

Project Report  
On  
**Drug Classification Model**

Submitted in partial fulfilment of the requirements for the award of

**BACHELOR OF TECHNOLOGY**  
in  
**COMPUTER SCIENCE & ENGINEERING**  
(Artificial Intelligence & Machine Learning)  
by

**Ms. G TEJASWINI (22WH1A6608)**

**Ms. G ANUSHA (22WH1A6641)**

**Ms. G VAISHNAVI (22WH1A6658)**

**Ms. B ANITHA (22WH1A6659)**

**Under the esteemed guidance of**  
**Ms. A Naga Kalyani**  
**Assistant Professor, CSE(AI&ML)**



**Department of Computer Science & Engineering**  
**(Artificial Intelligence & Machine Learning)**

**BVRIT HYDERABAD COLLEGE OF ENGINEERING FOR WOMEN**

**(Approved by AICTE, New Delhi and Affiliated to JNTUH, Hyderabad)**

**Accredited by NBA and NAAC with A Grade**

**Bachupally, Hyderabad – 500090**

**2024-25**

**Department of Computer Science & Engineering**  
**(Artificial Intelligence & Machine Learning)**  
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**Bachupally, Hyderabad – 500090**  
**2024-25**



**CERTIFICATE**

This is to certify that the major project entitled “**Drug Classification Model**” is a bonafide work carried out by **Ms. G. Tejaswini (22WH1A6608), Ms. G. Anusha (22WH1A6641), Ms. G. Vaishnavi (22WH1A6658), Ms. B. Anitha (22WH1A6659)** in partial fulfilment for the award of B. Tech degree in **Computer Science & Engineering (AI&ML), BVRIT HYDERABAD College of Engineering for Women, Bachupally, Hyderabad**, affiliated to Jawaharlal Nehru Technological University Hyderabad, Hyderabad under my guidance and supervision. The results embodied in the project work have not been submitted to any other University or Institute for the award of any degree or diploma.

**Supervisor**

**Ms. A Naga Kalyani**  
**Assistant Professor**  
**Dept of CSE(AI&ML)**

**Head of the Department**

**Dr. B. Lakshmi Praveena**  
**HOD & Professor**  
**Dept of CSE(AI&ML)**

**External Examiner**

## DECLARATION

We hereby declare that the work presented in this project entitled “**Drug Classification Model**” submitted towards completion of Project work in III Year of B.Tech of CSE(AI&ML) at **BVRIT HYDERABAD College of Engineering for Women**, Hyderabad is an authentic record of our original work carried out under the guidance of **Ms. A Naga Kalyani, Assistant Professor, Department of CSE(AI&ML)**.

Sign with Date:

G. Tejaswini  
(22WH1A6608)

Sign with Date:

G. Anusha  
(22WH1A6641)

Sign with Date:

G. Vaishnavi  
(22WH1A6658)

Sign with Date:

B. Anitha  
(22WH1A6659)

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We are extremely thankful to our Internal Guide, **Ms. A Naga Kalyani, Assistant Professor, CSE(AI&ML), BVRIT HYDERABAD College of Engineering for Women**, for her constant guidance and encouragement throughout the project.

Finally, we would like to thank our Major Project Coordinator, all Faculty and Staff of CSE(AI&ML) department who helped us directly or indirectly. Last but not least, we wish to acknowledge our **Parents** and **Friends** for giving moral strength and constant encouragement.

**G. Tejaswini (22WH1A6608)**

**G. Anusha (22WH1A6641)**

**G. Vaishnavi (22WH1A6658)**

**B. Anitha (22WH1A6659)**

## **ABSTRACT**

This project focuses on developing a drug classification model that uses machine learning techniques to predict the appropriate drug for a patient based on their health parameters. The dataset used includes features such as age, sex, blood pressure, cholesterol levels, and the sodium-to-potassium ratio, with the target variable being the prescribed drug. The model employs Decision Tree and Random Forest algorithms to classify the drugs into different categories. The process involves data preprocessing, such as encoding categorical variables, followed by model training and evaluation using metrics like accuracy, precision, recall, and F1-score. The model's performance is validated using a test dataset, achieving high accuracy in drug classification. This system can assist healthcare professionals in selecting suitable medications, promoting more accurate and timely drug prescriptions while reducing human errors in decision-making.

## **PROBLEM STATEMENT**

The goal of this project is to develop a drug classification model that predicts the therapeutic category of a drug based on patient-specific data, including age, sex, blood pressure, cholesterol levels, and sodium-to-potassium ratio. Accurate drug classification is crucial in medical practice, as it helps healthcare professionals select the most appropriate medication for patients. To achieve this, the project utilizes machine learning algorithms, specifically Decision Trees and Random Forest Classifiers, to train a model on a dataset of patient information and their corresponding drug prescriptions. The model can then predict the drug category for new patients, providing automated and reliable assistance in drug prescription.

## DATASET

| 1  | Age | Sex | BP     | Cholesterol | Na_to_K | Drug  |
|----|-----|-----|--------|-------------|---------|-------|
| 2  | 23  | F   | HIGH   | HIGH        | 25.355  | drugY |
| 3  | 47  | M   | LOW    | HIGH        | 13.093  | drugC |
| 4  | 47  | M   | LOW    | HIGH        | 10.114  | drugC |
| 5  | 28  | F   | NORMAL | HIGH        | 7.798   | drugX |
| 6  | 61  | F   | LOW    | HIGH        | 18.043  | drugY |
| 7  | 22  | F   | NORMAL | HIGH        | 8.607   | drugX |
| 8  | 49  | F   | NORMAL | HIGH        | 16.275  | drugY |
| 9  | 41  | M   | LOW    | HIGH        | 11.037  | drugC |
| 10 | 60  | M   | NORMAL | HIGH        | 15.171  | drugY |
| 11 | 43  | M   | LOW    | NORMAL      | 19.368  | drugY |
| 12 | 47  | F   | LOW    | HIGH        | 11.767  | drugC |
| 13 | 34  | F   | HIGH   | NORMAL      | 19.199  | drugY |
| 14 | 43  | M   | LOW    | HIGH        | 15.376  | drugY |
| 15 | 74  | F   | LOW    | HIGH        | 20.942  | drugY |
| 16 | 50  | F   | NORMAL | HIGH        | 12.703  | drugX |
| 17 | 16  | F   | HIGH   | NORMAL      | 15.516  | drugY |
| 18 | 69  | M   | LOW    | NORMAL      | 11.455  | drugX |
| 19 | 43  | M   | HIGH   | HIGH        | 13.972  | drugA |
| 20 | 23  | M   | LOW    | HIGH        | 7.298   | drugC |

**DataSet Link:** [DataSet](#)

## Code

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

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

## Code

```
def calculate_additional_metrics(model_name, y_true, y_pred):
    precision = metrics.precision_score(y_true,
y_pred,average='weighted')

    recall = metrics.recall_score(y_true, y_pred,average='weighted')

    f1_score = metrics.f1_score(y_true, y_pred,average='weighted')

    print(f"{model_name} - Additional Metrics:")

    print(f"Precision: {precision}")

    print(f"Recall: {recall}")

    print(f"F1-Score: {f1_score}\n")
```

## Code

```
ds=pd.read_csv("/content/drug200.csv")

ds.head()
```

## Output

|   | Age | Sex | BP     | Cholesterol | Na_to_K | Drug  |
|---|-----|-----|--------|-------------|---------|-------|
| 0 | 23  | F   | HIGH   | HIGH        | 25.355  | drugY |
| 1 | 47  | M   | LOW    | HIGH        | 13.093  | drugC |
| 2 | 47  | M   | LOW    | HIGH        | 10.114  | drugC |
| 3 | 28  | F   | NORMAL | HIGH        | 7.798   | drugX |
| 4 | 61  | F   | LOW    | HIGH        | 18.043  | drugY |





## Code

```
print("Description:\n",ds.describe())
```

## Output

```
Description:
           Age      Na_to_K
count  200.000000  200.000000
mean    44.315000   16.084485
std     16.544315    7.223956
min     15.000000    6.269000
25%     31.000000   10.445500
50%     45.000000   13.936500
75%     58.000000   19.380000
max     74.000000   38.247000
```

## Code

```
print("null values:\n",ds.isnull().sum())
```

## Output

```
null values:
Age      0
Sex      0
BP       0
Cholesterol  0
Na_to_K   0
Drug     0
dtype: int64
```

## Code

```
print("data:\n",ds)
```

## Output

```
data:
   Age Sex  BP Cholesterol  Na_to_K  Drug
0   23  F  HIGH         HIGH   25.355 drugY
1   47  M  LOW         HIGH   13.093 drugC
2   47  M  LOW         HIGH   10.114 drugC
3   28  F  NORMAL       HIGH    7.798 drugX
4   61  F  LOW         HIGH   18.043 drugY
..  ...  ..  ...         ...     ...  ...
195  56  F  LOW         HIGH   11.567 drugC
196  16  M  LOW         HIGH   12.006 drugC
197  52  M  NORMAL       HIGH    9.894 drugX
198  23  M  NORMAL       NORMAL  14.020 drugX
199  40  F  LOW         NORMAL  11.349 drugX

[200 rows x 6 columns]
```

## Code

```
from sklearn import preprocessing

sex_encoding=preprocessing.LabelEncoder()

ds['Sex'] = sex_encoding.fit_transform(ds['Sex'])

BP_encoding=preprocessing.LabelEncoder()

ds['BP']=BP_encoding.fit_transform(ds['BP'])

Cholesterol_encoding=preprocessing.LabelEncoder()

ds['Cholesterol']=Cholesterol_encoding.fit_transform(ds['Cholesterol'])

#display the preprocessed dataset

print(ds)
```

## Output

|     | Age | Sex | BP | Cholesterol | Na_to_K | Drug  |
|-----|-----|-----|----|-------------|---------|-------|
| 0   | 23  | 0   | 0  | 0           | 25.355  | drugY |
| 1   | 47  | 1   | 1  | 0           | 13.093  | drugC |
| 2   | 47  | 1   | 1  | 0           | 10.114  | drugC |
| 3   | 28  | 0   | 2  | 0           | 7.798   | drugX |
| 4   | 61  | 0   | 1  | 0           | 18.043  | drugY |
| ..  | ... | ... | .. | ...         | ...     | ...   |
| 195 | 56  | 0   | 1  | 0           | 11.567  | drugC |
| 196 | 16  | 1   | 1  | 0           | 12.006  | drugC |
| 197 | 52  | 1   | 2  | 0           | 9.894   | drugX |
| 198 | 23  | 1   | 2  | 1           | 14.020  | drugX |
| 199 | 40  | 0   | 1  | 1           | 11.349  | drugX |

[200 rows x 6 columns]

## Code

```
xcols=[col for col in ds.columns if col not in ['Drug']]

x=ds[xcols]

y=ds['Drug']
```

## Code

```
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_s
tate=3)
```

## Code

```
from sklearn.tree import DecisionTreeClassifier

dectree=DecisionTreeClassifier()
```

```
dectree.fit(x_train,y_train)

dr_prediction=dectree.predict(x_test)

print("DecisionTrees's Accuracy: ", metrics.accuracy_score(y_test,
dr_prediction))

calculate_additional_metrics('Decision Tree', y_test, dr_prediction)
```

## Output

```
DecisionTrees's Accuracy:  0.9833333333333333
Decision Tree - Additional Metrics:
Precision: 0.9840579710144927
Recall: 0.9833333333333333
F1-Score: 0.9833152664859981
```

## Code

```
from sklearn.ensemble import RandomForestClassifier

ranforest = RandomForestClassifier()

ranforest.fit(x_train, y_train)

rf_prediction= ranforest.predict(x_test)

print("Random Forest's Accuracy:", metrics.accuracy_score(y_test,
rf_prediction))

calculate_additional_metrics('Random Forest', y_test, rf_prediction)
```

## Output

```
Random Forest's Accuracy: 0.9833333333333333
Random Forest - Additional Metrics:
Precision: 0.9840579710144927
Recall: 0.9833333333333333
F1-Score: 0.9833152664859981
```

## Code

```
import pandas as pd

import ipywidgets as widgets

from IPython.display import display

# Create widgets with styled labels
```

```

age_label = widgets.Label(value="Age:", style={'font_size': '16px'})
age_widget = widgets.IntText(value=0,
layout=widgets.Layout(width='200px'))

sex_label = widgets.Label(value="Sex:", style={'font_size': '16px'})
sex_widget = widgets.Dropdown(options=["M", "F"],
layout=widgets.Layout(width='200px'))

bp_label = widgets.Label(value="Blood Pressure (BP):",
style={'font_size': '16px'})
bp_widget = widgets.Dropdown(options=["LOW", "NORMAL", "HIGH"],
layout=widgets.Layout(width='200px'))

cholesterol_label = widgets.Label(value="Cholesterol:",
style={'font_size': '16px'})
cholesterol_widget = widgets.Dropdown(options=["NORMAL", "HIGH"],
layout=widgets.Layout(width='200px'))

na_to_k_label = widgets.Label(value="Na_to_K:", style={'font_size':
'16px'})
na_to_k_widget = widgets.FloatText(value=0.0,
layout=widgets.Layout(width='200px'))

submit_button = widgets.Button(description="Submit",
                                button_style='success', # 'success',
'info', 'warning', 'danger', or ''
                                layout=widgets.Layout(width='200px',
margin='10px 0px'))

output = widgets.Output()

# Header
header = widgets.HTML(
    value="<h2 style='text-align: center;'>Patient Data Input</h2>",
    layout=widgets.Layout(margin='10px 0px')
)

# Submit button callback

```

```

def on_submit_clicked(b):
    with output:
        output.clear_output()

        age = age_widget.value

        sex = sex_widget.value

        bp = bp_widget.value

        cholesterol = cholesterol_widget.value

        na_to_k = na_to_k_widget.value

        new_patient_data = pd.DataFrame({
            'Age': [age],
            'Sex': [sex],
            'BP': [bp],
            'Cholesterol': [cholesterol],
            'Na_to_K': [na_to_k]
        })

        print("Submitted Data:")

        print(new_patient_data)

submit_button.on_click(on_submit_clicked)

# Arrange widgets in a vertical box

form = widgets.VBox([
    header,

    widgets.HBox([age_label, age_widget],
layout=widgets.Layout(justify_content='space-between', width='400px')),

    widgets.HBox([sex_label, sex_widget],
layout=widgets.Layout(justify_content='space-between', width='400px')),

    widgets.HBox([bp_label, bp_widget],
layout=widgets.Layout(justify_content='space-between', width='400px')),

    widgets.HBox([cholesterol_label, cholesterol_widget],
layout=widgets.Layout(justify_content='space-between', width='400px')),

```

```

        widgets.HBox([na_to_k_label, na_to_k_widget],
layout=widgets.Layout(justify_content='space-between', width='400px')),

        submit_button,

        output

], layout=widgets.Layout(align_items='center', width='50%',
margin='auto'))

# Display the form

display(form)

```

## Output

Patient Data Input

Age:

30

Sex:

F

▼

Blood Pressure (BP):

HIGH

▼

Cholesterol:

NORMAL

▼

Na\_to\_K:

15

Submit

Submitted Data:

|   | Age | Sex | BP   | Cholesterol | Na_to_K |
|---|-----|-----|------|-------------|---------|
| 0 | 30  | F   | HIGH | NORMAL      | 15.0    |

---

## Code

```

new_patient_data['Sex'] =
sex_encoding.fit_transform(new_patient_data['Sex'])

new_patient_data['BP'] =
BP_encoding.fit_transform(new_patient_data['BP'])

new_patient_data['Cholesterol'] =
Cholesterol_encoding.fit_transform(new_patient_data['Cholesterol'])

print("New Patient Data:")

print(new_patient_data)

```

## Output

```
New Patient Data:  
  Age  Sex  BP  Cholesterol  Na_to_K  
0   30   0   0             0      15.0
```

## Code

```
dt_prediction=dectree.predict(new_patient_data)  
rf_prediction = ranforest.predict(new_patient_data)  
print("\nPredictions for the New Patient:")  
print("Decision Tree Prediction:", dt_prediction[0])  
print("Random Forest Prediction:", rf_prediction[0])
```

## Output

```
Predictions for the New Patient:  
Decision Tree Prediction: drugY  
Random Forest Prediction: drugY
```





