+00,
               -3.69203222e-01,  1.53512599e+00,  1.89048899e+00,
               -3.75611957e-01, -4.30444219e-01, -1.46748968e-01,
                1.08708430e+00, -2.43889668e-01,  2.81189987e-01],
              [ 1.57988811e+00,  4.56186952e-01,  1.56650313e+00,
                1.55888363e+00,  9.42210440e-01,  1.05292554e+00,
                1.36347845e+00,  2.03723076e+00,  9.39684817e-01,
               -3.98007910e-01,  1.22867595e+00, -7.80083377e-01,
                8.50928301e-01,  1.18133606e+00, -2.97005012e-01,
                8.14973504e-01,  2.13076435e-01,  1.42482747e+00,
                2.37035535e-01,  2.93559404e-01,  1.51187025e+00,
               -2.39743838e-02,  1.34747521e+00,  1.45628455e+00,
                5.27407405e-01,  1.08293217e+00,  8.54973944e-01,
                1.95500035e+00,  1.15225500e+00,  2.01391209e-01],
              [-7.68909287e-01,  2.53732112e-01, -5.92687167e-01,
               -7.64463792e-01,  3.28355348e+00,  3.40290899e+00,
```

```

1.91589718e+00, 1.45170736e+00, 2.86738293e+00,
4.91091929e+00, 3.26373441e-01, -1.10409044e-01,
2.86593405e-01, -2.88378148e-01, 6.89701660e-01,
2.74428041e+00, 8.19518384e-01, 1.11500701e+00,
4.73268037e+00, 2.04751088e+00, -2.81464464e-01,
1.33984094e-01, -2.49939304e-01, -5.50021228e-01,
3.39427470e+00, 3.89339743e+00, 1.98958826e+00,
2.17578601e+00, 6.04604135e+00, 4.93501034e+00],
[ 1.75029663e+00, -1.15181643e+00, 1.77657315e+00,
1.82622928e+00, 2.80371830e-01, 5.39340452e-01,
1.37101143e+00, 1.42849277e+00, -9.56046689e-03,
-5.62449981e-01, 1.27054278e+00, -7.90243702e-01,
1.27318941e+00, 1.19035676e+00, 1.48306716e+00,
-4.85198799e-02, 8.28470780e-01, 1.14420474e+00,
-3.61092272e-01, 4.99328134e-01, 1.29857524e+00,
-1.46677038e+00, 1.33853946e+00, 1.22072425e+00,
2.20556166e-01, -3.13394511e-01, 6.13178758e-01,
7.29259257e-01, -8.68352984e-01, -3.97099619e-01]])

```

```

[20]: import seaborn as sns
# Viz.
x_cols = ['radius_mean', 'texture_mean', 'perimeter_mean', 'area_mean',
'smoothness_mean', 'compactness_mean', 'concavity_mean',
'concave points_mean', 'symmetry_mean', 'fractal_dimension_mean',
'radius_se', 'texture_se', 'perimeter_se', 'area_se', 'smoothness_se',
'compactness_se', 'concavity_se', 'concave points_se', 'symmetry_se',
'fractal_dimension_se', 'radius_worst', 'texture_worst',
'perimeter_worst', 'area_worst', 'smoothness_worst',
'compactness_worst', 'concavity_worst', 'concave points_worst',
'symmetry_worst', 'fractal_dimension_worst']

melted_df = df.melt(id_vars="diagnosis", value_vars=x_cols)

g = sns.FacetGrid(melted_df, col="variable", col_wrap=3, sharex=False,
↪sharey=False, height=4)

def scatter_with_hue(data, **kws):
    sns.scatterplot(data=data, x="value", y="diagnosis", hue="diagnosis",
↪palette={0: "blue", 1: "red"}, legend=False, **kws)

g.map_dataframe(scatter_with_hue)

for ax in g.axes.flat:
    xlabel = ax.get_title().split('=')[1]
    ax.set_xlabel(xlabel)
    ax.set_title('')

```

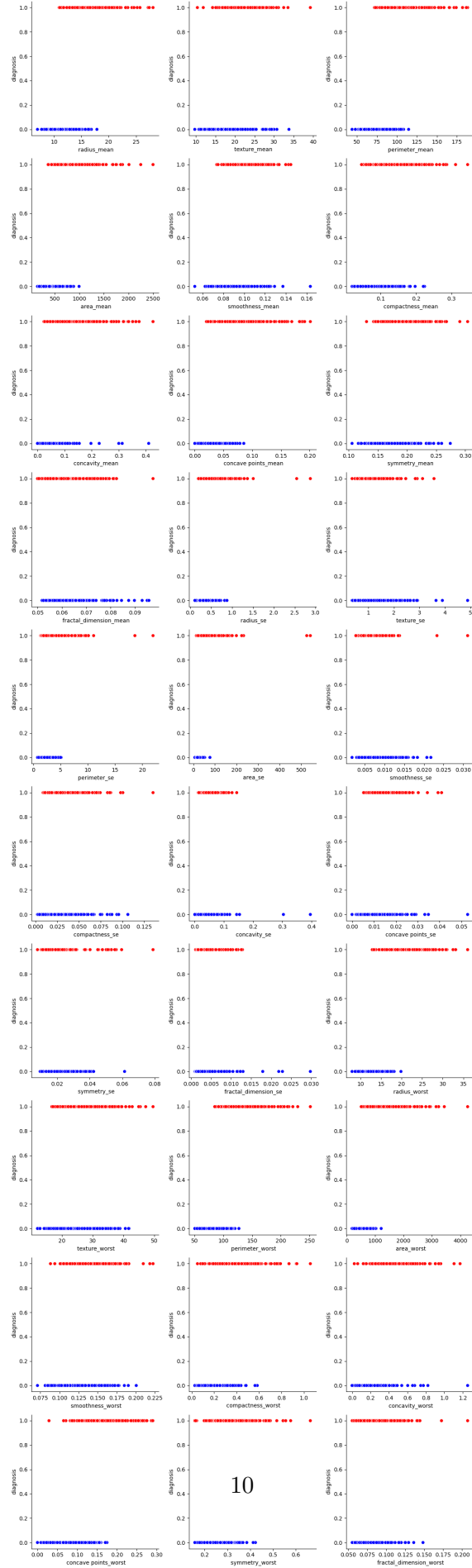


```
plt.subplots_adjust(hspace=0.8)
```

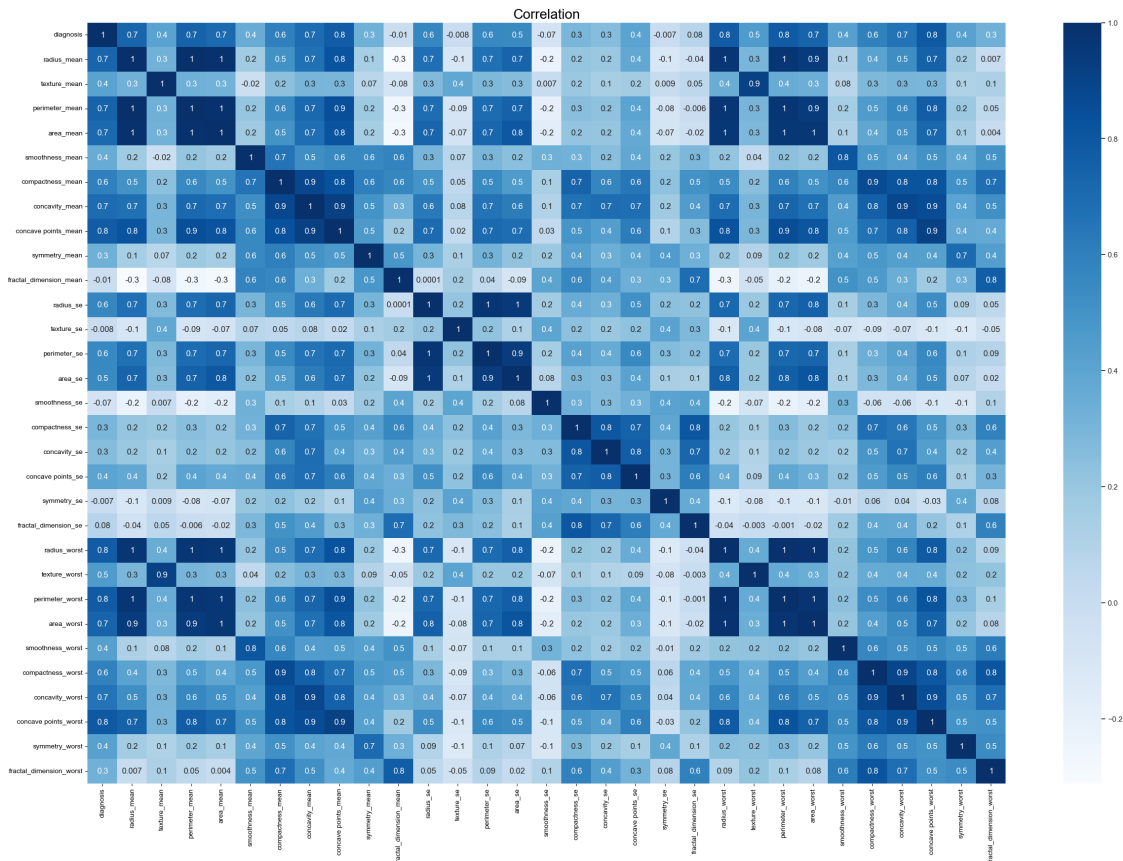
```
g.tight_layout()
```

```
plt.show()
```

C:\Users\moham\anaconda3\Lib\site-packages\seaborn\axisgrid.py:118: UserWarning:
The figure layout has changed to tight
self._figure.tight_layout(*args, **kwargs)



```
[21]: plt.figure(figsize=(30,20))
plt.title("Correlation", fontsize=20)
sns.set_theme(style="white")
corr = df.corr()
heatmap = sns.heatmap(corr, annot=True, cmap="Blues", fmt='.1g')
```



```
[22]: 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=4)
print ('Train set:', X_train.shape, y_train.shape)
print ('Test set:', X_test.shape, y_test.shape)
```

Train set: (455, 30) (455, 1)
Test set: (114, 30) (114, 1)

```
[25]: import numpy as np
from sklearn.neighbors import KNeighborsClassifier
from sklearn import metrics
```

```

Ks = 100
mean_acc = np.zeros((Ks-1))

for n in range(1, Ks):
    # Train the model and predict
    neigh = KNeighborsClassifier(n_neighbors=n).fit(X_train, y_train.values.
    ↪ravel())
    yhat = neigh.predict(X_test)
    mean_acc[n-1] = metrics.accuracy_score(y_test, yhat)

# Display the mean accuracy scores for different values of K
mean_acc

```

```

[25]: array([0.94736842, 0.97368421, 0.98245614, 0.98245614, 0.96491228,
            0.97368421, 0.95614035, 0.96491228, 0.97368421, 0.98245614,
            0.98245614, 0.98245614, 0.98245614, 0.99122807,
            0.99122807, 0.99122807, 0.98245614, 0.98245614, 0.98245614,
            0.98245614, 0.98245614, 0.98245614, 0.98245614, 0.98245614,
            0.98245614, 0.98245614, 0.98245614, 0.98245614, 0.98245614,
            0.97368421, 0.97368421, 0.97368421, 0.97368421, 0.96491228,
            0.96491228, 0.96491228, 0.96491228, 0.97368421, 0.97368421,
            0.97368421, 0.97368421, 0.98245614, 0.97368421,
            0.97368421, 0.97368421, 0.96491228, 0.97368421, 0.97368421,
            0.97368421, 0.97368421, 0.96491228, 0.96491228, 0.96491228,
            0.96491228, 0.96491228, 0.96491228, 0.96491228, 0.96491228,
            0.96491228, 0.96491228, 0.96491228, 0.96491228, 0.96491228,
            0.96491228, 0.97368421, 0.96491228, 0.97368421, 0.96491228,
            0.97368421, 0.96491228, 0.96491228, 0.96491228, 0.98245614,
            0.96491228, 0.96491228, 0.96491228, 0.97368421, 0.96491228,
            0.97368421, 0.96491228, 0.96491228, 0.96491228, 0.96491228,
            0.96491228, 0.96491228, 0.96491228, 0.96491228, 0.96491228,
            0.96491228, 0.96491228, 0.97368421, 0.96491228])

```

```

[26]: print( "The best accuracy was with", mean_acc.max(), "with k=", mean_acc.
    ↪argmax()+1)

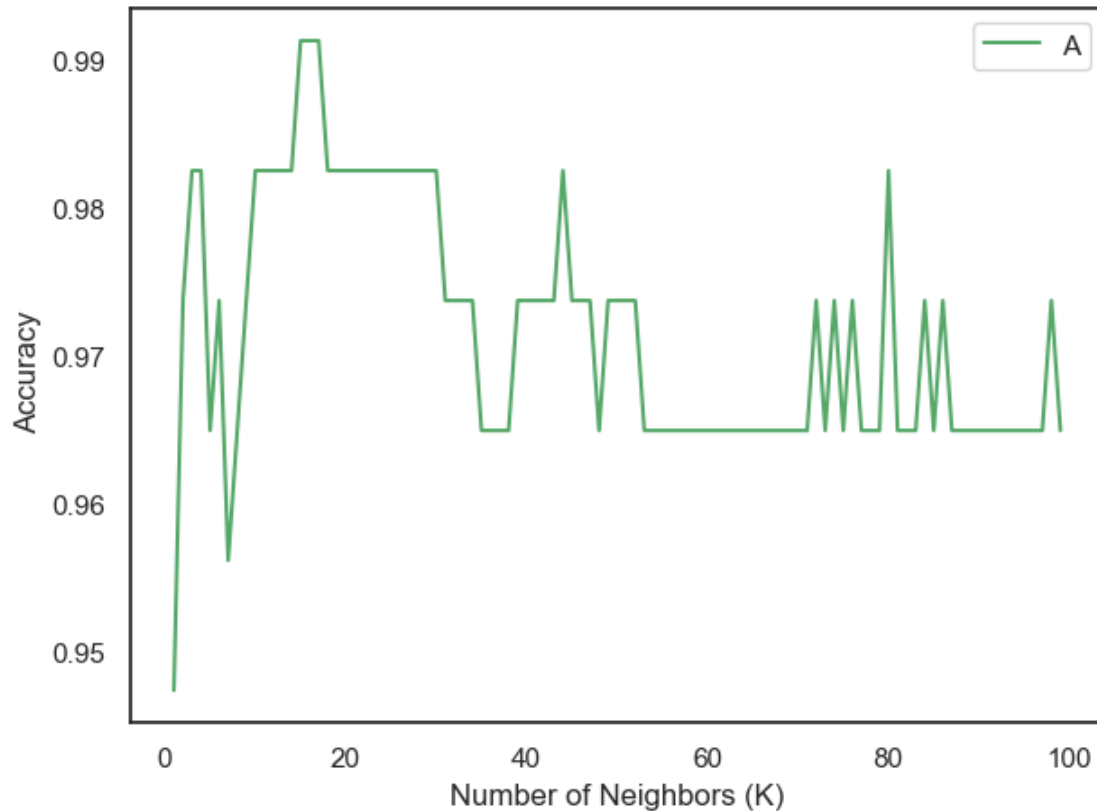
```

The best accuracy was with 0.9912280701754386 with k= 15

```

[27]: plt.plot(range(1,Ks),mean_acc,'g')
      plt.legend(('Accuracy '))
      plt.ylabel('Accuracy ')
      plt.xlabel('Number of Neighbors (K)')
      plt.tight_layout()
      plt.show()

```



```
[28]: final_k = 15
neigh = KNeighborsClassifier(n_neighbors = final_k).fit(X_train,y_train.values.
↪ravel())
yhat=neigh.predict(X_test)
```

```
[29]: jaccard_score = metrics.jaccard_score(y_test, yhat)
f1_score = metrics.f1_score(y_test, yhat)
print(f"Jaccard score: {jaccard_score}\nF1_Score: {f1_score}")
```

Jaccard score: 0.9705882352941176

F1_Score: 0.9850746268656716

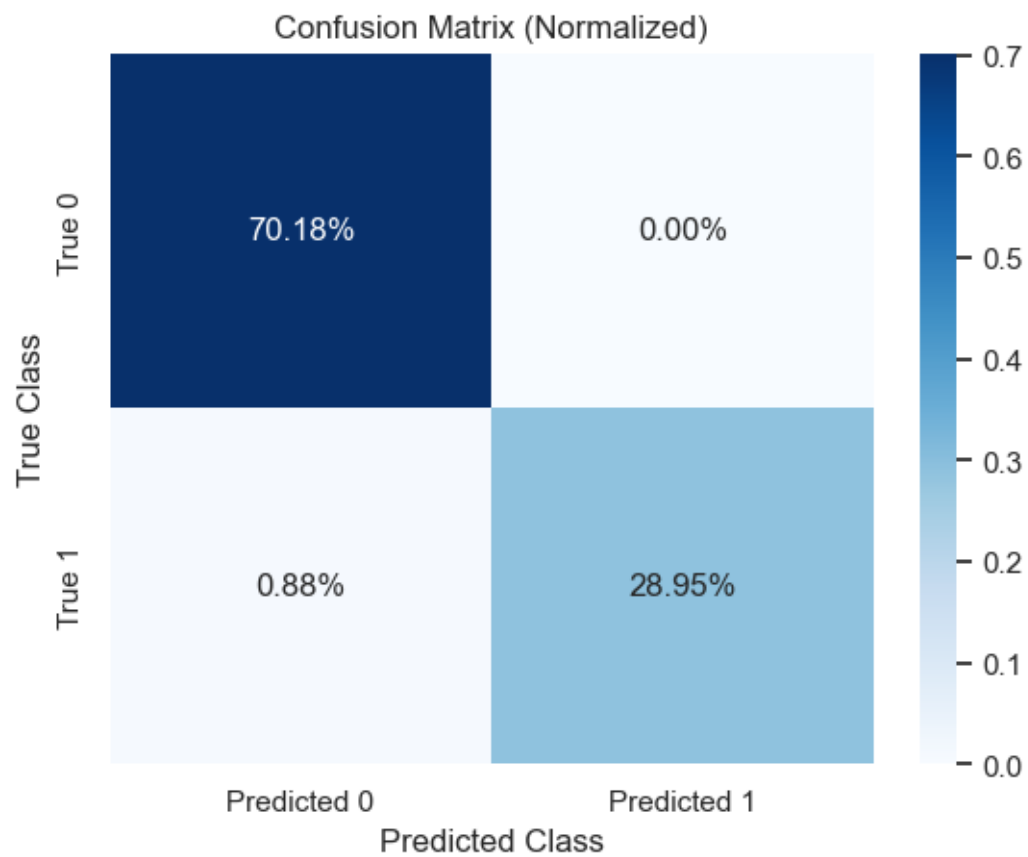
```
[30]: cf_matrix = metrics.confusion_matrix(y_test, yhat)

normalized_cf_matrix = cf_matrix / np.sum(cf_matrix)

sns.heatmap(normalized_cf_matrix, annot=True, fmt='.2%', cmap='Blues',
            xticklabels=['Predicted ' + str(label) for label in np.
↪unique(yhat)],
            yticklabels=['True ' + str(label) for label in np.unique(y_test)])
```

```
plt.xlabel('Predicted Class')
plt.ylabel('True Class')

plt.title('Confusion Matrix (Normalized)')
plt.show()
```



[]: