

▼ K-Markov Sampling for Letters dataset

```
1 # Import all dependencies
2 import pandas as pd
3 import numpy as np
4 from sklearn.svm import SVC
5 from sklearn.model_selection import train_test_split
6 from sklearn import metrics
7 from sklearn.metrics import confusion_matrix
8 from sklearn.model_selection import KFold
9 from sklearn.model_selection import cross_val_score
10 from sklearn.model_selection import GridSearchCV
11 import matplotlib.pyplot as plt
12 import seaborn as sns
13 from sklearn.preprocessing import scale
```

```
1 # mount drive for easy import and export of data
2 from google.colab import drive
3 drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive",

```
1 # initialise dataframe with letter dataset
2 letters = pd.read_csv("/content/drive/MyDrive/DM/letter-recognition.csv")
3 letters.columns = ['letter', 'xbox', 'ybox', 'width', 'height', 'onpix', 'xbar', 'ybar', 'x2bar', 'y2bar', 'xyb
```

Step-I

```
1 # initialise parameters
2 markov= pd.DataFrame(columns = letters.columns)
3 uniqChar=list(np.sort(letters['letter'].unique()))
4 classCNT=len(uniqChar)
```

```
4 classCNT=len(uniqChar)
5 limit=100
6 m=classCNT*limit
7 charNo={}
8 c=0
9 for i in uniqChar:
10     charNo[i]=c
11     c+=1
12 mAZ={i:0 for i in uniqChar}
```

```
1 # Chose parameters for markov sampling
2 k=5
3 q=1.2
4 acc=0
```

```
1 # Train a linear Model on N[here 2000] size train set
2 X = letters.drop("letter", axis = 1)
3 y = letters['letter']
4
5 X_scaled = scale(X)
6
7 # train test split
8 X_train, X_test, y_train, y_test = train_test_split(X_scaled, y, test_size = 0.9, random_state = 101)
9 model_linear = SVC(kernel='linear')
10 model_linear.fit(X_train, y_train)
11
12 # predict
13 y_pred = model_linear.predict(X_test)
```

```
1 y_pred

array(['X', 'M', 'E', ..., 'M', 'B', 'J'], dtype=object)
```

Step-II

```
1 # Chosing a random sample as first of markov chain
```

```

2 i=np.random.randint(letters.shape[0])
3 z0=letters.iloc[i]
4 y0=model_linear.predict(np.array([z0.drop('letter')]))[0]
5 if m%classCNT==0:
6     mAZ[z0['letter']] += 1

```

```

1 d={}
2 for i,val in z0.items():
3     print(i,val)
4     d[i]=val
5 markov.append(d,ignore_index=True)
6 markov

```

```

letter J
xbox 3
ybox 6
width 4
height 4
onpix 1
xbar 8
ybar 6
x2bar 5
y2bar 6
xybar 15
x2ybar 7
xy2bar 12
xedge 1
xedgey 6
yedge 1
yedgey 7

```

```

letter xbox ybox width height onpix xbar ybar x2bar y2bar xybar x2ybar xy2bar xedge xedgey yedge

```

```

1 predProb=[]

```

```

1 # Utility Function for loop condition
2 def exist(dic,limit):
3     for i,val in dic.items():
4         if val<limit:
5             return True

```

```
5         return True
6     return False
```

```
1 # Utility loss Function
2 def lossF(actual,pred):
3     if actual==pred:
4         return 1.0
5     return np.exp(-2)
```

```
1 # Utility Function for getting class index
2 def getNo(ch):
3     return charNo[ch]
```

```
1 # Utility function for training subsequent models
2 def train(data):
3
4     X = data.drop("letter", axis = 1)
5     y = data['letter']
6
7     # train test split
8     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.8, random_state = 101)
9     model_linear = SVC(kernel='linear')
10    model_linear.fit(X_train, y_train)
11
12    return model_linear
```

```
1 lst=[]
```

Step-III TO Step-VII

```
1 # Loop markov chain generator k times
2 for km in range(2):
3     # Reset parameters for next markov chain
4     if km!=0:
5         predProb=[]
```

```

6     model_linear=train(markov)
7     markov= pd.DataFrame(columns = letters.columns)
8     mAZ={i:0 for i in uniqChar}
9     lst=[]
10
11    # Chosing a random sample as first of markov chain
12    i=np.random.randint(letters.shape[0])
13    z0=letters.iloc[i]
14    y0=model_linear.predict(np.array([z0.drop('letter')]))[0]
15    if m%classCNT==0:
16        mAZ[z0['letter']] +=1
17    print("Entering...")
18    while exist(mAZ,limit):
19        # choosing a random sample
20        i=np.random.randint(letters.shape[0])
21        while i in lst:
22            i=np.random.randint(letters.shape[0])
23        z1=letters.iloc[i]
24        y1=model_linear.predict(np.array([z1.drop('letter')]))[0]
25        n=lossF(z1['letter'],y1)
26        d=lossF(z0['letter'],y0)
27        p=n/d
28
29        # Deciding of acceptance of chosen sample and its probability in markov chain
30        if acc==k:
31            acc=0
32            p2=q*p
33            p2=min(p2,1)
34            predProb.append([z1['letter'],y1,p2])
35            markov=markov.append(z1)
36            z0=z1
37            mAZ[z1['letter']] +=1
38            acc+=1
39            lst.append(i)
40        elif p==1 and z0['letter']==z1['letter']:
41            n=np.exp(-getNo(y1)*getNo(z1['letter']))
42            d=np.exp(-getNo(y0)*getNo(z0['letter']))
43
44            p1=n/d

```

```
45         p1=min(p1,1)
46         predProb.append([z1['letter'],y1,p1])
47         markov=markov.append(z1)
48         z0=z1
49         mAZ[z1['letter']] +=1
50         acc+=1
51         lst.append(i)
52     elif p<1:
53         predProb.append([z1['letter'],y1,p])
54         markov=markov.append(z1)
55         z0=z1
56         mAZ[z1['letter']] +=1
57         acc+=1
58         lst.append(i)
59     elif p==1 and z0['letter']!=z1['letter']:
60         predProb.append([z1['letter'],y1,p])
61         markov=markov.append(z1)
62         z0=z1
63         mAZ[z1['letter']] +=1
64         acc+=1
65         lst.append(i)
```

```
Entering...
Entering...
```

```
-----
KeyboardInterrupt                                Traceback (most recent call last)
<ipython-input-19-eed1c805cd77> in <module>()
```

```
1 markov
```

```
----> 23         z1=letters.iloc[i1]
```

```
1 predProb
```

Save data from generated markov chain

```
raise_cast_failure)
```

```
1 markov.to_csv("/content/drive/MyDrive/DM/KmarkovSamplesLetters.csv")
```

```
--> 530         if maybe_castable(arr) and not copy and dtype is None:
```

```
1 prob=[]
2 for i in predProb:
3     prob.append(i[2])
```

```
1 markov['probability']=prob
```

```
1 markov.to_csv("/content/drive/MyDrive/DM/KmarkovSamplesLettersProbability.csv")
```

```
1 for i in lst:
2     letters=letters.drop([i])
3 letters.to_csv('/content/drive/MyDrive/DM/KmarkovLettersRemaining.csv')
```

▼ SVM

```
1 train = pd.read_csv("/content/drive/MyDrive/DM/KmarkovSamplesLetters.csv")
2 test = pd.read_csv("/content/drive/MyDrive/DM/KmarkovLettersRemaining.csv")
```

```
1 train = train.drop(train.columns[[0]], axis=1)
2 test = test.drop(test.columns[[0]], axis=1)
3 train.columns = ['letter', 'xbox', 'ybox', 'width', 'height', 'onpix', 'xbar', 'ybar', 'x2bar', 'y2bar', 'xybar']
4 test.columns = ['letter', 'xbox', 'ybox', 'width', 'height', 'onpix', 'xbar', 'ybar', 'x2bar', 'y2bar', 'xybar']
```

```
1 X_train = train.drop("letter", axis = 1)
2 y_train = train["letter"]
3
4 X_test = test.drop("letter", axis = 1)
5 y_test = test["letter"]
```

Linear kernel

```
1 model_linear = SVC(kernel='linear')
2 model_linear.fit(X_train, y_train)
3
4 # predict
5 y_pred = model_linear.predict(X_test)
6 print("accuracy:", metrics.accuracy_score(y_true=y_test, y_pred=y_pred), "\n")
```

accuracy: 0.8147690305790501

RBF kernel

```
1 model_linear = SVC(kernel='rbf')
2 model_linear.fit(X_train, y_train)
3
4 # predict
5 y_pred = model_linear.predict(X_test)
6 print("accuracy:", metrics.accuracy_score(y_true=y_test, y_pred=y_pred), "\n")
```

accuracy: 0.8815875081327261

Chi-squared kernel

```
1 from sklearn.metrics.pairwise import chi2_kernel
2
3 model_linear = SVC(kernel=chi2_kernel)
4 model_linear.fit(X_train, y_train)
5
6 y_pred = model_linear.predict(X_test)
7 print("accuracy:", metrics.accuracy_score(y_true=y_test, y_pred=y_pred), "\n")

accuracy: 0.8559531554977229
```

Hellinger kernel

```
1 def hellinger(X1, X2):
2
3     return np.sqrt(np.dot(X1,X2.T))
4
5
6 model_linear = SVC(kernel=hellinger)
7 model_linear.fit(X_train, y_train)
8
9 # predict
10 y_pred = model_linear.predict(X_test)
11 print("accuracy:", metrics.accuracy_score(y_true=y_test, y_pred=y_pred), "\n")

☐➤ accuracy: 0.7474951203643462
```

Intersection kernel

```
1 from sklearn.metrics.pairwise import euclidean_distances
2
3 def intersection(X1, X2):
```

```

3 def intersection(X1,X2):
4
5     # X1= n1 x m
6     # X2= n2 x m
7     # result= n1xn2
8
9     result = np.zeros((X1.shape[0],X2.shape[0]))
10    X2=X2.T
11
12    for i in range(len(X1)):
13        # iterate through columns of Y
14        for j in range(len(X2[0])):
15            # iterate through rows of Y
16            val=float('+inf')
17            for k in range(len(X2)):
18                val = min(val,X1[i][k] * X2[k][j])
19
20            result[i][j]=val
21
22    return result
23
24
25 model_linear = SVC(kernel=intersection)
26 model_linear.fit(X_train, y_train)
27
28 # predict
29 y_pred = model_linear.predict(X_test)
30 print("accuracy:", metrics.accuracy_score(y_true=y_test, y_pred=y_pred), "\n")
31
32 # Taking too much time.

```

KeyboardInterrupt

Traceback (most recent call last)

[<ipython-input-27-8790cc471789>](#) in <module>()

24

25 model_linear = SVC(kernel=intersection)

---> 26 model_linear.fit(X_train, y_train)

27

28 # predict

3 frames

[<ipython-input-27-8790cc471789>](#) in intersection(X1, X2)

16 val=float('+inf')

17 for k in range(len(X2)):

---> 18 val = min(val,X1[i][k] * X2[k][j])

19

20 result[i][j]=val

KeyboardInterrupt:

SEARCH STACK OVERFLOW