











Ungraded Lab: Decision Trees

In this notebook you will visualize how a decision tree is splitted using information gain.

We will revisit the dataset used in the video lectures. The dataset is:

	Ear Shape	Face Shape	Whiskers	Category
	Pointy	Round	Present	1
	Floppy	Not Round	Present	1
	Floppy	Round	Absent	0
	Pointy	Not Round	Present	0
	Pointy	Round	Present	1
	Pointy	Round	Absent	1
	Floppy	Not Round	Absent	0
	Pointy	Round	Absent	1
	Floppy	Round	Absent	0
	Floppy	Round	Absent	0

For simplicity, we will consider:

- Ear Shape: Pointy = 1, Floppy = 0
- Face Shape: Round = 1, Not Round = 0
- Whiskers: Present = 1, Absent = 0

Therefore, we have two sets:

- `X_train` : for each example, contains 3 features:
 - Ear Shape (1 if pointy, 0 otherwise)
 - Face Shape (1 if round, 0 otherwise)
 - Whiskers (1 if present, 0 otherwise)
- `y_train` : whether the animal is a cat
 - 1 if the animal is a cat
 - 0 otherwise

```
In [ ]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from utils import *
```

```
In [ ]: X_train = np.array([[1, 1, 1],
[0, 0, 1],
[0, 1, 0],
[1, 0, 1],
[1, 1, 1],
[1, 1, 0],
[0, 0, 0],
[1, 1, 0],
[0, 1, 0],
[0, 0, 0]])
```