

▼ Import Libraries

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

▼ Import Data

```
dataset = pd.read_csv('patient_data.csv')
# dataset.isnull().values.any() no missing data
print('dataset size: %d\n' % dataset.size)
print('Overview of data \n', dataset.head())
print('Class types:\n', dataset['Class'].value_counts())
# we don't need sample code number
x = dataset.iloc[:, 1:-1].values
y = dataset.iloc[:, -1].values
```

dataset size: 7513

Overview of data

	Sample code number	Clump Thickness	...	Mitoses	Class
0	1000025	5	...	1	2
1	1002945	5	...	1	2
2	1015425	3	...	1	2
3	1016277	6	...	1	2
4	1017023	4	...	1	2

[5 rows x 11 columns]

Class types:

2 444

4 239

Name: Class, dtype: int64

We are going to assume that class 2 is normal and class 4 is cancerous

▼ Split into train and test sets

```
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=0)
```

▼ Feature scaling

```
# although the variance of features in x is similar so it may not be required
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x_test = sc.fit_transform(x_test)
```

▼ Model creation

```
# Logistic regression
from sklearn.linear_model import LogisticRegression
lr = LogisticRegression().fit(x_train, y_train)
# KNN
from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors = 5, metric = 'minkowski', p = 2).fit(x_train, y_train)
# SVM
from sklearn.svm import SVC
svc = SVC(kernel='rbf').fit(x_train, y_train)
# Naive Bayes
from sklearn.naive_bayes import GaussianNB
nb = GaussianNB().fit(x_train, y_train)
# Random forest
from sklearn.ensemble import RandomForestClassifier
rf = RandomForestClassifier(criterion='entropy').fit(x_train, y_train)
```

▼ Confusion matrix and accuracy

```
from sklearn.metrics import confusion_matrix, accuracy_score
# logistic regression
print("Logistic Regression performance")
y_pred_lr = lr.predict(x_test)
print(confusion_matrix(y_test, y_pred_lr))
print(accuracy_score(y_test, y_pred_lr))

# knn
print("KNN performance")
y_pred_knn = knn.predict(x_test)
print(confusion_matrix(y_test, y_pred_knn))
print(accuracy_score(y_test, y_pred_knn))

# SVM
print("SVM performance")
y_pred_svc = svc.predict(x_test)
print(confusion_matrix(y_test, y_pred_svc))
```

```
print(accuracy_score(y_test, y_pred_svc))
```

```
# Naive Bayes
```

```
print("Naive Bayes performance")
```

```
y_pred_nb = nb.predict(x_test)
```

```
print(confusion_matrix(y_test, y_pred_nb))
```

```
print(accuracy_score(y_test, y_pred_nb))
```

```
# Random Forest
```

```
print("Random Forest performance")
```

```
y_pred_rf = rf.predict(x_test)
```

```
print(confusion_matrix(y_test, y_pred_rf))
```

```
print(accuracy_score(y_test, y_pred_rf))
```

```
Logistic Regression performance
```

```
[[84  3]
```

```
 [ 3 47]]
```

```
0.9562043795620438
```

```
KNN performance
```

```
[[82  5]
```

```
 [ 2 48]]
```

```
0.948905109489051
```

```
SVM performance
```

```
[[82  5]
```

```
 [ 1 49]]
```

```
0.9562043795620438
```

```
Naive Bayes performance
```

```
[[80  7]
```

```
 [ 1 49]]
```

```
0.9416058394160584
```

```
Random Forest performance
```

```
[[84  3]
```

```
 [ 1 49]]
```

```
0.9708029197080292
```

