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SEEM2460 Lab Assignment 1

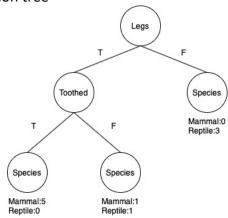
## Question 1

a) 
$$Gini_{split"Toothed"} = \frac{8}{10} \left[ 1 - \left( \frac{5}{8} \right)^2 - \left( \frac{3}{8} \right)^2 \right] + \frac{2}{10} \left[ 1 - \left( \frac{1}{2} \right)^2 - \left( \frac{1}{2} \right)^2 \right] \approx 0.475$$

$$Gini_{split"Legs"} = \frac{7}{10} \left[ 1 - \left( \frac{6}{7} \right)^2 - \left( \frac{1}{7} \right)^2 \right] + \frac{3}{10} \left[ 1 - \left( \frac{3}{3} \right)^2 - \left( \frac{0}{3} \right)^2 \right] \approx 0.1714$$

Since  $Gini_{split"Toothed"} > Gini_{split"Legs"}$ , splitting using "Legs" has lower impurity in the children nodes. Thus, "Legs" has been chosen as the first splitting attribute

b) The following is the two-level decision tree



## c) The following is the result in 'classifier output' window

=== Run information ===

Scheme: weka.classifiers.trees.J48 -C 0.25 -M 2

Relation: data-weka. filters. unsupervised. attribute. Discretize-F-B10-M-1.0-R first-last-precision 6-weka. filters. unsupervised. attribute. Discretize-B10-M-1.0-R first-last-last-precision 6-weka. filters. unsupervised. attribute. Discretize-B10-M-1.0-R first-last-last-precision 6-weka. filters. unsupervised. attribute. Discretize-B10-M-1.0-R first-last-last-precision 6-weka. filters. unsupervised. attribute. Discretize-B10-M-1.0-R first-last-precision 6-weka. Filters. Unsupervised. Attribute. Discretize-B10-M-1.0-R first-precision 6-weka. Filters. Unsupervised. The filters filters. Unsupervised. The filters filters filters filters filters. Unsupervised. The filters filter

precision6 Instances: 10

Attributes: 3
Toothed

Legs

Species

Test mode: evaluate on training data

=== Classifier model (full training set) ===

J48 pruned tree

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Legs = T: Mammal (7.0/1.0)

Legs = F: Reptile (3.0)

Number of Leaves: 2

Size of the tree: 3

Time taken to build model: 0.01 seconds

```
=== Evaluation on training set ===
Time taken to test model on training data: 0 seconds
=== Summary ===
                                 9
Correctly Classified Instances
                                           90
                                                 %
Incorrectly Classified Instances
                                 1
                                           10
                                                 %
Kappa statistic
                            0.7826
Mean absolute error
                              0.1714
Root mean squared error
                                0.2928
Relative absolute error
                              35.468 %
Root relative squared error
                                59.7269 %
Total Number of Instances
                                 10
=== Detailed Accuracy By Class ===
         TP Rate FP Rate Precision Recall F-Measure MCC
                                                               ROC Area PRC
Area Class
         1.000 0.250 0.857
                                 1.000 0.923
                                                0.802
                                                        0.875
                                                                0.857
Mammal
         0.750 0.000 1.000
                                 0.750 0.857
                                                 0.802
                                                                        Reptile
                                                        0.875
                                                                0.850
Weighted Avg. 0.900 0.150 0.914
                                                                       0.854
                                       0.900 0.897
                                                       0.802 0.875
=== Confusion Matrix ===
a b <-- classified as
60 \mid a = Mammal
1 3 \mid b = Reptile
```

## Question 2

The following is the source code

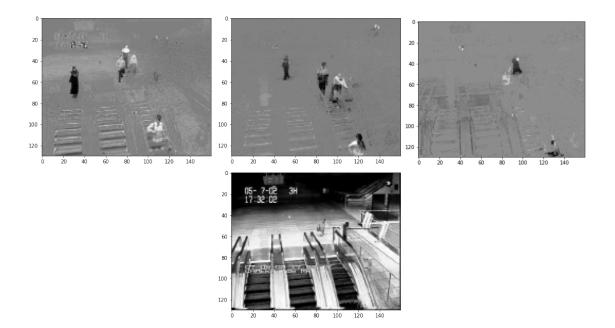
The above coding return  $[-0.000000 \ 0.000000 \ -2.000000 \ 1.000000 \ 3.000000 \ 1.000000]$ Thus, we obtain x = (0, 0, -2, 1, 3, 1)'

## Question 3

a) The following is the source code

```
import numpy as np
import cv2
import cvxpy as cp
from cvxpy import *
import matplotlib.pyplot as plt
im1 = cv2.imread('/content/Figure1.png',cv2.IMREAD GRAYSCALE)
im2 = cv2.imread('/content/Figure2.png',cv2.IMREAD GRAYSCALE)
im3 = cv2.imread('/content/Figure3.png',cv2.IMREAD GRAYSCALE)
M \text{ size} = im1.shape
size a = M \text{ size}[0]
size b = M \text{ size}[1]
n = size a*size b
M1 = im1.reshape(n, -1)
M2 = im2.reshape(n, -1)
M3 = im3.reshape(n, -1)
w = cp.Variable((n, 1))
# Please trying to implementing you code here:
obj = 0
for i in range(3):
 obj += norm((M1 - w), 1) + norm((M2 - w), 1) + norm((M3 - w), 1)
prob = cp.Problem(cp.Minimize(obj))
prob.solve()
# End of modification here
plt.figure(figsize=(6,6))
plt.imshow((M1 - w.value).reshape(size a, size b), cmap='gray')
plt.figure(figsize=(6,6))
plt.imshow((M2 - w.value).reshape(size a, size b), cmap='gray')
plt.figure(figsize=(6,6))
plt.imshow((M3 - w.value).reshape(size a, size b), cmap='gray')
plt.figure(figsize=(6,6))
plt.imshow((w.value).reshape(size a, size b), cmap='gray')
```

The above code return the following graphs



b) Since I am out off relatively static figures, I grab some figures from Github Source: <a href="https://github.com/sayibet/fight-detection-surv-dataset">https://github.com/sayibet/fight-detection-surv-dataset</a>

The following are figures that being inputted into the programme



The following are the output



We can see that there is still some afterglow on the final output, the extracted background. The result is not perfect at all, but it is still acceptable to see how the original background looks like.

My source code for you reference

https://colab.research.google.com/drive/1PIu2pvpGYR8VnzCtS6vmt626AmRPkwun