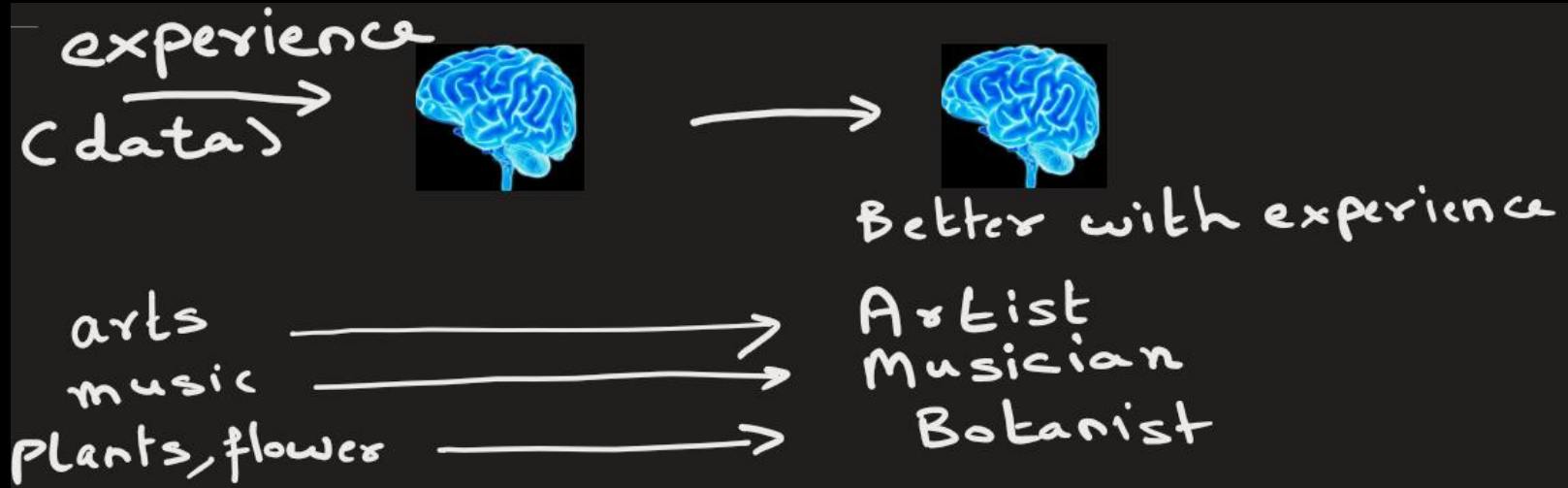


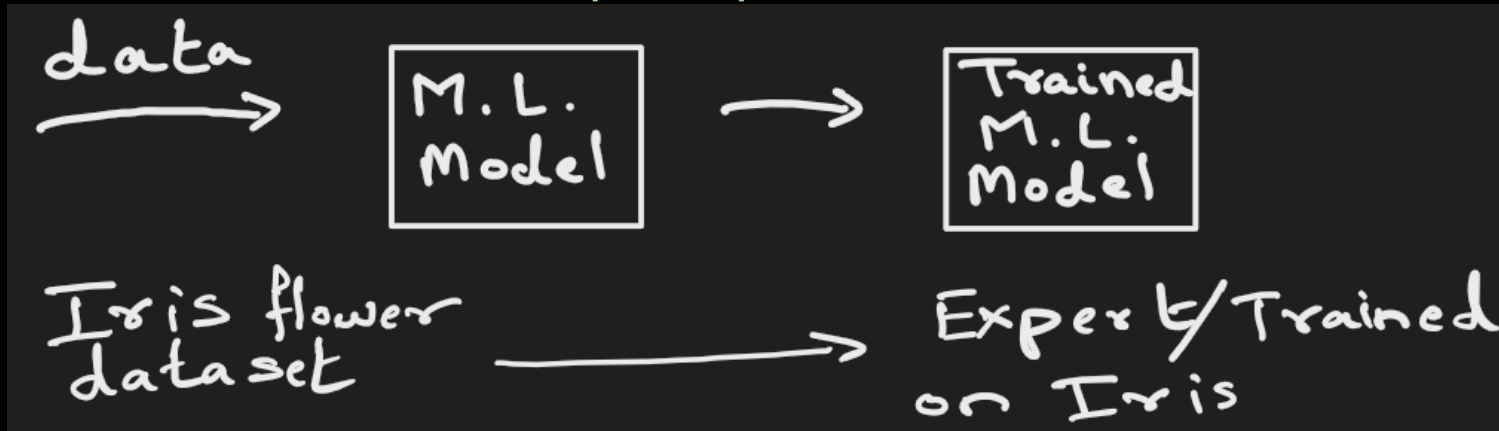
Machine Learning – General Idea

Human brain learns by experience.

This experience comes in the form of data. The brain becomes better with more data.



ML Model works on the same principle.





Task: You are given data of Iris flower. Make a ML model that would become expert on Iris flower.

The ML model should be able to recognize the type of flower (Setosa, Versicolor or Virginica) based on input: sepal and petal dimensions

| sepal_length | sepal_width | petal_length | petal_width | species |
|--------------|-------------|--------------|-------------|------------|
| 6.2 | 2.8 | 4.8 | 1.8 | virginica |
| 7.4 | 2.8 | 6.1 | 1.9 | virginica |
| 6.2 | 2.2 | 4.5 | 1.5 | versicolor |
| 6.3 | 2.9 | 5.6 | 1.8 | virginica |
| 5.5 | 2.6 | 4.4 | 1.2 | versicolor |
| 5.7 | 2.5 | 5.0 | 2.0 | virginica |
| 5.1 | 3.3 | 1.7 | 0.5 | setosa |
| 6.8 | 2.8 | 4.8 | 1.4 | versicolor |
| 5.2 | 3.5 | 1.5 | 0.2 | setosa |
| 5.4 | 3.9 | 1.7 | 0.4 | setosa |



Iris Versicolor



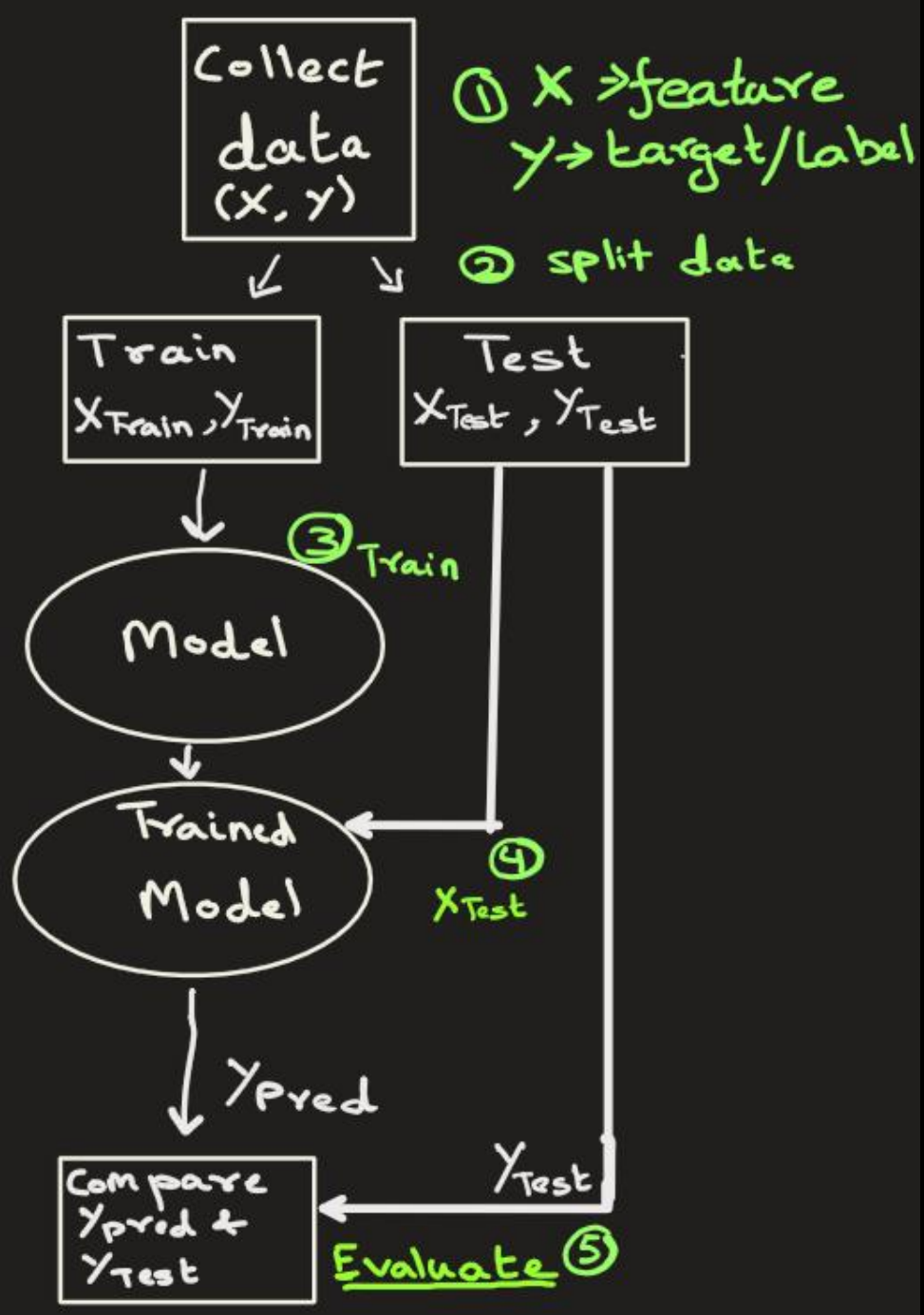
Iris Setosa

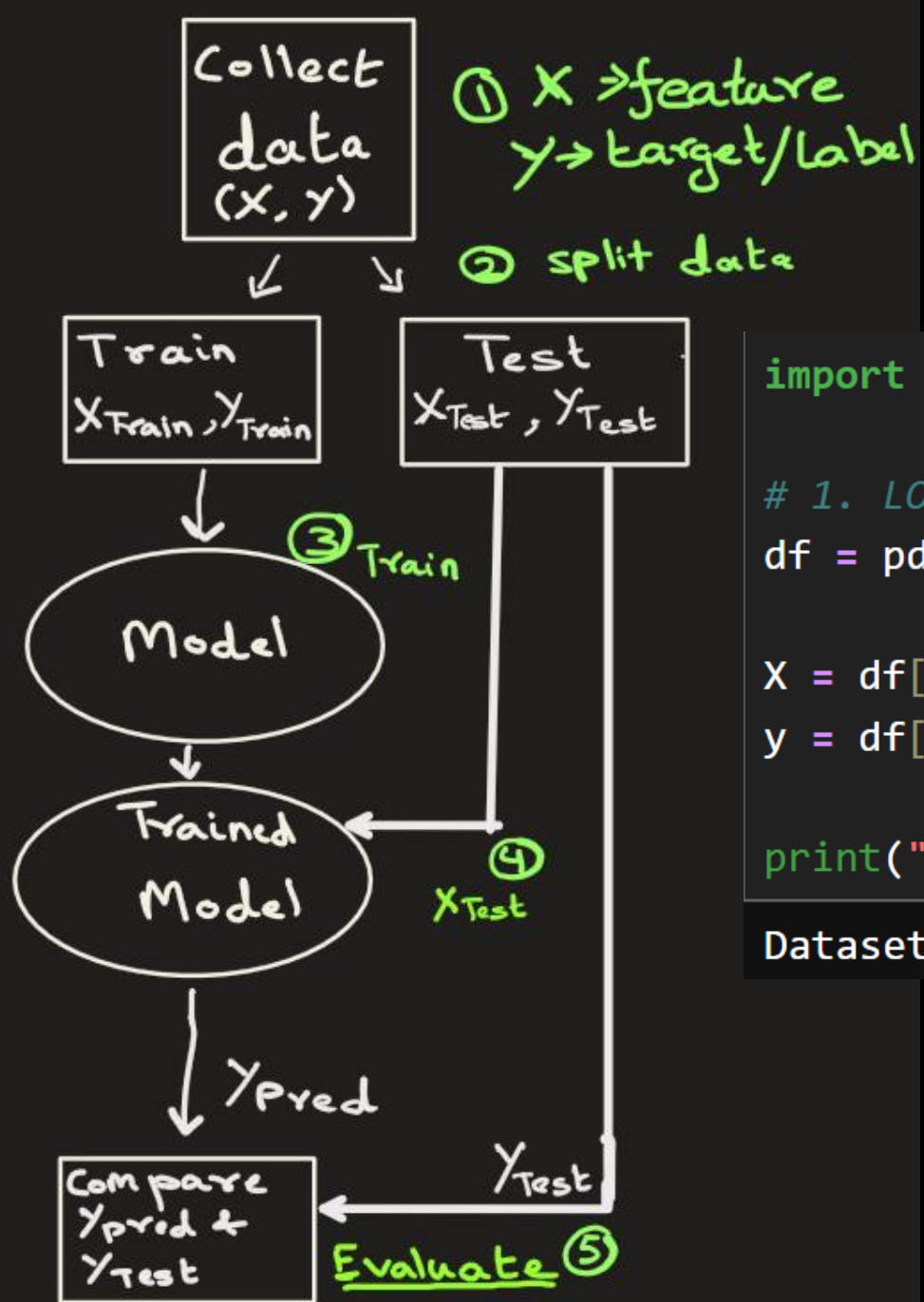


Iris Virginica



General Idea of ML





```
import pandas as pd
```

```
# 1. LOAD DATASET
```

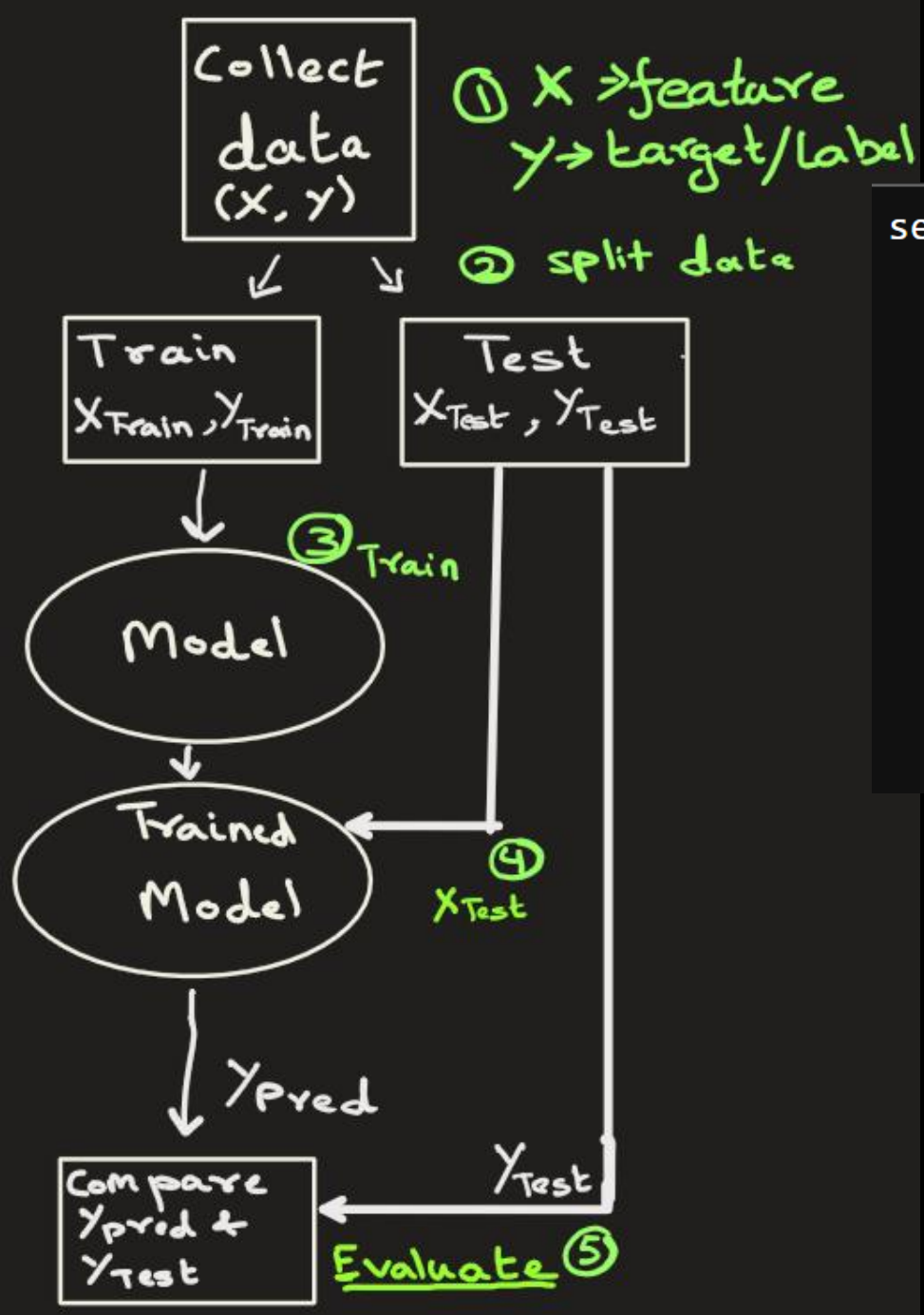
```
df = pd.read_csv("iris.csv")
```

```
X = df[["sepal_length", "sepal_width", "petal_length", "petal_width"]]
```

```
y = df['species']
```

```
print("Dataset shape:", df.shape)
```

```
Dataset shape: (150, 5)
```

| sepal_length | sepal_width | petal_length | petal_width | species |
|--------------|-------------|--------------|-------------|------------|
| 6.2 | 2.8 | 4.8 | 1.8 | virginica |
| 7.4 | 2.8 | 6.1 | 1.9 | virginica |
| 6.2 | 2.2 | 4.5 | 1.5 | versicolor |
| 6.3 | 2.9 | 5.6 | 1.8 | virginica |
| 5.5 | 2.6 | 4.4 | 1.2 | versicolor |
| 5.7 | 2.5 | 5.0 | 2.0 | virginica |
| 5.1 | 3.3 | 1.7 | 0.5 | setosa |
| 6.8 | 2.8 | 4.8 | 1.4 | versicolor |
| 5.2 | 3.5 | 1.5 | 0.2 | setosa |
| 5.4 | 3.9 | 1.7 | 0.4 | setosa |

Following is a sample taken from Iris flower dataset.

- This dataset has total of 150 records
- It contains records for 3 types of flowers: Setosa, Versicolor and Virginica
- Each type of flower has 50 records.



Collect
data
(X, y)

① $X \rightarrow \text{feature}$
 $y \rightarrow \text{target/label}$

② split data

Train
 $X_{\text{Train}}, y_{\text{Train}}$

Test
 $X_{\text{Test}}, y_{\text{Test}}$

③ Train

Model

Trained
Model

④

X_{Test}

y_{Pred}

Compare
 y_{pred} &
 y_{Test}

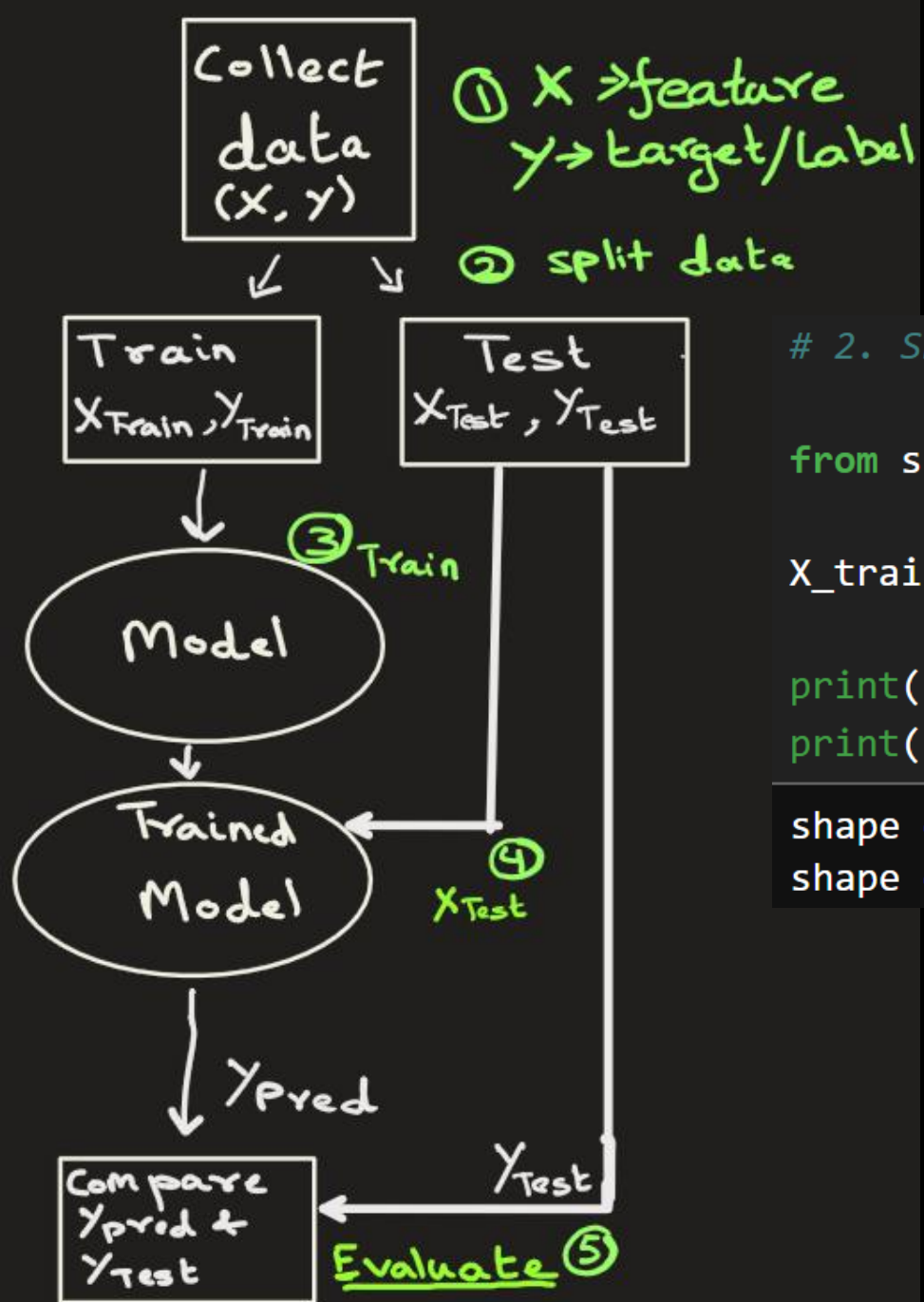
Evaluate ⑤

| X | | | | y |
|--------------|-------------|--------------|-------------|------------|
| sepal_length | sepal_width | petal_length | petal_width | species |
| 6.2 | 2.8 | 4.8 | 1.8 | virginica |
| 7.4 | 2.8 | 6.1 | 1.9 | virginica |
| 6.2 | 2.2 | 4.5 | 1.5 | versicolor |
| 6.3 | 2.9 | 5.6 | 1.8 | virginica |
| 5.5 | 2.6 | 4.4 | 1.2 | versicolor |
| 5.7 | 2.5 | 5.0 | 2.0 | virginica |
| 5.1 | 3.3 | 1.7 | 0.5 | setosa |
| 6.8 | 2.8 | 4.8 | 1.4 | versicolor |
| 5.2 | 3.5 | 1.5 | 0.2 | setosa |
| 5.4 | 3.9 | 1.7 | 0.4 | setosa |

X is called **features** (independent)

y is called **target/response/label** (dependent)

y is function of X.



2. SPLIT DATA INTO TRAINING AND TESTING

```
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)
```

```
print("shape of X_train:", X_train.shape)
```

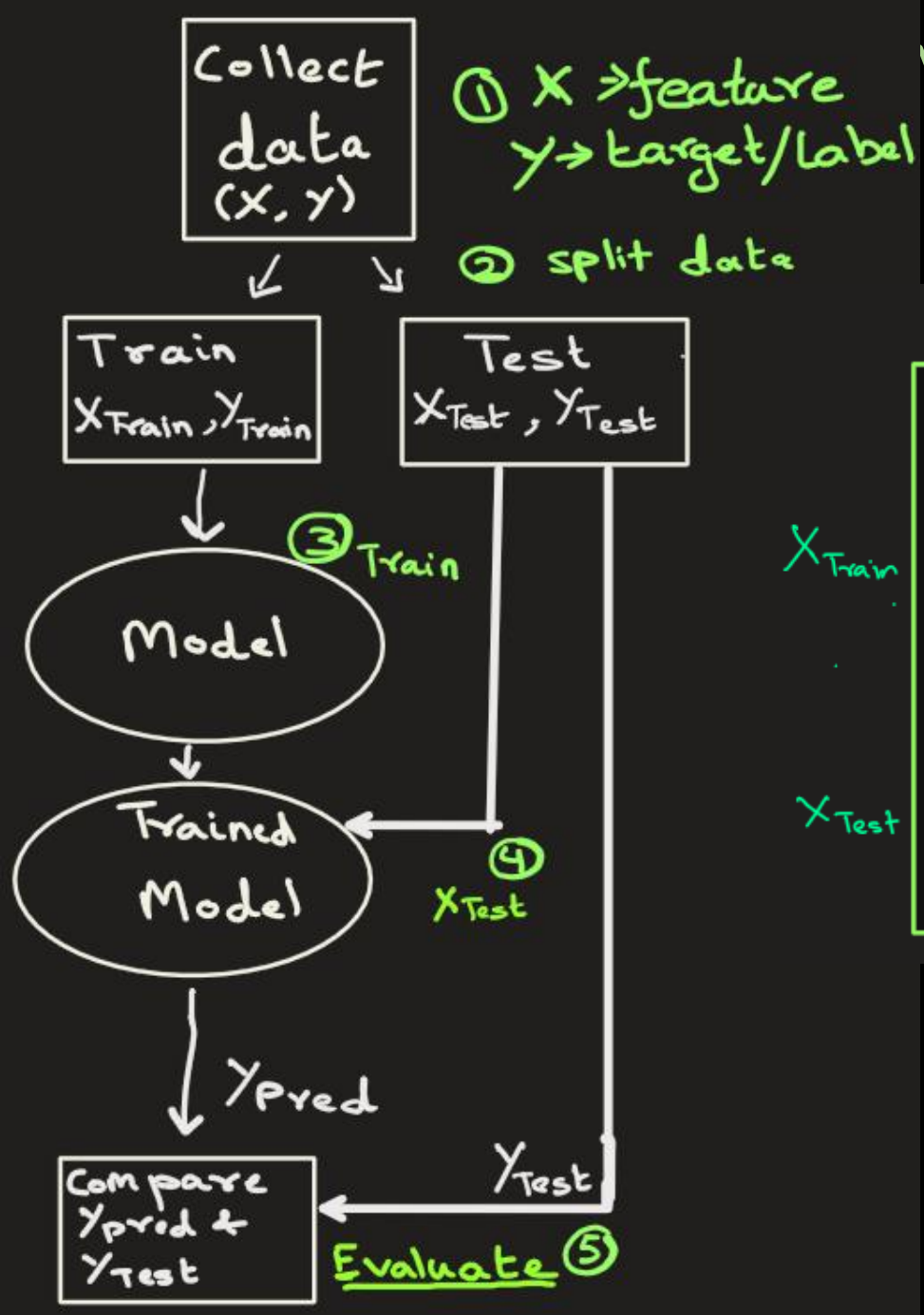
```
print("shape of X_test :", X_test.shape)
```

```
shape of X_train: (120, 4)
```

```
shape of X_test : (30, 4)
```



ML Overview



| X | | | | y |
|--------------|-------------|--------------|-------------|------------|
| sepal_length | sepal_width | petal_length | petal_width | species |
| 6.2 | 2.8 | 4.8 | 1.8 | virginica |
| 7.4 | 2.8 | 6.1 | 1.9 | virginica |
| 6.2 | 2.2 | 4.5 | 1.5 | versicolor |
| 6.3 | 2.9 | 5.6 | 1.8 | virginica |
| 5.5 | 2.6 | 4.4 | 1.2 | versicolor |
| 5.7 | 2.5 | 5.0 | 2.0 | virginica |
| 5.1 | 3.3 | 1.7 | 0.5 | setosa |
| 6.8 | 2.8 | 4.8 | 1.4 | versicolor |
| 5.2 | 3.5 | 1.5 | 0.2 | setosa |
| 5.4 | 3.9 | 1.7 | 0.4 | setosa |

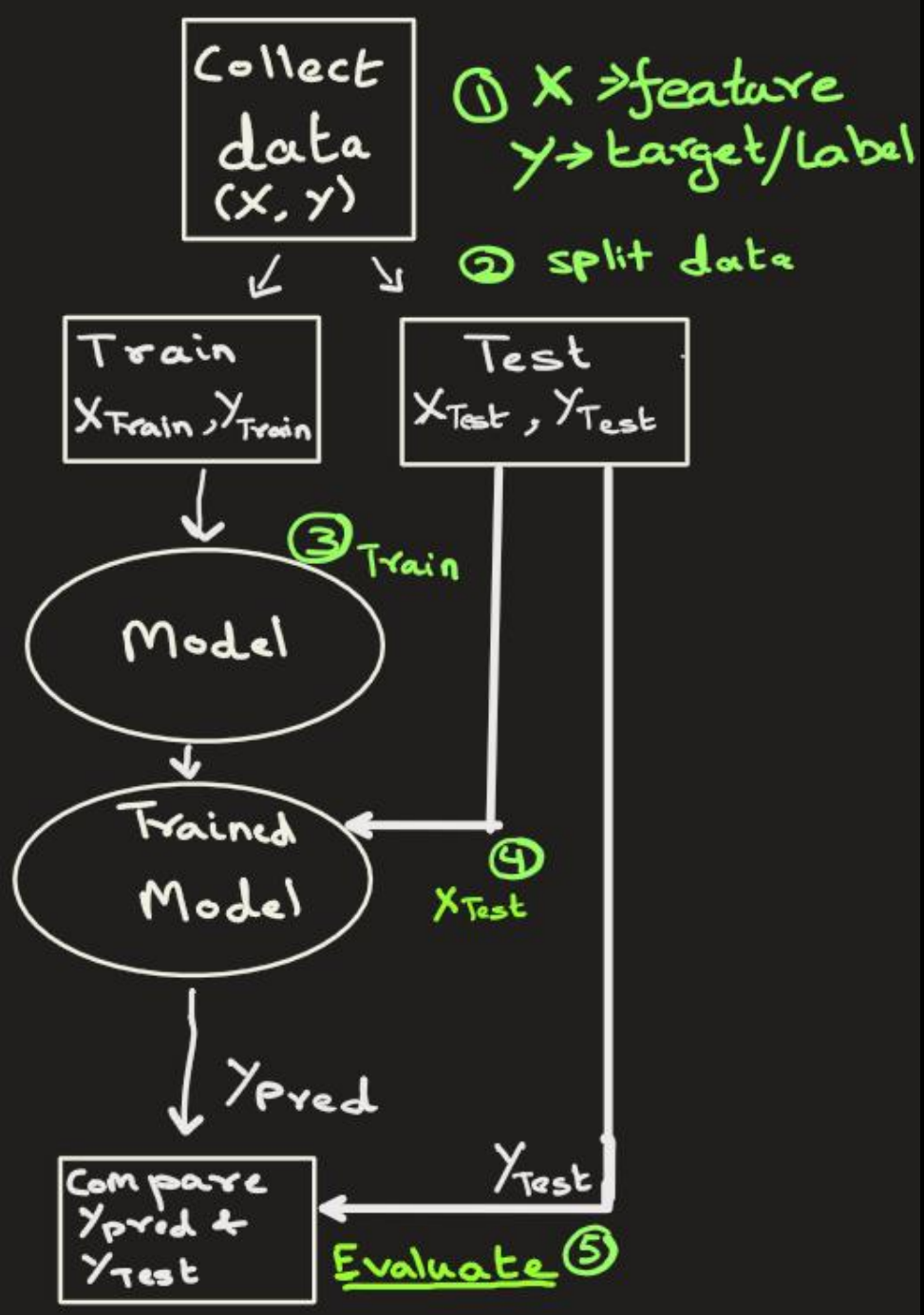
X_{Train}

X_{Test}

y_{Train}

y_{Test}

X and y are split into training and testing part.
There would be total 4 parts now.



3. TRAIN THE MODEL

```
from sklearn.ensemble import RandomForestClassifier
```

```
model = RandomForestClassifier()
```

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

▼ RandomForestClassifier ⓘ ?

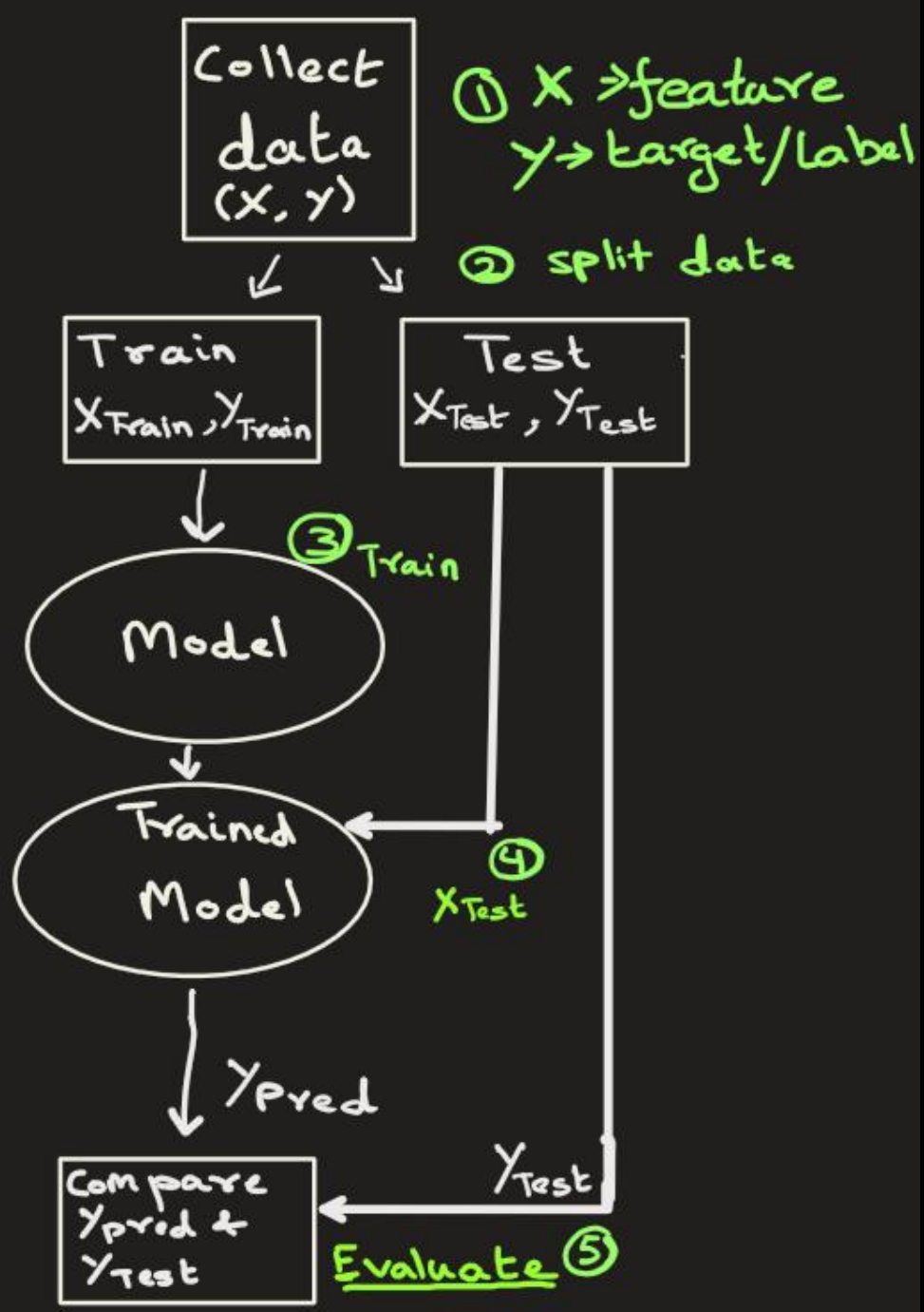
RandomForestClassifier()

Training data is fed to the ML Model.
It may take some time for model to learn/train.

| X | | | | Y | |
|---|--|--|--|------------|--|
| sepal_length sepal_width petal_length petal_width | | | | species | |
| 6.2 2.8 4.8 1.8 | | | | virginica | |
| 7.4 2.8 6.1 1.9 | | | | virginica | |
| 6.2 2.2 4.5 1.5 | | | | versicolor | |
| 6.3 2.9 5.6 1.8 | | | | virginica | |
| 5.5 2.6 4.4 1.2 | | | | versicolor | |
| 5.7 2.5 5.0 2.0 | | | | virginica | |
| 5.1 3.3 1.7 0.5 | | | | setosa | |
| 6.8 2.8 4.8 1.4 | | | | versicolor | |
| 5.2 3.5 1.5 0.2 | | | | setosa | |
| 5.4 3.9 1.7 0.4 | | | | setosa | |

X_{Train} Y_{Train} Training Data

X_{Test} Y_{Test} Testing Data



```
# Lets make prediction on 2 samples that I copied
#      sepal length, sepal width, petal length, petal width
X_sample = [[5.1, 3.5, 1.4, 0.2], # actual data is setosa
            [6.3, 3.3, 6 , 2.5], # actual data is virginica
            ]
```

```
X_df = pd.DataFrame(sample, columns=X_train.columns)
```

```
prediction = model.predict(sample_df)
print("Prediction:", prediction)
```

```
Prediction: ['setosa' 'virginica']
```



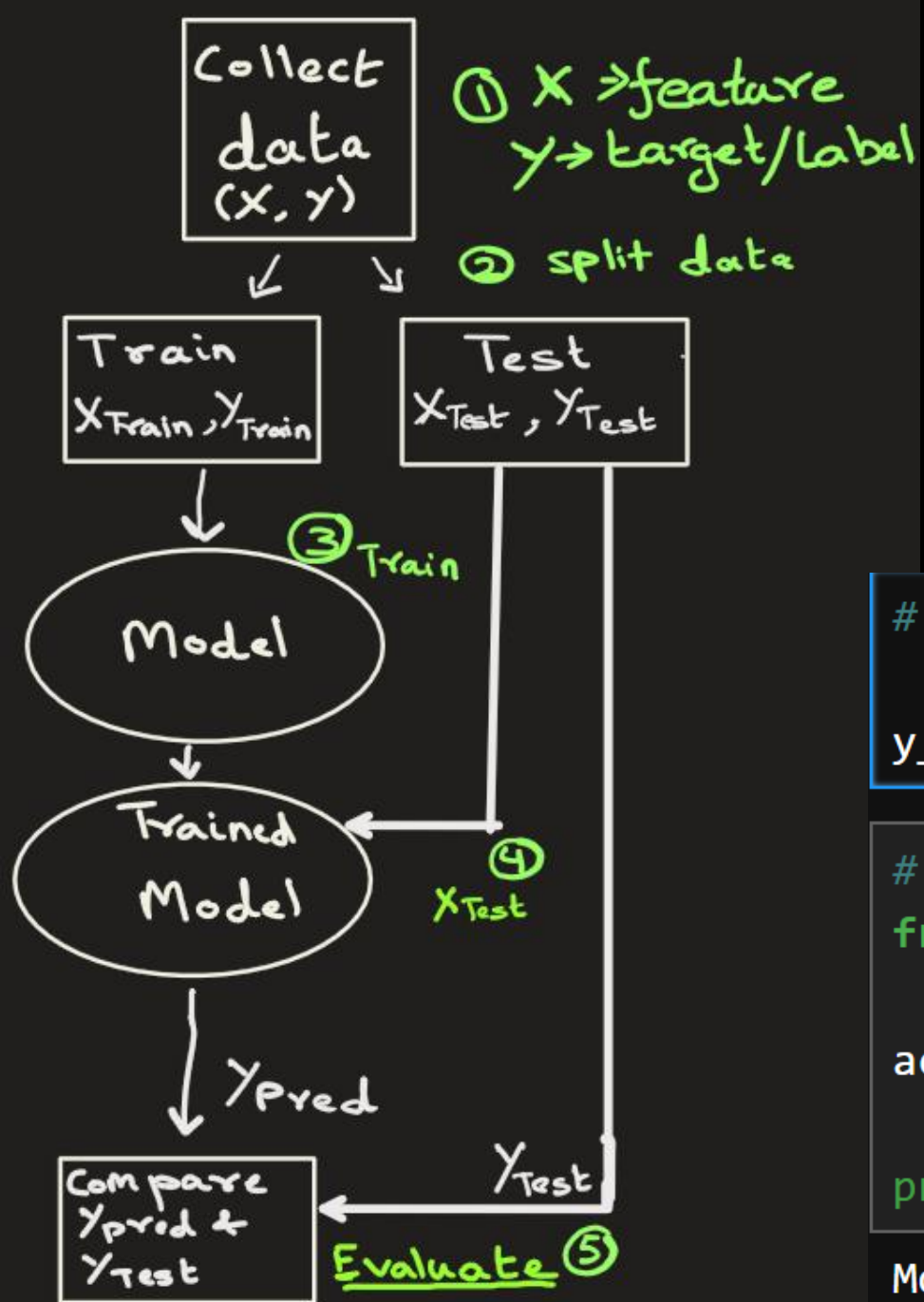
Iris Versicolor



Iris Setosa



Iris Virginica



4. FEED THE TEST DATA X_{test}

```
y_pred = model.predict(X_test)
```

5. EVALUATE THE MODEL

```
from sklearn.metrics import accuracy_score
```

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

```
print("Model Accuracy:", accuracy)
```

Model Accuracy: 0.9333333333333333



ML Overview

