

Part 1

1.

```
[ ] 1 print("Gini of data is ", gini(data))
```

Gini of data is 0.4628099173553719

```
▶ 1 print("Entropy of data is ", entropy(data))
```

↗ Entropy of data is 0.9456603046006401

2-1.

```
▶ 1 clf_depth3 = DecisionTree('gini', 3,  
2                               x_train, y_train  
3                               , x_test, y_test)  
4 clf_depth10 = DecisionTree('gini', 10,  
5                               x_train, y_train  
6                               , x_test, y_test)  
7 print(clf_depth3.accuracy, clf_depth10.accuracy)
```

↗ 0.78 0.69

Depth 10 suffers from overfitting.

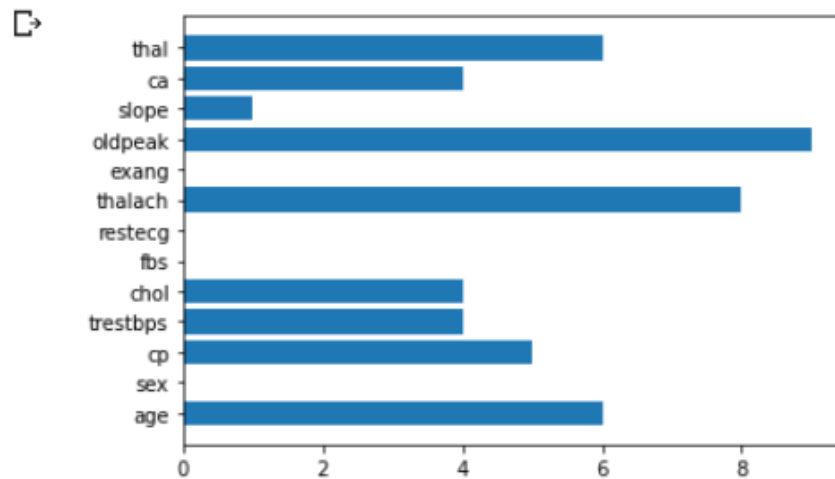
2-2.

```
[ ] 1 clf_gini = DecisionTree('gini', 3,  
2                               x_train, y_train,  
3                               x_test, y_test)  
4 clf_entropy = DecisionTree('entropy', 3,  
5                               x_train, y_train,  
6                               x_test, y_test)  
7 print(clf_gini.accuracy, clf_entropy.accuracy)
```

0.78 0.76

3.

```
1 plt.barh(range(len(feature_used)),
2           feature_used,
3           tick_label=list(dictionary.keys()))
4 plt.show()
```



4.

```
[ ] 1 es_ten = AdaBoost(10)
     2 es_hun = AdaBoost(100)
```

▼ Question 4.1

Show the accuracy score of test data by `n_estimators=10` and `n_estimators=100`, respectively.

```
[ ] 1 print(es_ten.accuracy, es_hun.accuracy)
```

0.78 0.78

5-1.

```
[22] 1 clf_10tree = RandomForest(10, np.sqrt(x_train.shape[1]), True, 'gini',
2                                None, x_train, y_train, x_test, y_test)
     3 clf_100tree = RandomForest(100, np.sqrt(x_train.shape[1]), True, 'gini',
4                                None, x_train, y_train, x_test, y_test)
```

```
1 print(clf_10tree.accuracy, clf_100tree.accuracy)
```

0.8 0.74

5-2.

```
✓ [26] 1 clf_random_features = RandomForest(10, np.sqrt(x_train.shape[1]), True, 'gini',  
4      2 None, x_train, y_train, x_test, y_test)  
秒      3 clf_all_features = RandomForest(10, x_train.shape[1], True, 'gini',  
        4 None, x_train, y_train, x_test, y_test)
```

- Note: Use majority votes to get the final prediction, you may get slightly different results when re-building the random forest model

```
✓ 0 1 print(clf_random_features.accuracy, clf_all_features.accuracy)  
秒  
0.82 0.77
```

6.

```
✓ 1 1 best_result = RandomForest(20, np.sqrt(x_train.shape[1]), True, 'entropy',  
秒 2 3, x_train, y_train, x_test, y_test)  
3 print(best_result.accuracy)
```

```
0.86
```

Part 2

1. $A: (300, 100) \rightarrow \text{class } 1$

$(100, 300) \rightarrow \text{class } 2$

$$\text{misclassification} = \frac{100}{400} \times \frac{400}{800} + \frac{100}{400} \times \frac{400}{800} = \frac{1}{4}$$

$$\text{cross-entropy} = -\frac{1}{2} \left(\frac{1}{4} \log_2 \frac{1}{4} + \frac{3}{4} \log_2 \frac{3}{4} \right) - \frac{1}{2} \left(\frac{3}{4} \log_2 \frac{3}{4} + \frac{1}{4} \log_2 \frac{1}{4} \right) \approx 0.811$$

$$\text{Gini-index} = \frac{1}{2} \left(1 - \frac{1}{16} - \frac{9}{16} \right) + \frac{1}{2} \left(1 - \frac{1}{16} - \frac{9}{16} \right) = \frac{3}{8}$$

$B: (200, 400) \rightarrow \text{class } 2$

$(200, 0) \rightarrow \text{class } 1$

$$\text{misclassification} = \frac{200}{600} \times \frac{600}{800} + \frac{0}{200} \times \frac{200}{800} = \frac{1}{4}$$

$$\text{cross-entropy} = -\frac{3}{4} \left(\frac{1}{3} \log_2 \frac{1}{3} + \frac{2}{3} \log_2 \frac{2}{3} \right) - \frac{1}{4} (1 \cdot \log_2 1) \approx 0.689$$

$$\text{Gini-index} = \frac{3}{4} \left(1 - \frac{1}{9} - \frac{4}{9} \right) + \frac{1}{4} (1 - 1) = \frac{1}{3}$$

$0.689 < 0.811$ and $\frac{1}{3} < \frac{3}{8} \Rightarrow$ both criteria are lower #

$$2. E_{x,t}[e^{-ty(x)}] = \int_t \int e^{-ty(x)} p(t|x) p(x) dx$$

$$= \int [e^{-y(x)} p(t=1|x) + e^{y(x)} p(t=-1|x)] p(x) dx$$

$$\text{Let } e^{y(x)} = \mu, \quad p(t=1|x) = \varepsilon \Rightarrow p(t=-1|x) = 1-\varepsilon$$

To minimize the function, $\frac{\varepsilon}{\mu} + \mu(1-\varepsilon)$ should be minimal

$$\Rightarrow \frac{d\left[\frac{\varepsilon}{\mu} + \mu(1-\varepsilon)\right]}{d\mu} = -\varepsilon \cdot \mu^{-2} + (1-\varepsilon) = 0$$

$$\Rightarrow \mu = \left(\frac{\varepsilon}{1-\varepsilon}\right)^{\frac{1}{2}}$$

$$\Rightarrow y(x) = \ln\left(\frac{\varepsilon}{1-\varepsilon}\right)^{\frac{1}{2}}$$

$$= \frac{1}{2} \ln\left(\frac{\varepsilon}{1-\varepsilon}\right)$$

$$= \frac{1}{2} \ln \frac{p(t=1|x)}{p(t=-1|x)} \quad \#$$