

# PAD project

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## Topic

AI can assist in various situations. We can implement simple statistical analysis and achieve great results.

## Heart Disease

- Cleveland
- Hungarian
- Switzerland
- Long Beach VA
- Statlog (Heart) Data Set.

This heart disease dataset is curated by combining five popular heart disease datasets that were previously available independently but not combined before.

## Collecting and finding data

[Heart Disease Dataset](#)

The dataset is sourced from Kaggle.

## Downloading dataset

```
import kaggle

kaggle.api.authenticate()
kaggle.api.dataset_download_files(
    'mexwell/heart-disease-dataset',
    path='data',
    unzip=True
)
```

Dataset URL: <https://www.kaggle.com/datasets/mexwell/heart-disease-dataset>

The dataset is automatically downloaded to the local data folder. Ensure you set your Kaggle username and authentication key.

## Cleaning data and handling missing values

Let's explore the dataset for the most obvious errors.

### Visualize incorrect rows

```
coloms_to_remove = data.loc[
    (data['resting bp s']<10) | (data['ST slope']==0)
]
coloms_to_remove[['resting bp s', 'ST slope']]
```

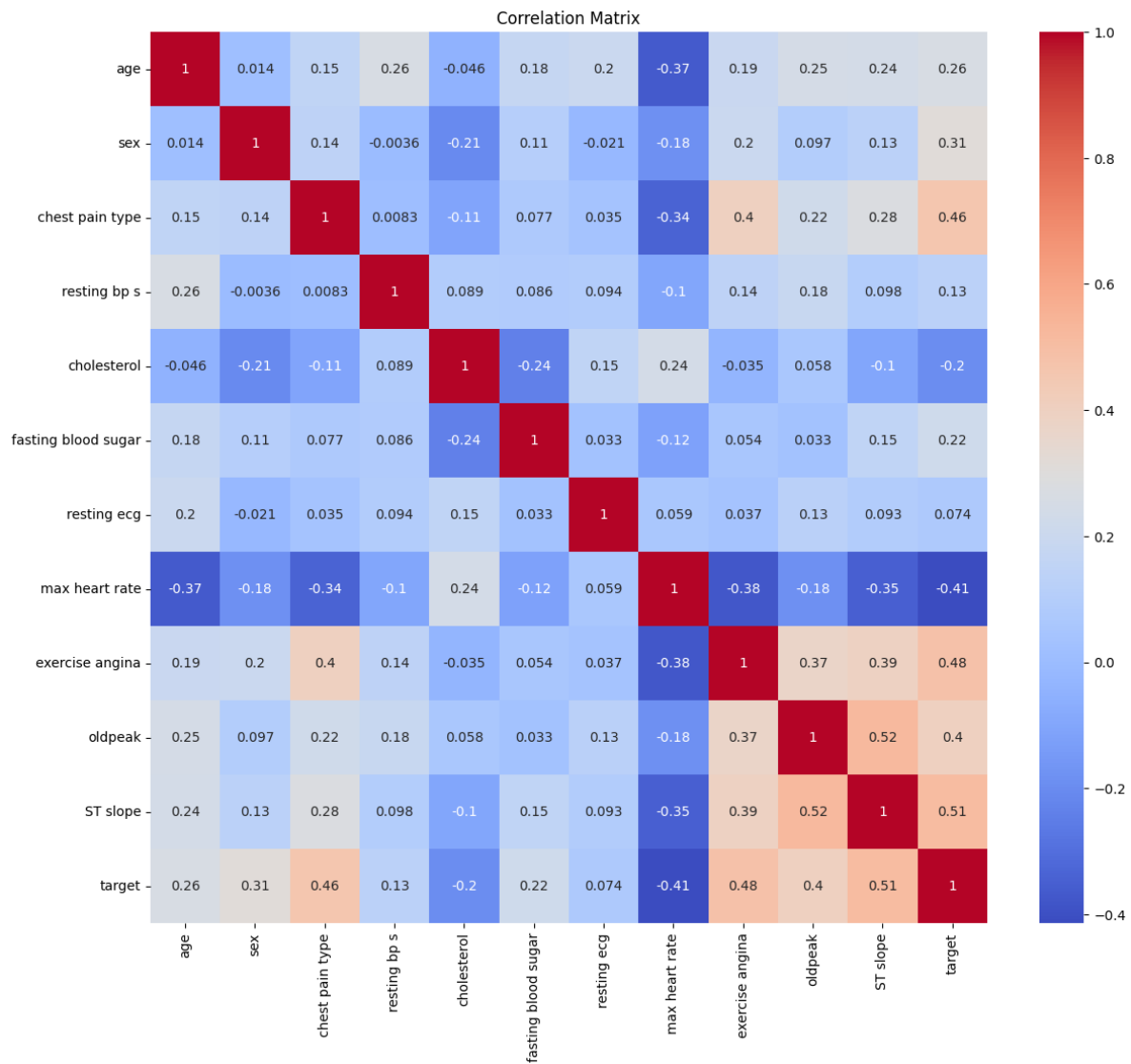
	resting bp s	ST slope
450	0	2
517	150	0

Two rows are clearly impossible for human beings. The first has a heart rate of zero beats per second, and the second has a slope of the peak exercise ST segment of zero, which is not possible according to the documentation.

## Data analysis – attribute dependencies

This section focuses on attribute dependencies.

## Heatmap



The main attributes describing variance in the data are chest pain type, exercise-induced angina, and the slope of the peak exercise ST segment.

## Creating a dashboard

```
python3 dashboard.py
```

A dashboard is created using Plotly, showing the distribution of variables.

## **\*Evaluation**

Problem type: classification

Let's move to the evaluation part.

### **Split data**

```
from sklearn.model_selection import train_test_split

X = data.drop('target', axis=1)
y = data['target']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

The data was split into training and testing datasets.

### **Data preparation**

```
from sklearn.preprocessing import StandardScaler
from sklearn.impute import SimpleImputer

imputer = SimpleImputer(strategy='mean')
X_train = imputer.fit_transform(X_train)
X_test = imputer.transform(X_test)
scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
```

The cleaned data was prepared for evaluation.

### **Training data**

```

from sklearn.metrics import accuracy_score, classification_report
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.neighbors import KNeighborsClassifier
from sklearn.ensemble import GradientBoostingClassifier
from sklearn.naive_bayes import GaussianNB

models = {
    'Logistic Regression': LogisticRegression(max_iter=10000),
    'Decision Tree': DecisionTreeClassifier(),
    'Random Forest': RandomForestClassifier(),
    'SVM': SVC(),
    'KNN': KNeighborsClassifier(),
    'Gradient Boosting': GradientBoostingClassifier(),
    'Naive Bayes': GaussianNB(),
}

for name, model in models.items():
    print(f"Training {name}...")
    model.fit(X_train, y_train)
    y_pred = model.predict(X_test)
    accuracy = accuracy_score(y_test, y_pred)
    print(f"{name} Accuracy: {accuracy:.2f}")
    print(classification_report(y_test, y_pred))

```

Training Logistic Regression...

Logistic Regression Accuracy: 0.85

	precision	recall	f1-score	support
0	0.84	0.84	0.84	112
1	0.86	0.86	0.86	126
accuracy			0.85	238
macro avg	0.85	0.85	0.85	238
weighted avg	0.85	0.85	0.85	238

Training Decision Tree...

Decision Tree Accuracy: 0.87

	precision	recall	f1-score	support
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0	0.83	0.90	0.87	112
1	0.91	0.84	0.87	126
accuracy			0.87	238
macro avg	0.87	0.87	0.87	238
weighted avg	0.87	0.87	0.87	238

Training Random Forest...

Random Forest Accuracy: 0.93

	precision	recall	f1-score	support
0	0.91	0.94	0.93	112
1	0.94	0.92	0.93	126
accuracy			0.93	238
macro avg	0.93	0.93	0.93	238
weighted avg	0.93	0.93	0.93	238

Training SVM...

SVM Accuracy: 0.87

	precision	recall	f1-score	support
0	0.88	0.83	0.85	112
1	0.86	0.90	0.88	126
accuracy			0.87	238
macro avg	0.87	0.86	0.86	238
weighted avg	0.87	0.87	0.87	238

Training KNN...

KNN Accuracy: 0.86

	precision	recall	f1-score	support
0	0.84	0.87	0.85	112
1	0.88	0.85	0.86	126
accuracy			0.86	238
macro avg	0.86	0.86	0.86	238
weighted avg	0.86	0.86	0.86	238

Training Gradient Boosting...

Gradient Boosting Accuracy: 0.91

	precision	recall	f1-score	support
0	0.91	0.90	0.91	112
1	0.91	0.92	0.92	126
accuracy			0.91	238
macro avg	0.91	0.91	0.91	238
weighted avg	0.91	0.91	0.91	238

Training Naive Bayes...

Naive Bayes Accuracy: 0.88

	precision	recall	f1-score	support
0	0.85	0.89	0.87	112
1	0.90	0.87	0.88	126
accuracy			0.88	238
macro avg	0.88	0.88	0.88	238
weighted avg	0.88	0.88	0.88	238

The data was trained on various models.

## Stat analysis

Training Logistic Regression...

Logistic Regression Accuracy: 0.85

	precision	recall	f1-score	support
0	0.84	0.84	0.84	112
1	0.86	0.86	0.86	126
accuracy			0.85	238
macro avg	0.85	0.85	0.85	238
weighted avg	0.85	0.85	0.85	238

Training Decision Tree...

Decision Tree Accuracy: 0.87

	precision	recall	f1-score	support
0	0.83	0.90	0.87	112
1	0.91	0.84	0.87	126

accuracy			0.87	238
macro avg	0.87	0.87	0.87	238
weighted avg	0.87	0.87	0.87	238

Training Random Forest...

Random Forest Accuracy: 0.93

	precision	recall	f1-score	support
0	0.91	0.94	0.93	112
1	0.94	0.92	0.93	126

accuracy			0.93	238
macro avg	0.93	0.93	0.93	238
weighted avg	0.93	0.93	0.93	238

Training SVM...

SVM Accuracy: 0.87

	precision	recall	f1-score	support
0	0.88	0.83	0.85	112
1	0.86	0.90	0.88	126

accuracy			0.87	238
macro avg	0.87	0.86	0.86	238
weighted avg	0.87	0.87	0.87	238

Training KNN...

KNN Accuracy: 0.86

	precision	recall	f1-score	support
0	0.84	0.87	0.85	112
1	0.88	0.85	0.86	126

accuracy			0.86	238
macro avg	0.86	0.86	0.86	238
weighted avg	0.86	0.86	0.86	238

Training Gradient Boosting...

Gradient Boosting Accuracy: 0.91

	precision	recall	f1-score	support
0	0.91	0.90	0.91	112
1	0.91	0.92	0.92	126



accuracy			0.91	238
macro avg	0.91	0.91	0.91	238
weighted avg	0.91	0.91	0.91	238

Training Naive Bayes...

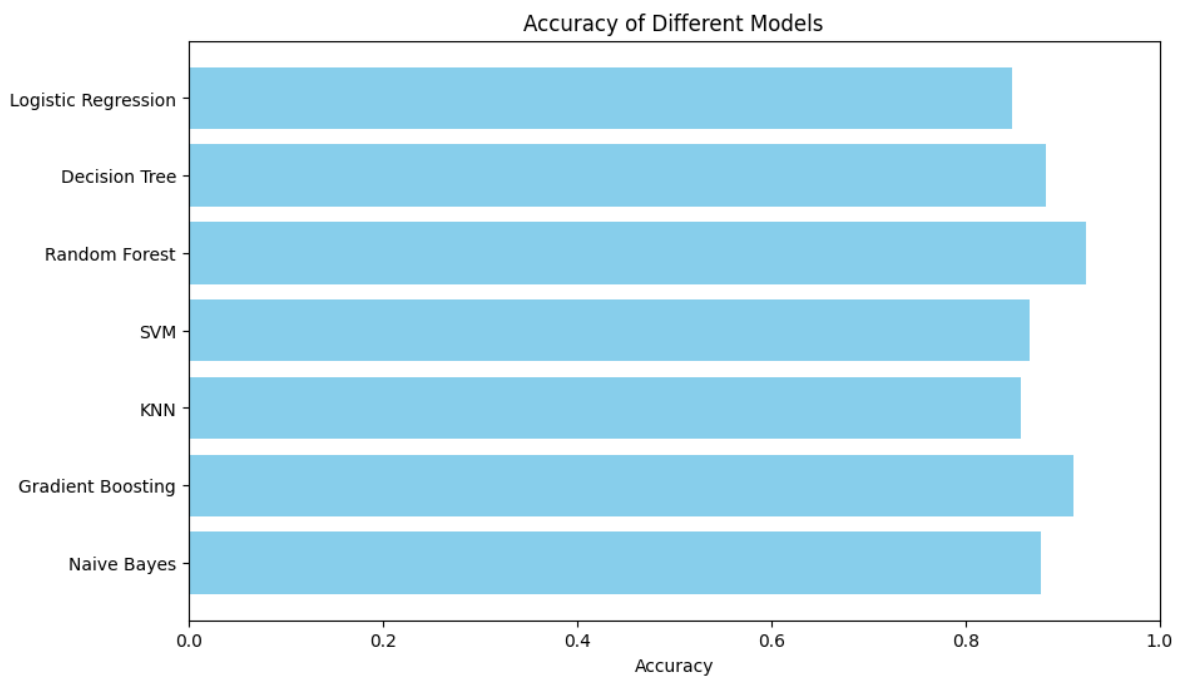
Naive Bayes Accuracy: 0.88

	precision	recall	f1-score	support
0	0.85	0.89	0.87	112
1	0.90	0.87	0.88	126

accuracy			0.88	238
macro avg	0.88	0.88	0.88	238
weighted avg	0.88	0.88	0.88	238

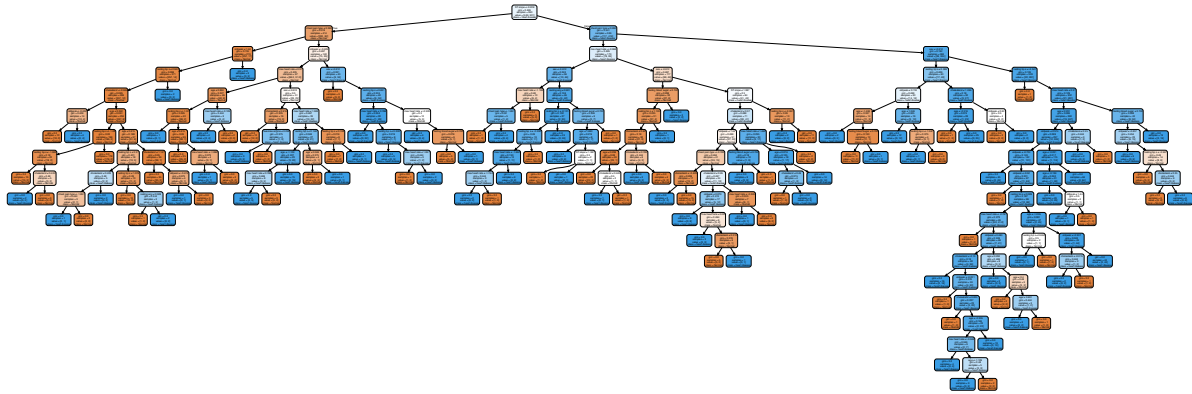
Stat analysis is done with use of precision, recall, f1-score

## Accuracy Comparison



The Random Forest algorithm shows the best results.

## Building a tree



Here is the decision tree itself.