

# Homework-2

## PART-1 Identity System Evaluation

(a) For each of the systems, plot the genuine and impostor score distributions.

```
In [1]: import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import pandas as pd
```

```
In [2]: sys1 = np.loadtxt('simMatrix1.txt', delimiter=' ')
sys2 = np.loadtxt('simMatrix2.txt', delimiter=' ')
```

```
In [3]: genuine_sys1 = sys1.diagonal()
genuine_sys2 = sys2.diagonal()
```

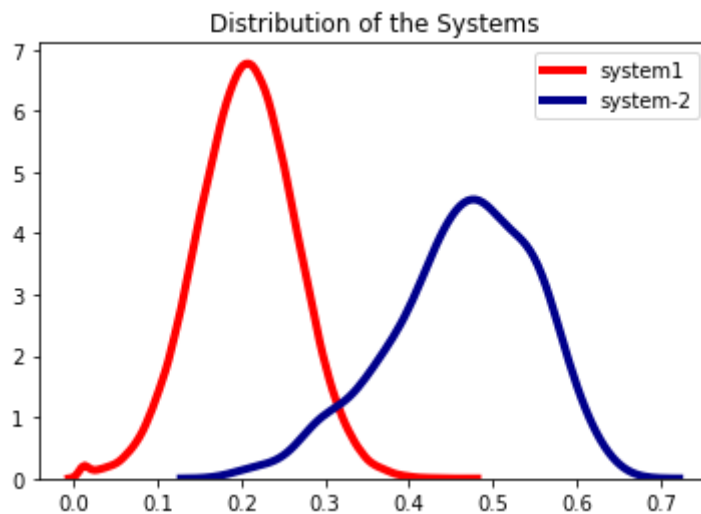
```
In [4]: non_diag = np.ones(shape=sys1.shape, dtype = bool)^np.identity(len(sys1)
).astype(bool)
imposter_sys1 = sys1[non_diag==True]
```

```
In [5]: non_diag = np.ones(shape=sys2.shape, dtype = bool)^np.identity(len(sys2)
).astype(bool)
imposter_sys2 = sys2[non_diag==True]
```

```
In [6]: # Density Plot
s1 = sns.distplot(imposter_sys1, hist=False,
                  bins=int(180/5), color = 'red',
                  hist_kws={'edgecolor': 'black'},
                  kde_kws={'linewidth': 4}, label = 'system1')

s2 = sns.distplot(genuine_sys1, hist=False,
                  bins=int(180/5), color = 'darkblue',
                  hist_kws={'edgecolor': 'black'},
                  kde_kws={'linewidth': 4}, label='system-2')
plt.title('Distribution of the Systems')
```

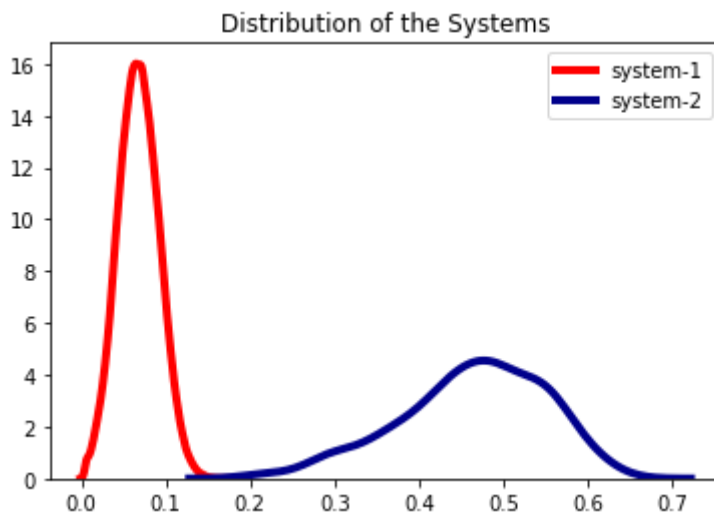
Out[6]: Text(0.5, 1.0, 'Distribution of the Systems')



```
In [7]: # Density Plot and Histogram of all arrival delays
s1 = sns.distplot(imposter_sys2, hist=False,
                  bins=int(180/5), color='red',
                  hist_kws={'edgecolor':'black'},
                  kde_kws={'linewidth': 4}, label='system-1')

s2 = sns.distplot(genuine_sys1, hist=False,
                  bins=int(180/5), color='darkblue',
                  hist_kws={'edgecolor':'black'},
                  kde_kws={'linewidth': 4}, label='system-2')
plt.title('Distribution of the Systems')
```

```
Out[7]: Text(0.5, 1.0, 'Distribution of the Systems')
```



## Cumulative Match Characteristic Curve

(b) For each of the systems, plot the Cumulative Match Characteristic curves.

(c) For each of the systems, calculate the  $d'$  (decidability index).

```
In [8]: ordered_sys1 = np.zeros((466,466))
for i in range(0,len(sys1)):
    row = sys1[i]
    sort_row = np.sort(row)[:-1]
    ordered_sys1[i] = sort_row
```

```
In [9]: ordered_sys2 = np.zeros((466,466))
for i in range(0,len(sys2)):
    row = sys2[i]
    sort_row = np.sort(row)[:-1]
    ordered_sys2[i] = sort_row
```

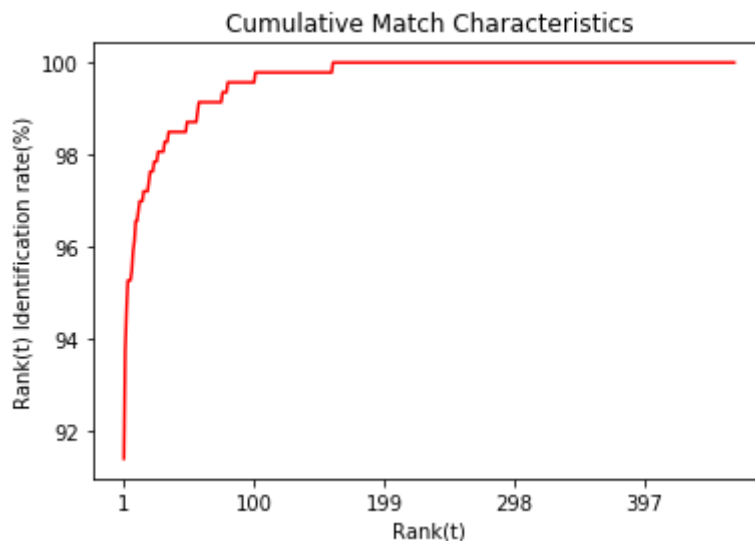
## System1

```
In [10]: rank_prob1 = []
for t in range(1, len(genuine_sys1)):
    prob = 0
    for r in range(0, len(ordered_sys1)):
        row = ordered_sys1[r]
        topt = row[0:t]
        if np.isin(genuine_sys1[r], topt):
            prob += 1
    prob = prob / 466 * 100
    rank_prob1.append(prob)
```

```
In [11]: t1 = np.arange(1, 466)
```

```
In [12]: plt.plot(t1, rank_prob1, 'r')
plt.xlabel('Rank(t)')
plt.ylabel('Rank(t) Identification rate(%)')
plt.xticks(np.arange(1, 466, step=99))
plt.title('Cumulative Match Characteristics')
```

```
Out[12]: Text(0.5, 1.0, 'Cumulative Match Characteristics')
```



## Decidability Index

```
In [13]: mean_genuine = np.mean(genuine_sys1)
mean_imposter = np.mean(imposter_sys1)
std_genuine = np.std(genuine_sys1)
std_imposter = np.std(imposter_sys1)
d1 = np.sqrt(2)*np.absolute(mean_genuine-mean_imposter)/np.sqrt(std_genu
ine**2+std_imposter**2)
print("Decidability Index =", d1)
```

```
Decidability Index = 3.516470545146112
```

(d) For each system, what is the lowest rank at which the system achieves performance greater than 70%?

System1 achieves performance greater than 70% from rank 1

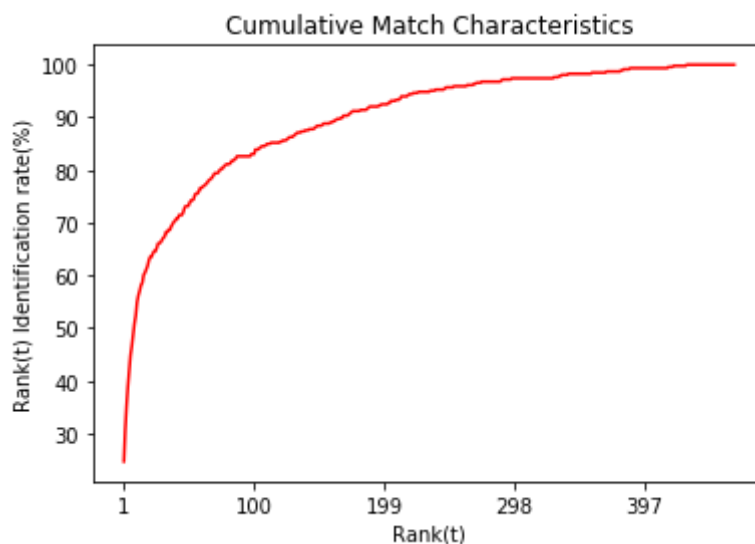
## System2

```
In [14]: rank_prob2 = []
for t in range(1, len(genuine_sys2)):
    prob = 0
    for r in range(0, len(ordered_sys2)):
        row = ordered_sys2[r]
        topt = row[0:t]
        if(np.isin(genuine_sys2[r], topt)):
            prob+=1
    prob = prob/466*100
    rank_prob2.append(prob)
```

```
In [15]: t2 = np.arange(1, 466)
```

```
In [16]: plt.plot(t2, rank_prob2, 'r')
plt.xlabel('Rank(t)')
plt.ylabel('Rank(t) Identification rate(%)')
plt.xticks(np.arange(1, 466, step=99))
plt.title('Cumulative Match Characteristics')
```

```
Out[16]: Text(0.5, 1.0, 'Cumulative Match Characteristics')
```



## Decidability Index

```
In [17]: mean_genuine = np.mean(genuine_sys2)
mean_imposter = np.mean(imposter_sys2)
std_genuine = np.std(genuine_sys2)
std_imposter = np.std(imposter_sys2)
d2 = np.sqrt(2)*np.absolute(mean_genuine-mean_imposter)/np.sqrt(std_genuine**2+std_imposter**2)
print("Decidability Index =",d2)
```

Decidability Index = 1.7116813276236182

**(d) For each system, what is the lowest rank at which the system achieves performance greater than 70%?**

**System1 achieves performance greater than 70% from rank 40**

## PART-2 Verification System Evaluation

### ROC Curve

**(e) For each of the systems, plot the Receiver Operating Curve (FAR vs. FRR)**

**(f) For each of the systems, calculate the Equal Error Rate. At what operating point is this rate achieved for each system?**

**(g) For each of the systems, determine what the FRR is when the FAR = 1%, FAR = 5%, FAR = 10%, and FAR = 20%. Present results in tabular format.**

### System1

```

In [18]: FAR=[]
          FRR=[]
          thresh = []
          t=0
          while(t<=1):
              thresh.append(t)
              far=len(imposter_sys1[imposter_sys1>t])/len(imposter_sys1)
              FAR.append(far)
              frr=len(genuine_sys1[genuine_sys1<t])/len(genuine_sys1)
              FRR.append(frr)
              t+=0.0001

          FAR=np.round(np.array(FAR)*100)
          FRR=np.round(np.array(FRR)*100)

```

```

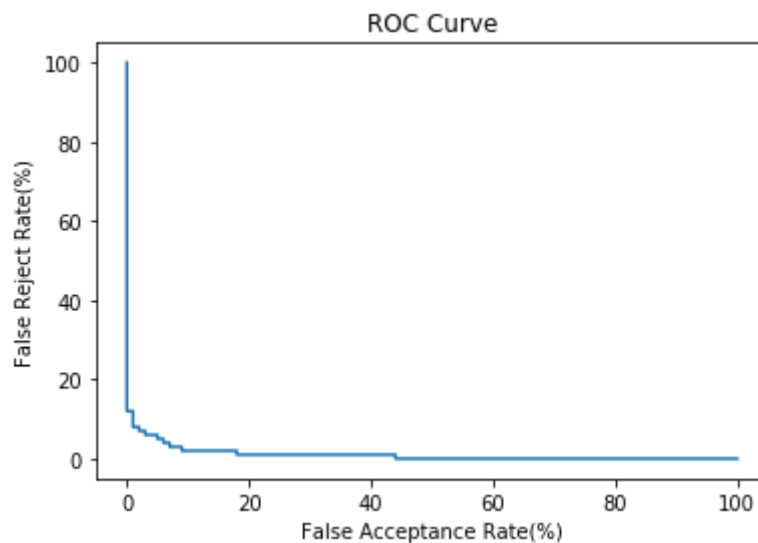
In [19]: plt.plot(FAR,FRR)
          plt.xlabel('False Acceptance Rate(%)')
          plt.ylabel('False Reject Rate(%)')
          plt.title('ROC Curve')

```

```

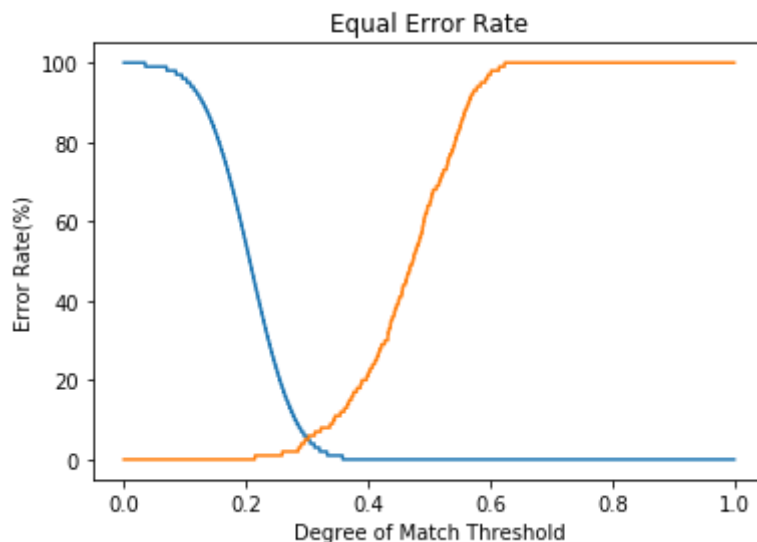
Out[19]: Text(0.5, 1.0, 'ROC Curve')

```



## Equal Error Rate

```
In [20]: p1 = plt.plot(thresh,FAR)
p2 = plt.plot(thresh,FRR)
plt.xlabel('Degree of Match Threshold')
plt.ylabel('Error Rate(%)')
plt.title("Equal Error Rate")
plt.show()
```



```
In [21]: eer = FRR[np.argmin(np.absolute(FAR-FRR))]
print('Equal Error Rate',eer)
op_point = thresh[np.argmin(np.absolute(FAR-FRR))]
print('Operating Point',op_point)
```

Equal Error Rate 5.0  
Operating Point 0.299199999999998337

**Equal Error rate for system 1 is at 5%. The Operating Point or match threshold is 0.29**

```
In [22]: frr1 = np.min(FRR[np.where(FAR==1)])
frr5 = np.min(FRR[np.where(FAR==5)])
frr10 = np.min(FRR[np.where(FAR==10)])
frr20 = np.min(FRR[np.where(FAR==20)])
```

```
In [23]: data = [['1%', frr1], ['5%', frr5], ['10%', frr10], ['20%', frr20]]
df = pd.DataFrame(data, columns = ['FRR', 'FAR'])
df
```

Out[23]:

	FRR	FAR
0	1%	8.0
1	5%	5.0
2	10%	2.0
3	20%	1.0

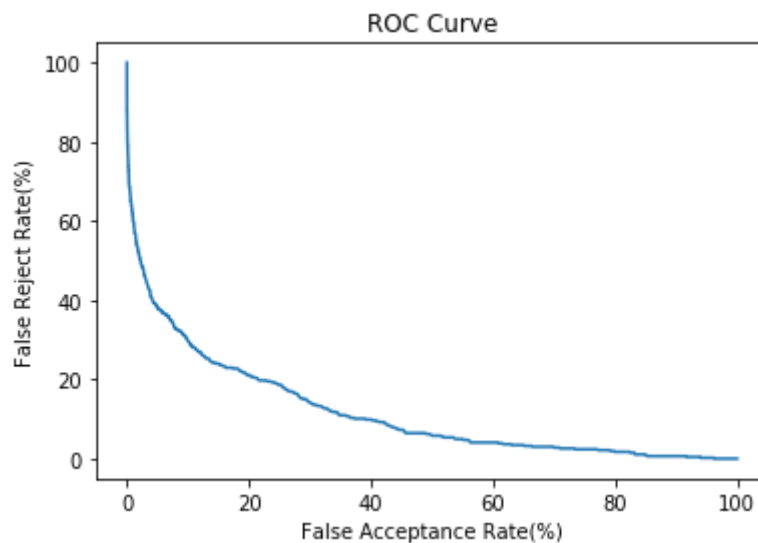


## Svstem2

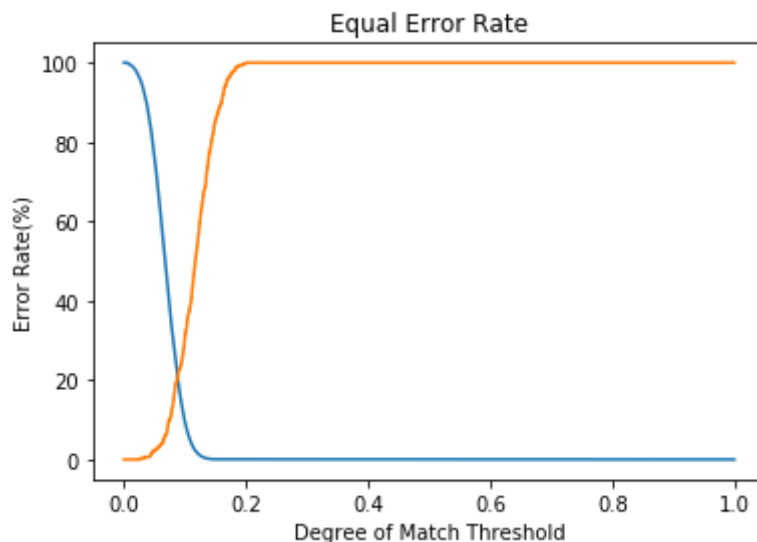
```
In [24]: FAR=[]  
FRR=[]  
thresh = []  
t=0  
while(t<=1):  
    thresh.append(t)  
    far=len(imposter_sys2[imposter_sys2>t])/len(imposter_sys2)  
    FAR.append(far)  
    frr=len(genuine_sys2[genuine_sys2<t])/len(genuine_sys2)  
    FRR.append(frr)  
    t+=0.0001  
  
FAR=np.round(np.array(FAR)*100,2)  
FRR=np.round(np.array(FRR)*100,2)
```

```
In [25]: plt.plot(FAR,FRR)  
plt.xlabel('False Acceptance Rate(%)')  
plt.ylabel('False Reject Rate(%)')  
plt.title('ROC Curve')
```

Out[25]: Text(0.5, 1.0, 'ROC Curve')



```
In [26]: p1 = plt.plot(thresh,FAR)
p2 = plt.plot(thresh,FRR)
plt.xlabel('Degree of Match Threshold')
plt.ylabel('Error Rate(%)')
plt.title("Equal Error Rate")
plt.show()
```



```
In [27]: eer = FRR[np.argmin(np.absolute(FAR-FRR))]
print('Equal Error Rate',eer)
op_point = thresh[np.argmin(np.absolute(FAR-FRR))]
print('Operating Point',op_point)
```

Equal Error Rate 20.6  
Operating Point 0.08800000000000015

**Equal Error rate for system 2 is at 20.6%. The Operating Point or match threshold is 0.088**

```
In [28]: frr1 = np.min(FRR[np.where(np.absolute(FAR-1)<=0.02)])
frr5 = np.min(FRR[np.where(np.absolute(FAR-5)<=0.02)])
frr10 = np.min(FRR[np.where(np.absolute(FAR-10)<=0.02)])
frr20 = np.min(FRR[np.where(np.absolute(FAR-20)<=0.05)])
```

```
In [29]: data = [['1%', frr1], ['5%', frr5], ['10%', frr10], ['20%', frr20]]
df = pd.DataFrame(data, columns = ['FRR', 'FAR'])
df
```

Out[29]:

	FRR	FAR
0	1%	60.30
1	5%	38.20
2	10%	29.83
3	20%	21.03

**System 1 performs better than System2. This can be verified by the fact that System1 has a lower equal error rate than System2. Also, the decidability index of system 1 is more than system 2.**

References: [1] <https://towardsdatascience.com/histograms-and-density-plots-in-python-f6bda88f5ac0>  
(<https://towardsdatascience.com/histograms-and-density-plots-in-python-f6bda88f5ac0>).

In [ ]: