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Research Article

NON-PARAMETRIC METHODS. K-NEAREST NEIGHBORS MODEL

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ABSTRACT

One of the machine learning algorithms k-Nearest Neighbors algorithm is widely used for classification tasks in the construction of artificial intelligence programs. In the k-NN algorithm, when determining which class a new object belongs to, the distances from this object to all objects are measured, and if there are more objects belonging to which class among the nearest k selected objects, the new object is considered to belong to this class, which makes it an intuitive and powerful tool for solving complex problems. In this article, a model for determining whether a patient is diagnosed with breast cancer or not is created using the k-NN algorithm. This problem is calculated using binary classification.

KEYWORDS

Dataset, testset, trainset, hyperparameters, classification, prediction.

INTRODUCTION

What is classification?

- Supervised Machine Learning type

- Classification of unknown elements into categories (classes)
- Classifiers can be of two types.
- Binary classifiers
- Multiclass classifiers

Generating k-Nearest Neighbors methods using scikit-learn

Implementation of breast cancer detection algorithm using k nearest neighbors method.

Description: Breast cancer is the most common cancer among women in the world. It accounts for

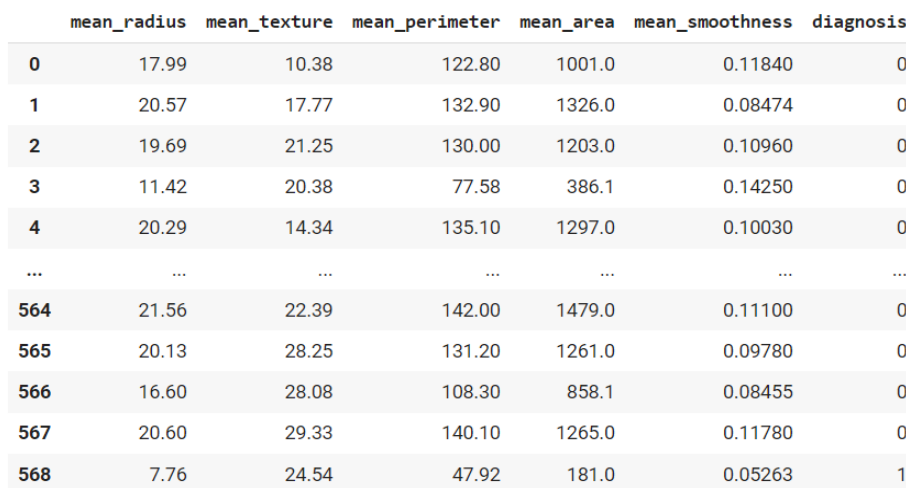
25% of all cancer cases. Breast cancer begins when cells in the breast grow out of control. These cells are usually detected by analyzing tumors that can be seen on X-rays.

```
import pandas as pd
```

```
import numpy as np
```

```
df=pd.read_csv("/content/Breast_cancer_data.csv")
```

```
df
```



	mean_radius	mean_texture	mean_perimeter	mean_area	mean_smoothness	diagnosis
0	17.99	10.38	122.80	1001.0	0.11840	0
1	20.57	17.77	132.90	1326.0	0.08474	0
2	19.69	21.25	130.00	1203.0	0.10960	0
3	11.42	20.38	77.58	386.1	0.14250	0
4	20.29	14.34	135.10	1297.0	0.10030	0
...	...	...	...	...	...	...
564	21.56	22.39	142.00	1479.0	0.11100	0
565	20.13	28.25	131.20	1261.0	0.09780	0
566	16.60	28.08	108.30	858.1	0.08455	0
567	20.60	29.33	140.10	1265.0	0.11780	0
568	7.76	24.54	47.92	181.0	0.05263	1

569 rows x 6 columns

```
df.shape
```

Result:

(569,32)

```
df['diagnosis'].value_counts()
```

Result:

B 357

M 212

Name: diagnosis, dtype: int64

Let's change these values to 0 and 1. M->1, B->0
For this, you can use the **LabelEncoder** in sklearn or the **.replace()** method in pandas. Both ways are shown below.

```
from sklearn.preprocessing import LabelEncoder
```

```
labelencoder = LabelEncoder()
```

1 212

```
df['diagnosis'] =  
labelencoder.fit_transform(df['diagnosis'].values  
)
```

Name: diagnosis, dtype: int64

Correlation (linear relationship) can be seen in the graph below

```
df['diagnosis'].value_counts()
```

```
corr_matrix = df.corr().abs()
```

Result:

```
corr_matrix.style.background_gradient(cmap='coolwarm')
```

0 357

	mean_radius	mean_texture	mean_perimeter	mean_area	mean_smoothness	diagnosis
mean_radius	1.000000	0.323782	0.997855	0.987357	0.170581	0.730029
mean_texture	0.323782	1.000000	0.329533	0.321086	0.023389	0.415185
mean_perimeter	0.997855	0.329533	1.000000	0.986507	0.207278	0.742636
mean_area	0.987357	0.321086	0.986507	1.000000	0.177028	0.708984
mean_smoothness	0.170581	0.023389	0.207278	0.177028	1.000000	0.358560
diagnosis	0.730029	0.415185	0.742636	0.708984	0.358560	1.000000

```
df.corrwith(df["diagnosis"]).abs().sort_values(ascending=False)
```

```
y=df["diagnosis"]
```

Using the module below, values are standardized because all column values are in different ranges.

```
diagnosis 1.000000
```

```
from sklearn.preprocessing import  
StandardScaler
```

```
mean_perimeter 0.742636
```

```
scaler=StandardScaler()
```

```
mean_radius 0.730029
```

```
X=scaler.fit_transform(X)
```

```
mean_area 0.708984
```

The train_test_split module below is used to split the DataSet into train_set and test_set.

```
mean_texture 0.415185
```

```
from sklearn.model_selection import  
train_test_split
```

```
mean_smoothness 0.358560
```

```
dtype: float64
```

```
X_train, X_test, y_train, y_test  
=train_test_split(X,y,test_size=0.20,
```

Data is extracted:

```
X=df.drop("diagnosis", axis=1).values
```

```
random_state=12)
```

```
1 1 1]
```

The k-NN model is recalled.

Predicted values:

```
from sklearn.neighbors import
KNeighborsClassifier
```

```
[0 1 1 1 1 0 1 1 1 0 1 0 0 0 0 1 1 0 1 1 1 1 1 0 0 1
0 1 0 1 0 0 1 0 1 1
```

```
knn=KNeighborsClassifier(n_neighbors=5)
```

```
0 0 1 1 1 1 1 1 0 0 1 1 1 1 1 1 0 0 1 1 1 0 0 1 0 1 1
0 1 1 1 0 1 1 1 0 0
```

X_train and Y_train data are fed to the model to train the model.

```
0 0 1 1 1 0 0 1 0 1 1 1 1 0 0 0 0 0 1 1 1 1 1 1 1 1 1
1 0 1 1 1 0 0 0 0 1
```

```
knn.fit(X_train, y_train)
```

```
1 1 1]
```

X_test data allocated for testing is passed to find predictive values

```
1 1 1]
```

Model evaluation.

Predictive values are found using the .predict() function.

Evaluation by the Jaccard index evaluation criterion:

```
y_prdict=knn.predict(X_test)
```

```
from sklearn.metrics import jaccard_score
```

Actual values and predicted values are as follows.

```
jaccard_score(y_test, y_prdict)
```

```
print(y_test.to_numpy())
```

Result:

```
print(y_prdict)
```

```
0.8958333333333334
```

Result:

Confusion matrix evaluation criteria:

Original values:

```
from sklearn.metrics import confusion_matrix
```

```
[0 1 1 1 1 1 0 1 1 0 1 0 0 0 0 1 1 0 1 1 1 0 1 1 0 0 1
0 1 0 1 0 0 1 0 1 1
```

```
import seaborn as sns
```

```
0 0 0 1 1 1 1 1 0 0 0 1 1 1 1 1 0 0 1 1 1 0 0 1 0 0 0
0 1 1 1 0 1 1 1 0 0
```

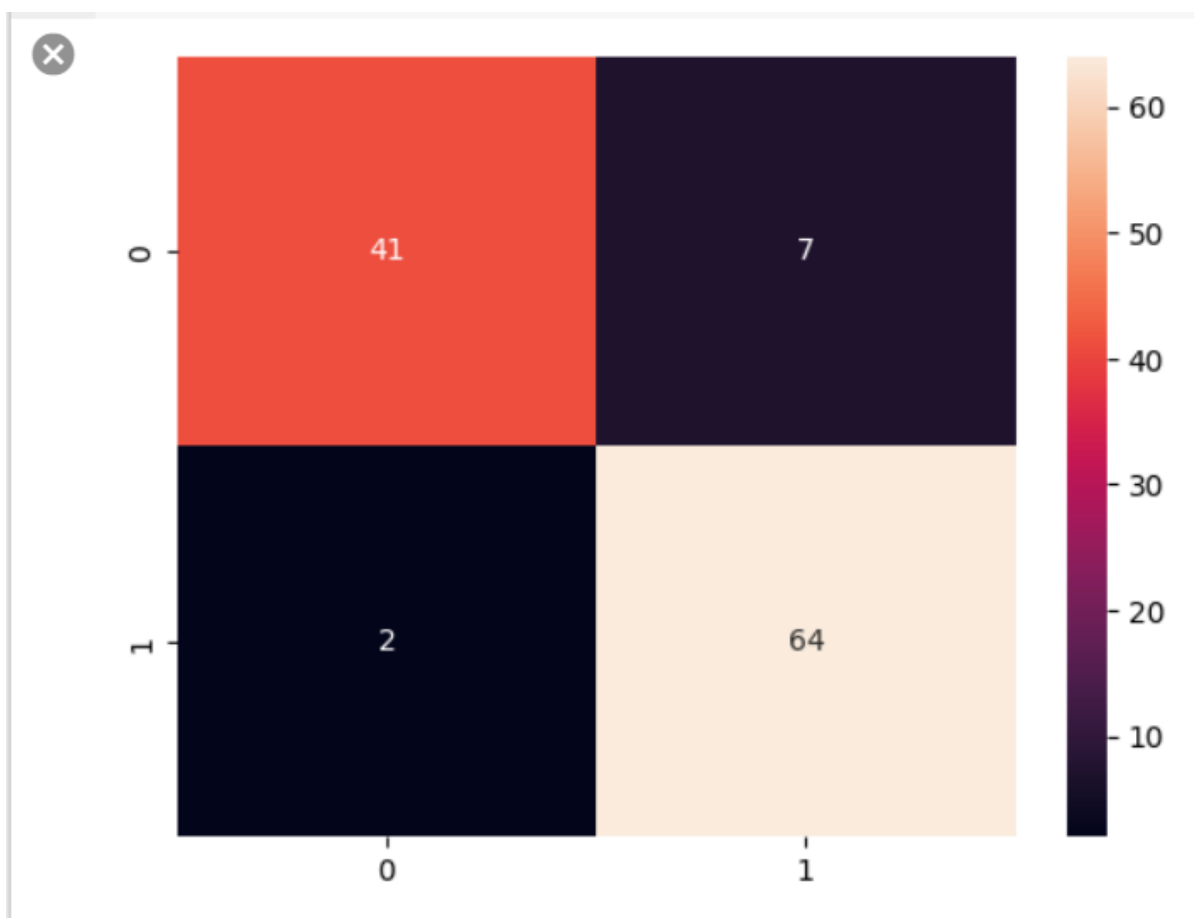
```
import matplotlib.pyplot as plt
```

```
0 0 1 1 1 0 0 1 0 1 1 1 1 0 0 0 0 0 0 1 1 1 1 1 1 1 1
1 0 1 1 1 0 1 0 0 1
```

```
sns.heatmap(confusion_matrix(y_test, y_prdict),
annot=True)
```

```
plt.show()
```

Result:



```
confusion_matrix(y_test, y_pred)
```

Result:

```
array([[41, 7],  
       [ 2, 64]])
```

Calculate Precision, Recall, F1 and Accuracy

```
from sklearn.metrics import precision_score,  
recall_score, f1_score, accuracy_score
```

```
precision=precision_score(y_test, y_pred)
```

```
recall=recall_score(y_test, y_pred)
```

```
f1=f1_score(y_test, y_pred)
```

```
accuracy=accuracy_score(y_test, y_pred)
```

```
print(f"Precision: {precision}")
```

```
print(f"Recall: {recall}")
```

```
print(f"f1: {f1}")
```

```
print(f"Accuracy: {accuracy}")
```

Result:

Precision: 0.9014084507042254

Recall: 0.9696969696969697

f1: 0.9343065693430657

Accuracy: 0.9210526315789473

We can actually get this information in general view.

```
from sklearn.metrics import classification_report
```

```
a=classification_report(y_test, y_predect)
```

```
print(a)
```

	precision	recall	f1-score	support
0	0.95	0.85	0.90	48
1	0.90	0.97	0.93	66
accuracy			0.92	114
macro avg	0.93	0.91	0.92	114
weighted avg	0.92	0.92	0.92	114

Selecting the best k by k nearest neighbors method

One of the most important parameters in the k-NN algorithm is the value of k. That is, the number of neighbors is important in determining whether a new object belongs to a certain class.

It is possible to find the best value of k by calculating the values of f1_score or

accuracy_score determined using the evaluation criterion seen above, by calculating at which value of k of the model the model achieves the best efficiency.

```
f11=[]
```

```
for k in range(1,25):
```

```
    knn=KNeighborsClassifier(n_neighbors=k) # k-  
    ni qiymati
```

```
    knn.fit(X_train, y_train)
```

```
    y_predect1=knn.predict(X_test)
```

```
    f11.append(accuracy_score(y_test,y_predect1))
```

```
plt.figure(figsize=(10,6))
```

```
plt.plot(range(1,25),f11,marker='o', linestyle='-',  
color='b')
```

```
plt.xticks(range(1,25))
```

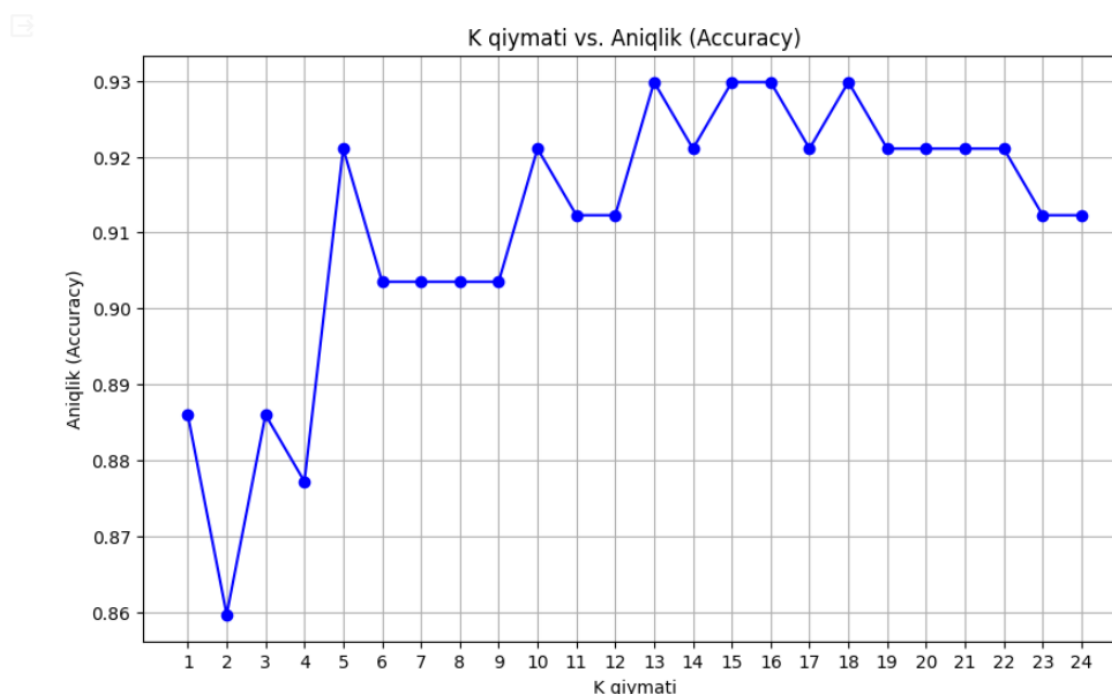
```
plt.title('K qiymati vs. Aniqlik (Accuracy)')
```

```
plt.xlabel('K qiymati')
```

```
plt.ylabel('Aniqlik (Accuracy)')
```

```
plt.grid()
```

```
plt.show()
```



As can be seen from the graph, the model achieves the highest accuracy at values $k=\{13,19\}$.

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