

Question 1

$$\begin{aligned} a) \quad E(X^2) &= E(Y^2) = \text{Var}(X) + E(X)^2 \\ &= \frac{|b-a|^2}{12} + \frac{b+a}{2} \\ &= \frac{1}{12} + \frac{1}{2} = \frac{13}{12} \end{aligned}$$

$$\begin{aligned} E(Z) &= E(X^2 - 2XY + Y^2) \\ &= E(X^2) - 2E(X)E(Y) + E(Y)^2 \\ &= \frac{13}{12} - 2 + \frac{13}{12} = \frac{1}{6} \end{aligned}$$

$$\text{Var}(X^2) = \text{Var}(Y^2)$$

$$= E(X^4) - E(X^2)^2$$

By fourth moment of mgf

$$E(X^4) = \frac{9}{5}$$

$$= \frac{9}{5} - \left(\frac{13}{12}\right)^2 = \frac{451}{720}$$

$$\text{Var}(X) = \frac{(b-a)^2}{12} = \frac{1}{12}$$

$$\text{Var}(Z) = \text{Var}(X - Y^2)$$

$$= \text{Var}(X^2 - 2XY + Y^2)$$

$$= \text{Var}(X^2) + 4\text{Var}(XY) + \text{Var}(Y^2)$$

$\because X$ and Y are independent

$$\therefore \text{Var}(XY) = \text{Var}(X)\text{Var}(Y)$$

$$= \text{Var}(X^2) + 4\text{Var}(X)\text{Var}(Y) + \text{Var}(Y^2)$$

$$= \frac{451}{720} + 4 \cdot \frac{1}{12} \cdot \frac{1}{12} + \frac{451}{720}$$

$$= \frac{461}{360}$$

$$b) E[R] = E[Z_1 + \dots + Z_d]$$

\because Since $X_1, Y_1, \dots, X_d, Y_d \sim \text{Uniform}(0, 1)$

$$\therefore E(|X_1 - Y_1|) = \dots = E(|X_d - Y_d|)$$

$$\therefore E(Z_1) = \dots = E(Z_d)$$

$$= d E(Z)$$

$$= d \cdot \frac{1}{6} = \frac{d}{6}$$

$$\text{Var}[R] = \text{Var}[Z_1 + \dots + Z_d]$$

$$\because \text{Var}[Z_1] = \dots = \text{Var}[Z_d]$$

$$\therefore \text{Var}[Z_1 + \dots + Z_d] = \text{Var}(dZ_1)$$

$$= d^2 \text{Var}[Z_1]$$

$$= d^2 \frac{461}{720}$$

$$d) \quad \sigma = \sqrt{d^2 \frac{461}{720}} = d \sqrt{\frac{461}{720}}$$

$$\mu = d \cdot \frac{1}{6}$$

$$\because \frac{1}{6} < \sqrt{\frac{461}{720}} \approx 0.8$$

$$\therefore \mu < \sigma$$

As dimension d increase, σ increase faster than μ .

$$\text{maximum distance} = d E(z)$$

Therefore as d increase,

average distance increase,

so every points get further away.

But they get relative same distance, which seems to be closer.

Question 2

$$a) H(X) = \sum_x p(x) \log_2 \left(\frac{1}{p(x)} \right)$$

$$\because 0 \leq p(x) \leq 1,$$

$$\therefore \frac{1}{p(x)} \geq 1$$

$$\therefore \log_2 \left(\frac{1}{p(x)} \right) \geq \log_2(1)$$

$$\Rightarrow \log_2 \left(\frac{1}{p(x)} \right) \geq 0$$

$$b) \quad H(X, Y) = H(X) + H(Y|X)$$

by properties $H(Y|X) = H(Y)$

$$H(X, Y) = H(X) + H(Y)$$

$$\begin{aligned} c) \quad H(X, Y) &= - \sum_{x \in X} \sum_{y \in Y} p(x, y) \log_2 p(x, y) \\ &= - \sum_{x \in X} \sum_{y \in Y} p(x, y) \log_2 p(x) p(y|x) \\ &= - \sum_{x \in X} \sum_{y \in Y} p(x, y) \log_2 p(x) - \sum_{x \in X} \sum_{y \in Y} p(x, y) \log_2 p(y|x) \\ &= - \sum_{x \in X} p(x) \log_2 p(x) - \sum_{x \in X} \sum_{y \in Y} p(x, y) \log_2 p(y|x) \\ &= H(X) + H(Y|X) \end{aligned}$$

d)

$$\begin{aligned} KL(p||q) &= \int p(x) \log_2 \left(\frac{p(x)}{q(x)} \right) dx \\ &= - \int p(x) \log_2 \left(\frac{q(x)}{p(x)} \right) dx \\ &= -E \left(\log_2 \frac{q}{p} \right) \end{aligned}$$

by Jensen's inequality

$$E \left(\log_2 \frac{q}{p} \right) \geq \log_2 \left(E \left(\frac{q}{p} \right) \right)$$

$$-E \left(\log_2 \frac{q}{p} \right) \leq -\log_2 \left(E \left(\frac{q}{p} \right) \right)$$

\therefore Since

$$\therefore -\log_2 \left(\sum p(x) \frac{q(x)}{p(x)} \right) = 0$$

$$\therefore -E \left(\log_2 \frac{q}{p} \right) \leq 0$$

$$\therefore E \left(\log_2 \frac{q}{p} \right) \geq 0$$

e)

$$\begin{aligned} KL(p(x,y) || p(x)p(y)) &= \sum p(x,y) \log_2 \left(\frac{p(x,y)}{p(x)p(y)} \right) \\ &= \sum p(x,y) \log_2 \left(\frac{p(y)p(x|y)}{p(x)p(y)} \right) \\ &= \sum p(x,y) \log_2 \left(\frac{p(x|y)}{p(x)} \right) \\ &= \sum p(x,y) \log_2 \left(\frac{1}{p(x)} \right) - \sum p(x,y) \log_2 \left(\frac{1}{p(x|y)} \right) \end{aligned}$$

$$= H(X) - H(X|Y)$$

$$= I(X; Y)$$

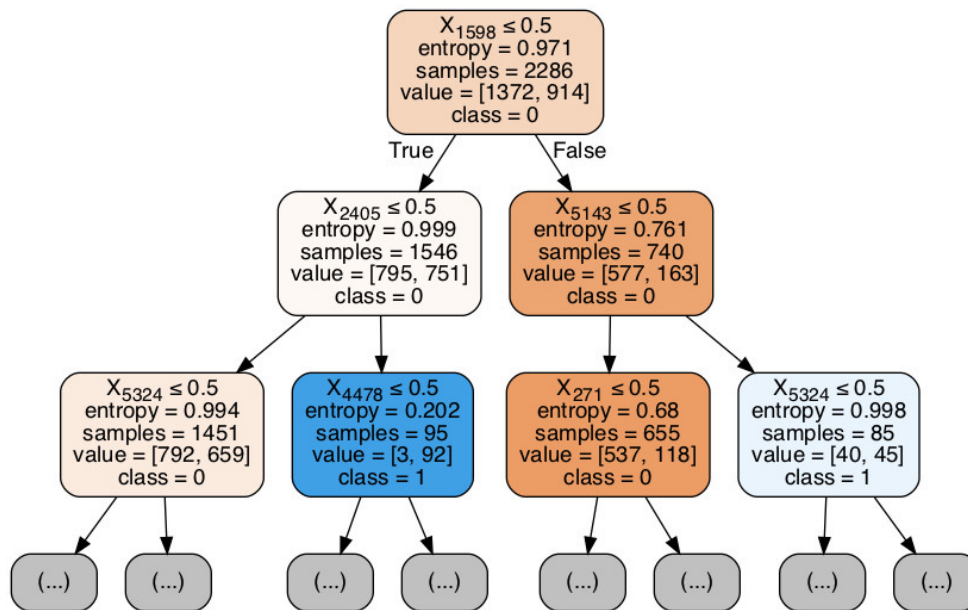
Question 3

b)

```
In [87]: models = [gini10, gini30, gini50, gini100, gini150, entro10, entro30, entro50, entro100, entro150]
scores = []
for model in models:
    score = model.score(x_validation_array, y_validation_array)
    scores.append(score)
    print(model.criterion, model.max_depth, 'Model Accuracy:', score)
best = max(scores)
best_model = models[scores.index(best)]
print('Best Model :', best_model)
print("Best Model Accuracy ", test_accuracy(best_knn_model, x_test_array, y_test_array))

gini 10 Model Accuracy: 0.7081632653061225
gini 30 Model Accuracy: 0.7653061224489796
gini 50 Model Accuracy: 0.7714285714285715
gini 100 Model Accuracy: 0.753061224489796
gini 150 Model Accuracy: 0.7714285714285715
entropy 10 Model Accuracy: 0.7204081632653061
entropy 30 Model Accuracy: 0.7612244897959184
entropy 50 Model Accuracy: 0.7673469387755102
entropy 100 Model Accuracy: 0.7775510204081633
entropy 150 Model Accuracy: 0.7673469387755102
Best Model : DecisionTreeClassifier(class_weight=None, criterion='entropy', max_depth=100,
                                     max_features=None, max_leaf_nodes=None,
                                     min_impurity_decrease=0.0, min_impurity_split=None,
                                     min_samples_leaf=1, min_samples_split=2,
                                     min_weight_fraction_leaf=0.0, presort=False,
                                     random_state=None, splitter='best')
Best Model Accuracy 0.5346938775510204
```

c)



```
In [83]: models = [gini10, gini30, gini50, gini100, gini150, entro10, entro30, entro50, entro100, entro150]
scores = []
for model in models:
    score = model.score(x_validation_array, y_validation_array)
    scores.append(score)
best = max(scores)
best_model = models[scores.index(best)]
print(best_model)
print("Best Model Accuracy ", test_accuracy(best_model, x_test_array, y_test_array))
```

```
DecisionTreeClassifier(class_weight=None, criterion='entropy', max_depth=100,
max_features=None, max_leaf_nodes=None,
min_impurity_decrease=0.0, min_impurity_split=None,
min_samples_leaf=1, min_samples_split=2,
min_weight_fraction_leaf=0.0, presort=False,
random_state=None, splitter='best')
Best Model Accuracy 0.5102040816326531
```

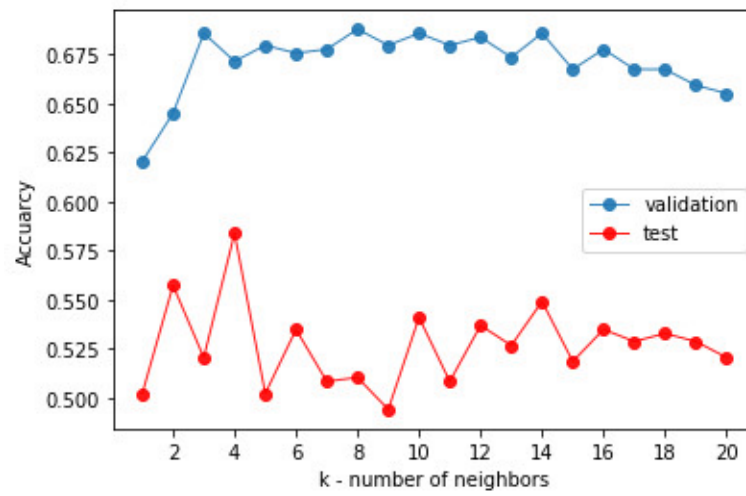

d)

```
In [213]: print(compute_info_gain(vectorizer.inverse_transform(x_train),y_train_array, 'the')) #Topmost|
          print(compute_info_gain(vectorizer.inverse_transform(x_train),y_train_array, 'hillary'))
          print(compute_info_gain(vectorizer.inverse_transform(x_train),y_train_array, 'donald'))

0.05330262393898977
0.040765700073505995
0.0501023886302121
```

```
In [ ]:
```

e)



```
In [74]: print(best_knn_model)
print("Best Model Accuracy ", test_accuracy(best_knn_model, x_test_array, y_test_array))
```

```
KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski',
metric_params=None, n_jobs=None, n_neighbors=17, p=2,
weights='uniform')
Best Model Accuracy 0.5285714285714286
```