

# Final Report : Solving Decision Tree Algorithm and Solving K-Nearest Neighbour Algorithm

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**Abstract—Code 1 :**Decision Tree algorithm belongs to the family of supervised learning algorithms. Unlike other supervised learning algorithms, the decision tree algorithm can be used for solving regression and classification problems too

**Code 2 :**The k-nearest neighbors (KNN) algorithm is a simple, supervised machine learning algorithm that can be used to solve both classification and regression problems. It's easy to implement and understand, but has a major drawback of becoming significantly slower as the size of that data in use grows.

**Index Terms—**Python, C++

## I. INTRODUCTION

Decision Tree algorithm belongs to the family of supervised learning algorithms. The goal of using a Decision Tree is to create a training model that can use to predict the class or value of the target variable by learning simple decision rules inferred from prior data

KNN is a lazy learning, non-parametric algorithm. It uses data with several classes to predict the classification of the new sample point.

## II. WHY KNN ALGORITHM AND DECISION TREE ALGORITHM IS USED?

Decision tree builds classification or regression models in the form of a tree structure. It breaks down a dataset into smaller and smaller subsets while at the same time an associated decision tree is incrementally developed. Decision trees can handle both categorical and numerical data.

The KNN algorithm can compete with the most accurate models because it makes highly accurate predictions. The quality of the predictions depends on the distance measure.

## III. TYPES OF DECISION TREE AND KNN ALGORITHM

There are 4 popular types of decision tree algorithms:

- 1.ID3,
2. CART (Classification and Regression Trees),
3. Chi-Square
- 4.Reduction in Variance.

The k-nearest neighbors (KNN) algorithm is a simple, supervised machine learning algorithm that can be used to solve both classification and regression problems.

## IV. RULES FOR SOLVING DECISION TREE ALGORITHM

Decision tree uses the tree representation to solve the problem in which each leaf node corresponds to a class label and attributes are represented on the internal node of the tree.

## V. CALCULATED A KNN ALGORITHM

Here is step by step on how to compute K-nearest neighbors KNN algorithm:

1. Determine parameter K = number of nearest neighbors.
2. Calculate the distance between the query-instance and all the training samples.

## VI. REPORT CODE OF DECISION TREE ALGORITHM

```
1 import pandas as pd
2 import numpy as np
3
4 df = pd.read_csv("dataset.csv")
5
6 df
7
8 X = df.iloc[:, :-1]
9
10 X
11
12 y = df.iloc[:, 3]
13
14 y
15
16 from sklearn.preprocessing import LabelEncoder
17
18 LabelEncoder_X = LabelEncoder()
19
20 X = X.apply(LabelEncoder_X.fit_transform)
21
22 X
23
24 from sklearn.tree import DecisionTreeClassifier
25
```

Fig. 1.

```
26 from sklearn.tree import DecisionTreeClassifier
27
28 regressor = DecisionTreeClassifier()
29 regressor.fit(X.iloc[:, 1:4], y)
30
31 X_in = np.array([1, 0, 1])
32
33 y_pred = regressor.predict([X_in])
34
35 y_pred
```

Fig. 2.

## VII. REPORT CODE OF KNN ALGORITHM ALGORITHM

```
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In [2]: # Import necessary modules
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_iris

# Loading data
irisData = load_iris()

# Create feature and target arrays
X = irisData.data
y = irisData.target

# Split into training and test set
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size = 0.2, random_state=42)

knn = KNeighborsClassifier(n_neighbors=7)
knn.fit(X_train, y_train)

# Predict on dataset which model has not seen before
print(knn.predict(X_test))

[1 0 2 1 1 0 1 2 1 2 0 0 0 1 2 1 1 2 0 2 0 2 2 2 2 0 0]
```

Fig. 3.

```
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In [3]: # Import necessary modules
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_iris

# Loading data
irisData = load_iris()

# Create feature and target arrays
X = irisData.data
y = irisData.target

# Split into training and test set
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size = 0.2, random_state=42)

knn = KNeighborsClassifier(n_neighbors=7)
knn.fit(X_train, y_train)

# Calculate the accuracy of the model
print(knn.score(X_test, y_test))

0.9666666666666667
```

Fig. 4.

```
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In [4]: # Import necessary modules
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_iris
import numpy as np
import matplotlib.pyplot as plt

irisData = load_iris()

# Create feature and target arrays
X = irisData.data
y = irisData.target

# Split into training and test set
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size = 0.2, random_state=42)

neighbors = np.arange(1, 9)
train_accuracy = np.empty(len(neighbors))
test_accuracy = np.empty(len(neighbors))

# Loop over K values
for i, k in enumerate(neighbors):
    knn = KNeighborsClassifier(n_neighbors=k)
    knn.fit(X_train, y_train)

# Compute training and test data accuracy
train_accuracy[i] = knn.score(X_train, y_train)
test_accuracy[i] = knn.score(X_test, y_test)

# Generate plot
plt.plot(neighbors, test_accuracy, label='Testing Accuracy (KNN)')
```

Fig. 5.

## VIII. ASSIGNMENT OUTPUT ( DECISION TREE ALGORITHM PROBLEM SOLVING)

```
localhost:8888/notebooks/Desktop/desicion%20tree%20all/DC_Tree_Classfier.ipynb

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In [8]: import pandas as pd
import numpy as np

In [9]: dataset = pd.read_csv("dtdata.csv")

In [10]: dataset

Out[10]:
```

	Income	gender	Metastus	Ages
0	High	Male	Single	<21
1	High	Male	Married	<21
2	High	Male	Single	21-35
3	Medium	Male	Single	Weak
4	Low	Female	Single	>35
5	Low	Female	Married	>35
6	Low	Female	Married	21-35
7	Medium	Male	Single	>21
8	Low	Female	Married	<21
9	Medium	Female	Single	>35
10	Medium	Female	Married	<21
11	Medium	Male	Married	21-35
12	High	Female	Single	21-35
13	Medium	Male	Married	>35

```
In [ ]:
```

Fig. 6.

```
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13 Medium Male Married

In [17]: y= dataset.iloc[:,3]

In [18]: y

Out[18]:
```

0	<21
1	<21
2	21-35
3	Weak
4	>35
5	>35
6	21-35
7	>21
8	<21
9	>35
10	<21
11	21-35
12	21-35
13	>35

```
Name: Ages, dtype: object

In [ ]:
```

Fig. 7.

```
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In [23]: from sklearn.preprocessing import LabelEncoder

In [24]: LabelEncoder_X = LabelEncoder()

In [25]: X = X.apply(LabelEncoder().fit_transform)

In [26]: X

Out[26]:
```

	Income	gender	Metastus
0	0	1	1
1	0	1	0
2	0	1	1
3	2	1	1
4	1	0	1
5	1	0	0
6	1	0	0
7	2	1	1
8	1	0	0
9	2	0	1
10	2	0	0
11	2	1	0
12	0	0	1
13	2	1	0

```
In [27]: from sklearn.tree import DecisionTreeClassifier

In [28]: regressor = DecisionTreeClassifier()

In [29]: regressor.fit(X.iloc[:,1:4], y)

Out[29]: DecisionTreeClassifier()
```

Fig. 8.

## IX. ASSIGNMENT OUTPUT ( KNN ALGORITHM OUTPUT)

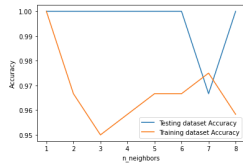


Fig. 9.

## ACKNOWLEDGMENT

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## REFERENCES

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