



Step 7 – Decision Tree

1. Train a Decision Tree classifier for the two classes.
 2. Visualize the tree and the **decision boundary**.
 3. Evaluate performance metrics: accuracy, precision, recall, and F1-score.
-

When to Use Decision Tree

-  Use it when:
 - You need a **model that is easy to interpret** and visualize.
 - The relationships between variables are **non-linear**.
 - You want a **fast baseline** or an estimator to use inside an ensemble (Random Forest, Gradient Boosting).
 -  Avoid it when:
 - The model shows **overfitting** on training data — use pruning or limit depth.
 - You need **smooth or continuous** decision boundaries.
 - Data is **highly noisy** or unstable across samples.
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Model Hyperparameters

- `max_depth = None` — expands until all leaves are pure
 - `criterion = "gini"` — impurity measure for node splits
 - `random_state = 42` — controls randomness of splits
-

```
%run 00-setup.py
```

```
from tasks.tree import run_tree
from ml.data import load_dataset
from sklearn.model_selection import train_test_split

from ml.viz import plt_dboundary, plt_cmatrix, plt_dtree, export_tree_text
```

```
X, y, _ = load_dataset("../data/data_bivariate_gaussian.npz")

X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42, stratify=y
)
```

```
res = run_tree(X_train, y_train, X_test, y_test,

               params={"max_depth": None,
                       "criterion": "gini",
```

```

        "seed": 42}
    )

```

```

res["test"]["metrics"]

```

```

{'accuracy': 0.95,
 'precision': 0.9655172413793104,
 'recall': 0.9333333333333333,
 'f1': 0.9491525423728814}

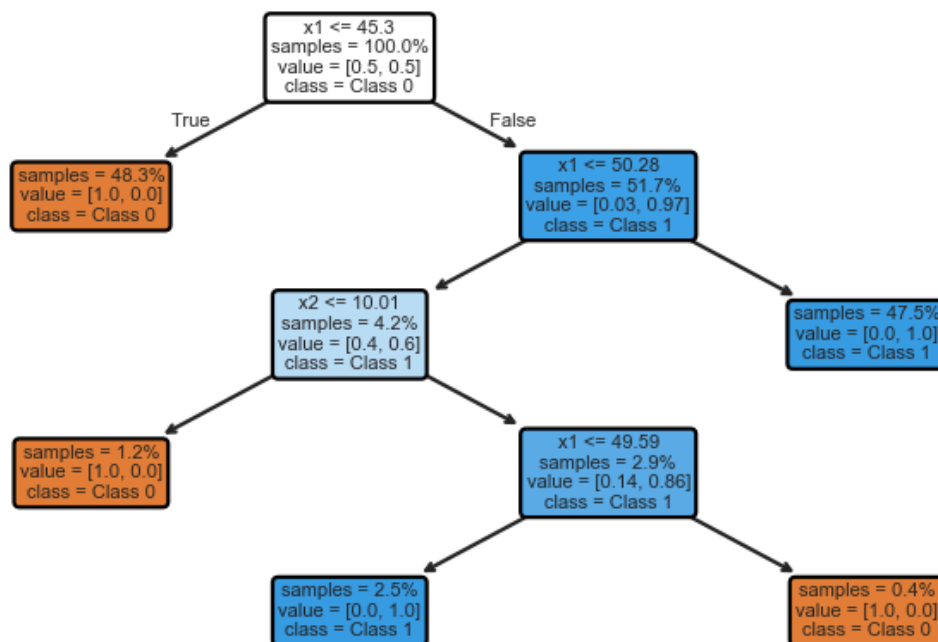
```

```

plt_dtrees(
    res["model"],
    feature_names=("x1", "x2"),
    class_names=("Class 0", "Class 1"),
    max_depth=None,
)

```

• Decision Tree



(<Figure size 800x480 with 1 Axes>, <Axes: title={'left': '• Decision Tree'}>)

```

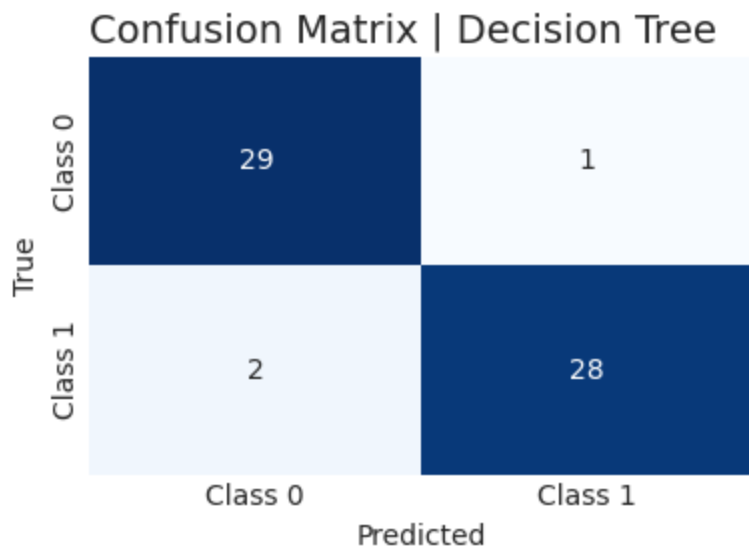
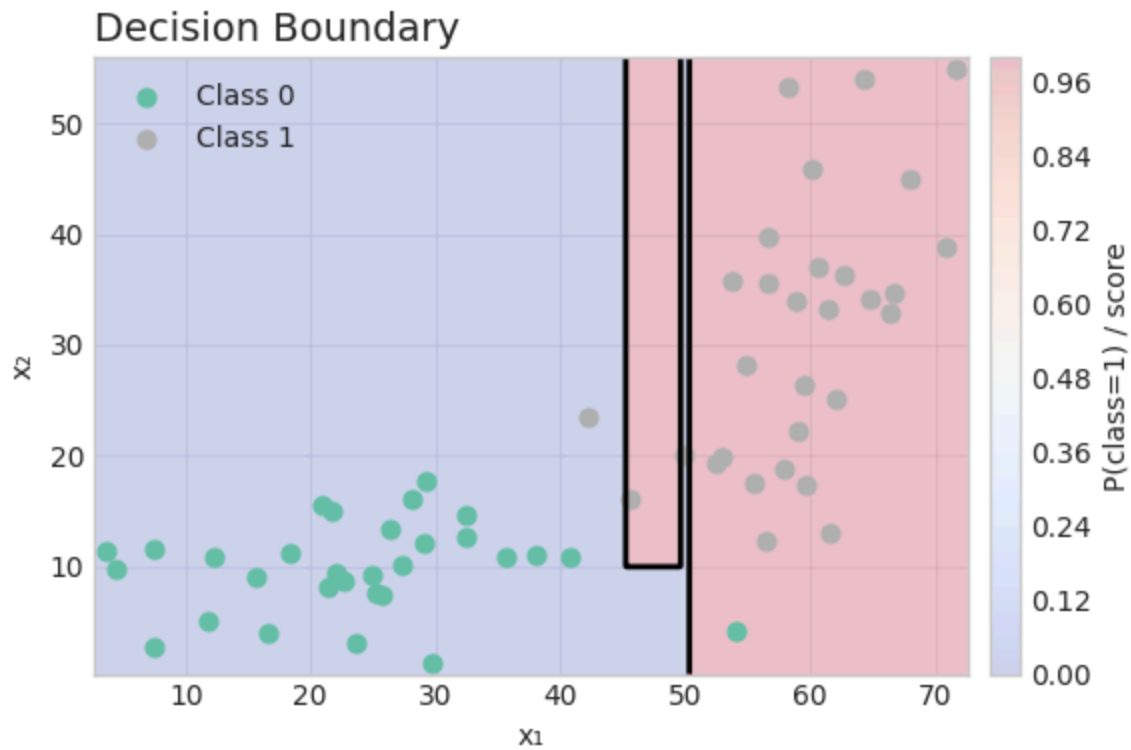
plt_dboundary(res["model"], X_test, y_test)

```

```

plt_cmatrix(
    y_true=y_test,
    y_pred=res["test"]["y_pred"],
    title="Confusion Matrix | Decision Tree"
)

```



```
(<Figure size 400x300 with 1 Axes>,  
<Axes: title={'left': 'Confusion Matrix | Decision Tree'}, xlabel='Predicted', ylabel='True'  
>)
```