

+ Code + Text

RAM Disk Editing

Problem Set 4

Team Lab 1/2:

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# Use this data

```
penguins = pd.read_csv("penguins_classification.csv")
culmen_columns = ["Culmen Length (mm)", "Culmen Depth (mm)"]
target_column = "Species"

from sklearn.model_selection import train_test_split

data, target = penguins[culmen_columns], penguins[target_column]
data_train, data_test, target_train, target_test = train_test_split(
    data, target, random_state=0
)
```

Create a decision tree classifier with a maximum depth of 2 levels and fit the training data. Once this classifier is trained, plot the data and the decision boundary to see the benefit of increasing the depth. To plot the decision tree.

✓ [14] # Write your code here

Did we make use of the feature "Culmen Length"? Plot the tree using the function `sklearn.tree.plot_tree` to find out!

✓ [15] # Write your code here

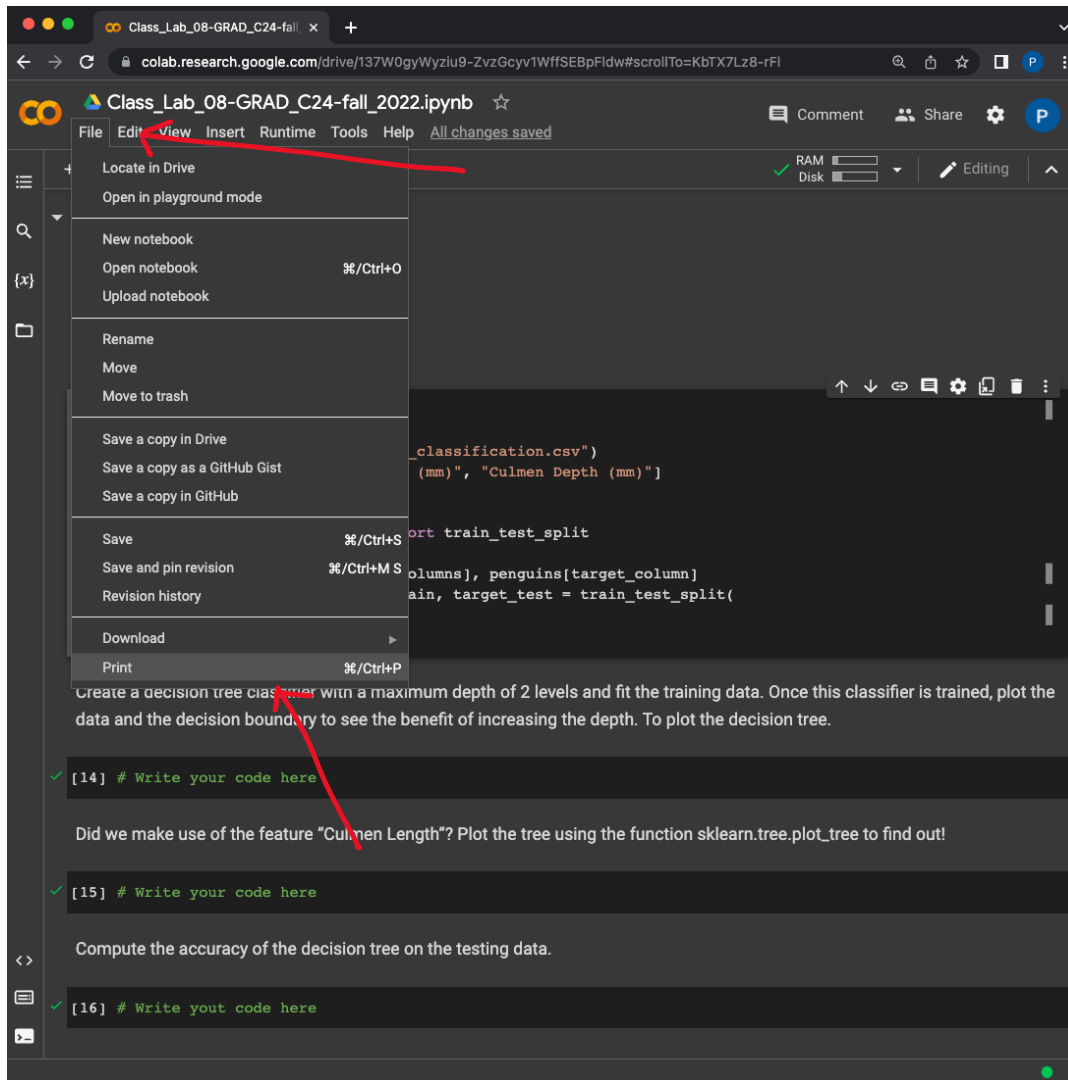
Compute the accuracy of the decision tree on the testing data.

✓ [16] # Write your code here

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Lab 08 - GRADC-24

Date: 11.11.2022

Lab Instructor: Paulina Garcia Corral

Decision Trees for classification and regression tasks, and random forests

Non-parametric models

A given split node in a decision tree classifier makes a binary decision considering a single feature at a time.

Sequence of simple decision rules: one feature and one threshold at a time.

No scaling required for numerical features, because it makes splits one feature at a time.

Well suited for tabular data with heterogeneous values.

Decision trees can be unstable because small variations in the data might result in a completely different tree being generated. This problem is mitigated by using decision trees within an ensemble.

Decision Trees are most useful as a building block for ensemble models

- Random Forests
- Gradient Boosting

Typically for controlling the underfitting/overfitting trade-off we can control for maximum depth of the decision tree and maximum leaf nodes.:

Classification Decision Tree

```
import numpy as np
import pandas as pd

penguins = pd.read_csv('penguin_classification.csv')
culmen_columns = ["Culmen Length (mm)", "Culmen Depth (mm)"]
target_column = "Species"

Split the data into subsets to see how trees predict values based on out-of-sample dataset.

from sklearn.model_selection import train_test_split
```

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[14] # Write your code here

Did we make use of the feature "Culmen Length"? Plot the tree using the function `sklearn.tree.plot_tree` to find out!

[15] # Write your code here

Compute the accuracy of the decision tree on the testing data.

[16] # Write your code here

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