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# Hazırlayanlar

- #### Grup Numarası: 3
- #### Anıl Dursun İpek 031890131
- #### Batuhan Arslandaş 032190097

#### Veri Setinin Yüklenmesi

```
In [1]: # Gerekli Kütüphaneler
    import numpy as np # lineer cebir kütüphanesi
    import matplotlib.pyplot as plt # görselleştirme kütüphanesi
    import pandas as pd
    import seaborn as sns
    from sklearn.metrics import confusion_matrix
    sns.set_style('white')
In [2]: letters = pd.read_csv('letter_recognition.data')
```

### Verinin İncelenmesi

- 1. Data içerisinde null değer bulunmamaktadır.
- 2. Toplam veri sayısı: 20000
- 3. Toplam feature sayısı: 17
- 4. Veri Türleri:
  - Letter: object
  - F1 F16: int64

```
6
                        20000 non-null
              F6
          7
               F7
                        20000 non-null
                                           int64
          8
               F8
                        20000 non-null
          9
               F9
                        20000 non-null int64
          10
              F10
                        20000 non-null
          11
             F11
                        20000 non-null
                                          int64
          12
              F12
                        20000 non-null
                                           int64
          13
              F13
                        20000 non-null
                                          int64
              F14
                        20000 non-null
          14
                                          int64
          15 F15
                        20000 non-null int64
          16 F16
                        20000 non-null int64
         dtypes: int64(16), object(1)
         memory usage: 2.6+ MB
        letters.describe()
                         F1
                                      F2
                                                   F3
                                                                F4
                                                                             F5
                                                                                          F6
                                                                                                       F7
         count 20000.000000 20000.000000 20000.000000 20000.000000 20000.000000 20000.000000
                                                                                              20000.000000 20000.00
                    4.023550
                                 7.035500
                                              5.121850
                                                           5.37245
                                                                        3.505850
                                                                                     6.897600
                                                                                                  7.500450
         mean
                                                                                                               4.62
                    1.913212
                                 3.304555
                                              2.014573
                                                           2.26139
                                                                        2.190458
                                                                                     2.026035
                                                                                                  2.325354
                                                                                                               2.69
           std
                    0.000000
                                 0.000000
                                              0.000000
                                                           0.00000
                                                                        0.000000
                                                                                     0.000000
                                                                                                  0.000000
                                                                                                               0.00
           min
                    3.000000
                                 5.000000
                                              4.000000
                                                           4.00000
                                                                        2.000000
                                                                                                               3.00
          25%
                                                                                     6.000000
                                                                                                  6.000000
          50%
                    4.000000
                                 7.000000
                                              5.000000
                                                           6.00000
                                                                        3.000000
                                                                                     7.000000
                                                                                                  7.000000
                                                                                                               4.00
          75%
                    5.000000
                                 9.000000
                                              6.000000
                                                           7.00000
                                                                        5.000000
                                                                                     8.000000
                                                                                                  9.000000
                                                                                                               6.00
                   15.000000
                                15.000000
                                             15.000000
                                                           15.00000
                                                                       15.000000
                                                                                    15.000000
                                                                                                 15.000000
                                                                                                              15.00
          max
In [5]: # Veri eğitim ve test olarak ayrıldı
         train = letters[:16000]
         test = letters[16000:]
         print("Train length -> ", len(train), " Test length -> ", len(test))
         Train length -> 16000 Test length -> 4000
In [6]: train.head()
                               F4
                                   F5
                                       F6
                                          F7
                                               F8
                                                   F9
                                                       F10
                                                            F11
                                                                 F12
                                                                      F13
                                                                           F14
                                                                                F15
                                                                                     F16
         0
                    2
                                                                                       8
                Τ
                        8
                            3
                                5
                                    1
                                        8
                                           13
                                                0
                                                    6
                                                         6
                                                             10
                                                                   8
                                                                        0
                                                                             8
                                                                                  0
                       12
                                    2
                                       10
                                                5
                                                        13
                                                                                      10
         2
                       11
                            6
                                8
                                    6
                                       10
                                            6
                                                2
                                                    6
                                                        10
                                                              3
                                                                   7
                                                                             7
                                                                  10
                                                                            10
                G
                    2
                            3
                                    1
                                                                                  5
                                                                                      10
         test.head()
                Letter
                           F2
                               F3
                                   F4
                                       F5
                                           F6
                                                   F8
                                                       F9
                                                           F10
                                                                F11
                                                                     F12
                                                                          F13
                                                                               F14
                                                                                    F15
                                                                                         F16
         16000
                    U
                           10
                                6
                                    7
                                        9
                                            9
                                                6
                                                        3
                                                             6
                                                                  7
                                                                       7
                                                                            9
                                                                                 8
                                                                                       5
                                                                                            6
```

13

5

1

2

4

5

In [4]:

Out[4]:

Out[6]:

In [7]:

Out[7]:

16001

6

9

8 7

F1

F2

F3

F4

F5

20000 non-null

20000 non-null

20000 non-null

20000 non-null

20000 non-null

int64

int64

int64

int64

```
16002
                                                                                  10
        16003
                                                                                   7
                      5
                             7
                                        9
                                                   5
                                                                     5
                                                                          8
        16004
                          9
                                               3
                                                       10
                                                            4
                                                                              1
In [8]:
        data = train.values
        data
        array([['T', 2, 8, ..., 8, 0, 8],
                ['I', 5, 12, ..., 8, 4, 10],
                ['D', 4, 11, ..., 7, 3, 9],
                ['G', 8, 14, ..., 7, 5, 8],
                ['E', 4, 7, ..., 8, 5, 8],
                ['C', 2, 1, ..., 9, 4, 10]], dtype=object)
```

```
# Veri seti içerisindeki harfler diziye aktarıldı
In [9]:
        letter unique = np.unique(train["Letter"])
        letter unique
```

```
array(['A', 'B', 'C', 'D', 'E', 'F', 'G', 'H', 'I', 'J', 'K', 'L', 'M',
Out[9]:
               'N', 'O', 'P', 'Q', 'R', 'S', 'T', 'U', 'V', 'W', 'X', 'Y', 'Z'],
              dtype=object)
```

### Verinin Görselleştirilmesi

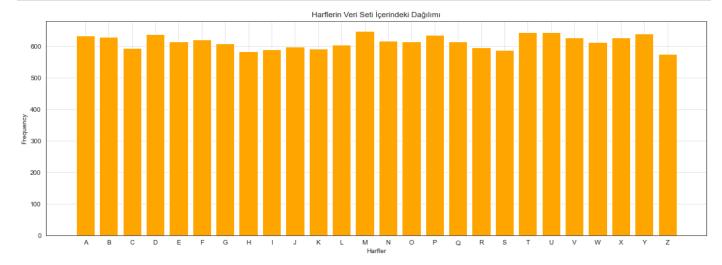
Out[8]:

```
# Eğitim veri seti içerisindeki her bir harf gruplandırıldı ve toplam sayıları diziye ak
In [10]:
        letter num = train.groupby('Letter')['Letter'].count()
        print(letter num)
        letter counts = letter num.values
        letter sum = sum(letter counts)
        print("Toplam Harf Sayısı(Train datası içerisindeki veri sayısı): ",letter sum)
        Letter
             633
        Α
             630
        В
        С
             594
```

D 638 Ε 616 F 622 G 609 Η 583 Т 590 J 599 K 593 L 604 М 648 617 Ν 0 614 Р 635 615 Q R 597 S 587 Т 645 U 645 V 628 613 628 Χ Υ 641 576

Name: Letter, dtype: int64 Toplam Harf Sayısı (Train datası içerisindeki veri sayısı): 16000

```
In [11]: plt.figure(figsize=(18,6))
   plt.bar(letter_num.index,letter_num.values,color="orange")
   plt.xlabel("Harfler")
   plt.ylabel("Frequency")
   plt.title("Harflerin Veri Seti İçerindeki Dağılımı")
   plt.grid(True,alpha=0.6)
   plt.show()
```



### Öncül Olasılıkların Hesabı

# Toy Bayés Yöntemi Kullanımı Ve Açıklaması

```
# Feature'lar algoritma için değişkene aktarıldı
In [13]:
         features = train.columns
         num features = features.shape[0]
        print(features)
         Index(['Letter', 'F1', 'F2', 'F3', 'F4', 'F5', 'F6', 'F7', 'F8', 'F9', 'F10',
                'F11', 'F12', 'F13', 'F14', 'F15', 'F16'],
              dtype='object')
In [14]:
         # Data set içerisindeki hesaplanan koşullu olasılıkların aktarılması için dataframe oluş
         df = pd.DataFrame(columns=["Letter", "Feature", "Value", "Prob"])
         for i in range(1, num features): # Her bir feature'u hesaplaması için oluşturulan döngü
             fi = train.groupby(['Letter', features[i]])[features[i]].count() # Veri seti her bi
             print(f"FEATURE {features[i]}")
             for x in letter unique: # Her bir harfi hesaplaması için oluşturulan döngü
                 values = np.sort(train[train.Letter == x][features[i]].unique()) # Harf içerisin
                 print("Harfin feature içerisindeki değer aralığı \n", values)
                 for y in range(16): # Her bir harfin feature içerisindeki her bir değer aralığın
                     print(f"Harf: {x} Deger {y}")
                     if (np.isin(values, y).sum()):
                         # Eğer değer aralığı içinde var ise toplam kaç adet olduğuna göre hesapl
                         # Örneğin A için F1=15 yok. Bu nedenle harfte F1 için 15'in oluşma sayıı
```

```
# Harf içerisinde oluşma sayısı 0 dan farklı olan feature'lar için oluşm
               result = (fi[fi.index == (x,y)].values[0] + 1) / (letter num[x] + 16)
            else:
               result = (0 + 1) / (letter num[x] + 16)
            print("Sonuç -> ", result)
            df.loc[len(df.index)] = [x,features[i],y,result]
        print("----")
FEATURE F1
Harfin feature içerisindeki değer aralığı
 [1 2 3 4 5 6 7 8 9]
Harf: A Deger 0
Sonuç -> 0.0015408320493066256
Harf: A Deger 1
Sonuç -> 0.07087827426810478
Harf: A Deger 2
Sonuç -> 0.2357473035439137
Harf: A Deger 3
Sonuç -> 0.325115562403698
Harf: A Deger 4
Sonuç -> 0.17257318952234207
Harf: A Deger 5
Sonuc -> 0.0847457627118644
Harf: A Deger 6
Sonuç -> 0.05238828967642527
Harf: A Deger 7
Sonuç -> 0.02773497688751926
Harf: A Deger 8
Sonuç -> 0.012326656394453005
Harf: A Deger 9
Sonuç -> 0.007704160246533128
Harf: A Deger 10
Sonuç -> 0.0015408320493066256
Harf: A Deger 11
Sonuç -> 0.0015408320493066256
Harf: A Deger 12
Sonuç -> 0.0015408320493066256
Harf: A Deger 13
Sonuç -> 0.0015408320493066256
Harf: A Deger 14
Sonuç -> 0.0015408320493066256
Harf: A Deger 15
Sonuç -> 0.0015408320493066256
-----
Harfin feature içerisindeki değer aralığı
[0 1 2 3 4 5 6 7 8 9 10 11]
Harf: B Deger 0
Sonuç -> 0.0030959752321981426
Harf: B Deger 1
Sonuç -> 0.06037151702786378
Harf: B Deger 2
Sonuç -> 0.13312693498452013
Harf: B Deger 3
Sonuç -> 0.21517027863777088
Harf: B Deger 4
```

Sonuc -> 0.24303405572755418

Sonuç -> 0.1609907120743034

Sonuç -> 0.08204334365325078

Sonuç -> 0.04179566563467492

Sonuç -> 0.030959752321981424

Sonuç -> 0.01393188854489164

Harf: B Deger 5

Harf: B Deger 6

Harf: B Deger 7

Harf: B Deger 8

Harf: B Deger 9

```
Harf: B Deger 10
Sonuç -> 0.0030959752321981426
Harf: B Deger 11
Sonuç -> 0.006191950464396285
Harf: B Deger 12
Sonuç -> 0.0015479876160990713
Harf: B Deger 13
Sonuc -> 0.0015479876160990713
Harf: B Deger 14
Sonuç -> 0.0015479876160990713
Harf: B Deger 15
Sonuç -> 0.0015479876160990713
-----
Harfin feature içerisindeki değer aralığı
[0 1 2 3 4 5 6 7 8 9 10]
Harf: C Deger 0
Sonuç -> 0.003278688524590164
Harf: C Deger 1
Sonuç -> 0.054098360655737705
Harf: C Deger 2
Sonuc -> 0.16229508196721312
Harf: C Deger 3
Sonuç -> 0.1819672131147541
Harf: C Deger 4
Sonuç -> 0.22131147540983606
Harf: C Deger 5
Sonuç -> 0.16229508196721312
Harf: C Deger 6
Sonuç -> 0.10655737704918032
Harf: C Deger 7
Sonuç -> 0.054098360655737705
Harf: C Deger 8
Sonuç -> 0.02459016393442623
Harf: C Deger 9
Sonuç -> 0.018032786885245903
Harf: C Deger 10
Sonuç -> 0.003278688524590164
Harf: C Deger 11
Sonuç -> 0.001639344262295082
Harf: C Deger 12
Sonuç -> 0.001639344262295082
Harf: C Deger 13
Sonuç -> 0.001639344262295082
Harf: C Deger 14
Sonuç -> 0.001639344262295082
Harf: C Deger 15
Sonuç -> 0.001639344262295082
-----
Harfin feature içerisindeki değer aralığı
[1 2 3 4 5 6 7 8 9 10]
Harf: D Deger 0
Sonuç -> 0.0015290519877675841
Harf: D Deger 1
Sonuc -> 0.03669724770642202
Harf: D Deger 2
Sonuç -> 0.13914373088685014
Harf: D Deger 3
Sonuc -> 0.1926605504587156
Harf: D Deger 4
Sonuç -> 0.25229357798165136
Harf: D Deger 5
Sonuç -> 0.18501529051987767
Harf: D Deger 6
Sonuç -> 0.10856269113149847
Harf: D Deger 7
Sonuç -> 0.03363914373088685
```

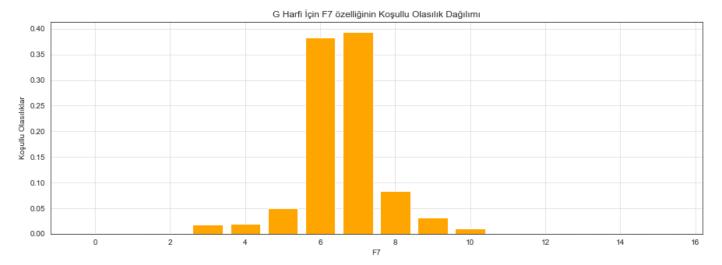
```
Harf: Y Deger 2
Sonuç -> 0.0015220700152207
Harf: Y Deger 3
Sonuç -> 0.060882800608828
Harf: Y Deger 4
Sonuç -> 0.1659056316590563
Harf: Y Deger 5
Sonuç -> 0.1689497716894977
Harf: Y Deger 6
Sonuc -> 0.0730593607305936
Harf: Y Deger 7
Sonuç -> 0.0715372907153729
Harf: Y Deger 8
Sonuc -> 0.410958904109589
Harf: Y Deger 9
Sonuç -> 0.0289193302891933
Harf: Y Deger 10
Sonuç -> 0.0060882800608828
Harf: Y Deger 11
Sonuç -> 0.0015220700152207
Harf: Y Deger 12
Sonuç -> 0.0030441400304414
Harf: Y Deger 13
Sonuç -> 0.0015220700152207
Harf: Y Deger 14
Sonuç -> 0.0015220700152207
Harf: Y Deger 15
Sonuç -> 0.0015220700152207
-----
Harfin feature içerisindeki değer aralığı
[ 3 4 5 6 7 8 9 10 11 12]
Harf: Z Deger 0
Sonuç -> 0.0016891891891893
Harf: Z Deger 1
Sonuç -> 0.0016891891891893
Harf: Z Deger 2
Sonuç -> 0.0016891891891893
Harf: Z Deger 3
Sonuç -> 0.006756756756756757
Harf: Z Deger 4
Sonuç -> 0.018581081081082
Harf: Z Deger 5
Sonuc -> 0.05067567567568
Harf: Z Deger 6
Sonuç -> 0.06418918918919
Harf: Z Deger 7
Sonuç -> 0.18074324324324326
Harf: Z Deger 8
Sonuç -> 0.5405405405406
Harf: Z Deger 9
Sonuç -> 0.0777027027027027
Harf: Z Deger 10
Sonuç -> 0.02533783783783784
Harf: Z Deger 11
Sonuç -> 0.016891891891893
Harf: Z Deger 12
Sonuç -> 0.008445945945946
Harf: Z Deger 13
Sonuç -> 0.0016891891891893
Harf: Z Deger 14
Sonuç -> 0.0016891891891893
Harf: Z Deger 15
Sonuç -> 0.0016891891891893
_____
```

```
Out[15]:
             Letter Feature Value
                                     Prob
          0
                 Α
                        F1
                                0 0.001541
          1
                        F1
                                1 0.070878
                 Δ
          2
                 Α
                        F1
                                2 0.235747
                                3 0.325116
          3
                 Α
                        F1
                                4 0.172573
          4
                 Α
                        F1
In [16]:
          df.tail()
                Letter Feature Value
                                         Prob
Out[16]:
          6651
                    Ζ
                          F16
                                  11 0.016892
          6652
                    Ζ
                           F16
                                  12 0.008446
                    Ζ
                          F16
                                  13 0.001689
          6653
                                  14 0.001689
          6654
                    Ζ
                           F16
          6655
                    Ζ
                          F16
                                  15 0.001689
          df[(df["Letter"] == "A") & (df["Feature"] == "F1")].Prob.values
In [17]:
          array([0.00154083, 0.07087827, 0.2357473 , 0.32511556, 0.17257319,
Out[17]:
                  0.08474576, 0.05238829, 0.02773498, 0.01232666, 0.00770416,
                  0.00154083, 0.00154083, 0.00154083, 0.00154083, 0.00154083,
                  0.00154083])
In [18]: # Örnek görselleştirme
          plt.figure(figsize=(15,5))
          plt.bar(range(16), df[(df["Letter"] == "A") & (df["Feature"] == "F1")].Prob.values, colo
          plt.xlabel("F1")
          plt.ylabel("Koşullu Olasılıklar")
          plt.title("A Harfi İçin F1 özelliğinin Koşullu Olasılık Dağılımı")
          plt.grid(True, alpha=0.6)
          plt.show()
                                                A Harfi İçin F1 özelliğinin Koşullu Olasılık Dağılımı
           0.30
           0.25
          Koşullu Olasılıkla
           0.20
           0.15
           0.10
           0.05
           0.00
                                                                             10
```

In [15]: df.head()

```
In [19]: # Örnek görselleştirme - 2
plt.figure(figsize=(15,5))
plt.bar(range(16), df[(df["Letter"] == "G") & (df["Feature"] == "F7")].Prob.values, colo
plt.xlabel("F7")
plt.ylabel("Koşullu Olasılıklar")
```

```
plt.title("G Harfi İçin F7 özelliğinin Koşullu Olasılık Dağılımı")
plt.grid(True,alpha=0.6)
plt.show()
```



# Test Verisi Üzerinden Tahminde Bulunma

```
predictions = pd.DataFrame(columns=["Real Val", "Prediction", "Accuracy"]) # Tahminlerin s
In [20]:
         for i in range(len(test)): # Test verisi üzerindeki her bir veri döngü ile test edildi
             total = 0
            maximum = 0
            probs = [] # Argmax almak için her bir harfin olasılık sonuçları diziye aktarıldı
             print(test.values[i])
             for j in letter unique: # Her bir test verisi için 26 harf test edildi ve olasılıksa
                 total = 0
                 for k in range(1, num features):
                     total += df[(df["Letter"] == j) & (df["Feature"] == features[k]) & (df["Valu
                 total += np.log10(letter num[test.values[i][0]] / sum(letter num.values))
                 probs.append(total)
             #print("probs ->",probs)
             predictions.loc[len(predictions.index)] = [test.values[i][0],letter unique[np.argmax
             # Diziye aktarılan değerler arasındaki maksimum orana sahip olan değer tahmin olarak
             print("İşlenen Veri -> ", (i+1), " || Max -> ", probs[np.argmax(probs)] , "Harf ->",
```

```
['U' 4 10 6 7 9 9 6 4 3 6 7 7 9 8 5 6]

İşlenen Veri -> 1 || Max -> 1.4165852428152563 Harf -> B

['N' 6 9 8 4 3 8 7 3 4 13 5 8 6 8 0 8]

İşlenen Veri -> 2 || Max -> 2.823788943753555 Harf -> I

['V' 6 9 8 8 10 7 7 5 4 7 6 8 7 9 7 10]

İşlenen Veri -> 3 || Max -> 2.649945271642327 Harf -> I

['I' 5 6 6 4 3 7 6 2 7 7 6 9 0 9 4 8]

İşlenen Veri -> 4 || Max -> 3.1789432501083326 Harf -> I

['N' 5 9 7 6 4 9 7 3 5 10 4 6 5 8 1 7]

İşlenen Veri -> 5 || Max -> 2.31602300128253 Harf -> N

['H' 5 8 8 6 6 5 8 3 6 10 8 8 4 8 4 6]

İşlenen Veri -> 6 || Max -> 2.0506971530713702 Harf -> H

['E' 0 0 1 0 0 5 7 5 6 7 6 12 0 8 6 10]

İşlenen Veri -> 7 || Max -> 2.9006422476603153 Harf -> I
```

\_\_\_\_\_\_

```
['Y' 1 0 2 0 0 7 10 3 1 7 12 8 1 11 0 8]
İşlenen Veri -> 8 || Max -> 2.5053255658887674 Harf -> Y
______
['G' 5 9 7 7 5 6 7 7 5 5 6 11 4 8 4 8]
İşlenen Veri -> 9 || Max -> 2.6344655639452044 Harf -> 0
_____
['E' 2 4 4 3 2 7 8 2 8 11 7 9 2 8 4 8]
İşlenen Veri -> 10 || Max -> 2.867106299128754 Harf -> E
_____
['N' 5 4 5 6 2 7 7 15 2 4 6 8 6 8 0 8]
İşlenen Veri -> 11 || Max -> 3.647221286987878 Harf -> I
-----
['B' 1 3 2 1 1 8 7 3 4 10 5 7 1 8 3 9]
İşlenen Veri -> 12 || Max -> 2.0286570993053967 Harf -> B
______
['G' 4 7 5 5 3 7 6 7 7 7 5 11 2 9 4 8]
İşlenen Veri -> 13 || Max -> 2.505297309976951 Harf -> G
_____
['L' 4 9 4 6 2 0 2 4 6 1 0 7 0 8 0 8]
İşlenen Veri -> 14 || Max -> 2.283187583027914 Harf -> I
______
['E' 4 7 6 5 5 8 9 7 3 6 6 11 4 7 7 10]
İşlenen Veri -> 15 || Max -> 1.5838759434546175 Harf -> Q
_____
['G' 6 8 8 7 9 7 6 6 4 7 7 9 10 8 9 9]
İşlenen Veri -> 16 || Max -> 1.7972694618756844 Harf -> E
['M' 4 10 6 8 6 7 6 6 5 7 7 9 8 5 2 8]
İşlenen Veri -> 17 || Max -> 2.842394782250813 Harf -> M
     _____
['D' 2 4 4 3 3 9 7 4 6 10 4 6 2 8 3 8]
İşlenen Veri -> 18 || Max -> 3.297948402487256 Harf -> D
_____
['Y' 6 7 6 5 2 3 12 5 5 13 12 6 2 11 2 6]
İşlenen Veri -> 19 || Max -> 1.5839292694334248 Harf -> F
______
['R' 5 5 5 6 3 5 12 8 3 7 2 9 3 7 6 11]
İşlenen Veri -> 20 || Max -> 2.0335182962711604 Harf -> R
_____
['P' 4 7 6 10 9 8 11 5 0 9 7 6 4 10 5 8]
İşlenen Veri -> 21 || Max -> 1.3728472910231484 Harf -> P
 -----
['D' 3 9 5 7 5 8 7 5 7 7 6 5 3 8 3 7]
İşlenen Veri -> 22 || Max -> 3.340761858144748 Harf -> D
______
['E' 4 9 4 7 3 3 6 6 12 7 7 15 0 8 7 7]
İşlenen Veri -> 23 || Max -> 2.276916425711032 Harf -> E
_____
['W' 4 5 6 4 3 7 11 2 3 7 9 8 7 11 1 8]
İşlenen Veri -> 24 || Max -> 3.103213305852951 Harf -> W
     _____
['D' 4 2 5 4 3 7 7 7 7 6 6 5 2 8 3 7]
İşlenen Veri -> 25 || Max -> 3.10528785202854 Harf -> D
______
['Q' 8 13 7 7 4 7 5 4 8 10 5 9 3 7 9 9]
İşlenen Veri -> 26 || Max -> 1.2013946110966536 Harf -> R
_____
['R' 4 7 5 5 5 7 8 5 6 6 4 8 3 6 5 9]
İşlenen Veri -> 27 || Max -> 2.2112533468039954 Harf -> H
______
['G' 4 6 4 4 3 6 7 5 5 9 7 10 2 9 4 10]
İşlenen Veri -> 28 || Max -> 2.89889730997695 Harf -> G
 -----
['Y' 3 7 4 5 2 8 10 2 2 6 12 8 2 11 0 8]
İşlenen Veri -> 29 || Max -> 2.7366802082023147 Harf -> Y
```

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```
['Q' 3 5 4 6 4 8 9 4 2 5 8 11 2 10 5 8]
İşlenen Veri -> 3968 || Max -> 1.9543203505107967 Harf -> V
______
['C' 6 9 6 7 4 4 7 5 7 10 9 14 4 9 5 5]
İşlenen Veri -> 3969 || Max -> 1.2580174456838165 Harf -> U
______
['S' 4 9 6 6 7 9 4 4 4 9 6 9 4 7 10 9]
İşlenen Veri -> 3970 || Max -> 1.1942081675313305 Harf -> R
_____
['F' 7 10 9 8 7 9 7 2 6 12 4 6 5 9 4 9]
İşlenen Veri -> 3971 || Max -> 1.257866120689328 Harf -> D
-----
['C' 5 10 7 9 8 5 6 4 4 7 6 11 5 11 8 10]
İşlenen Veri -> 3972 || Max -> 0.9686812406503924 Harf -> K
______
['V' 4 7 6 5 6 8 6 4 2 7 8 8 7 9 4 6]
İşlenen Veri -> 3973 || Max -> 2.153458103052654 Harf -> W
_____
['T' 4 4 5 3 2 5 12 2 8 11 9 4 0 10 2 4]
İşlenen Veri -> 3974 || Max -> 1.5146681737342598 Harf -> T
______
['N' 5 9 5 4 2 9 11 5 3 5 6 9 5 11 2 6]
İşlenen Veri -> 3975 || Max -> 1.246743567983101 Harf -> Y
______
['E' 1 0 1 0 0 5 8 5 7 7 6 12 0 8 6 10]
İşlenen Veri -> 3976 || Max -> 2.8395861420497543 Harf -> I
['L' 3 8 3 6 2 0 2 4 6 1 0 8 0 8 0 8]
İşlenen Veri -> 3977 || Max -> 2.824441708440455 Harf -> I
     _____
['A' 3 9 5 6 2 6 5 3 1 6 1 8 2 7 2 7]
İşlenen Veri -> 3978 || Max -> 2.8530618475463303 Harf -> A
_____
['K' 5 11 5 8 5 3 8 7 3 6 4 11 3 8 2 11]
İşlenen Veri -> 3979 || Max -> 1.8743534299201599 Harf -> K
______
['M' 6 9 10 7 12 7 5 3 2 7 5 8 15 7 4 6]
İşlenen Veri -> 3980 || Max -> 1.4839157911612462 Harf -> H
______
['R' 2 3 3 2 2 7 7 5 5 7 5 6 2 7 4 8]
İşlenen Veri -> 3981 || Max -> 2.53693371355281 Harf -> 0
 -----
['S' 6 12 6 7 3 6 8 3 6 13 7 7 2 9 3 7]
İşlenen Veri -> 3982 || Max -> 1.5136984464605423 Harf -> C
______
['Y' 3 9 5 6 3 7 9 1 6 6 11 8 2 11 2 7]
İşlenen Veri -> 3983 || Max -> 2.1902570727380826 Harf -> Y
_____
['V' 7 10 5 5 2 6 11 5 4 11 9 4 4 11 3 10]
İşlenen Veri -> 3984 || Max -> 1.160659937578967 Harf -> P
    _____
['S' 2 0 2 1 1 8 7 4 6 5 6 8 0 8 7 8]
İşlenen Veri -> 3985 || Max -> 3.9539570624860794 Harf -> I
______
['M' 5 6 8 4 5 9 6 2 4 9 5 7 8 6 2 8]
İşlenen Veri -> 3986 || Max -> 2.3288405653833433 Harf -> M
-----
['0' 9 15 6 8 5 5 7 7 4 10 7 10 5 9 5 8]
İşlenen Veri -> 3987 || Max -> 1.6903975948344494 Harf -> 0
______
['L' 3 7 3 5 1 0 1 6 6 0 0 6 0 8 0 8]
İşlenen Veri -> 3988 || Max -> 2.368852439836175 Harf -> L
['D' 6 9 8 8 8 7 6 5 7 7 5 9 6 5 10 3]
İşlenen Veri -> 3989 || Max -> 1.060038045462828 Harf -> M
```

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```
['P' 2 1 3 2 1 4 10 3 5 10 8 5 0 9 3 7]
İşlenen Veri -> 3990 || Max -> 1.6503778805670852 Harf -> F
______
['W' 3 8 5 6 5 11 11 2 2 5 8 7 7 12 1 7]
İşlenen Veri -> 3991 || Max -> 1.721655277236099 Harf -> W
_____
['0' 4 3 5 4 2 7 6 8 8 6 5 7 3 8 4 8]
İşlenen Veri -> 3992 || Max -> 3.0142071186439736 Harf -> 0
_____
['E' 4 9 5 6 3 5 9 2 10 10 8 9 2 8 5 5]
İşlenen Veri -> 3993 || Max -> 1.8164733877363481 Harf -> E
______
['J' 2 11 3 8 2 15 4 4 5 13 1 8 0 7 0 8]
İşlenen Veri -> 3994 || Max -> 2.3554850575551685 Harf -> I
______
['T' 5 8 7 7 7 7 9 4 8 7 7 8 3 10 8 6]
İşlenen Veri -> 3995 || Max -> 1.8675432378224146 Harf -> H
_____
['D' 2 2 3 3 2 7 7 7 6 6 6 4 2 8 3 7]
İşlenen Veri -> 3996 || Max -> 2.6229229182874594 Harf -> 0
_____
['C' 7 10 8 8 4 4 8 6 9 12 9 13 2 9 3 7]
İşlenen Veri -> 3997 || Max -> 1.6532730197023178 Harf -> C
_____
['T' 6 9 6 7 5 6 11 3 7 11 9 5 2 12 2 4]
İşlenen Veri -> 3998 || Max -> 1.6614155262304782 Harf -> T
['S' 2 3 4 2 1 8 7 2 6 10 6 8 1 9 5 8]
İşlenen Veri -> 3999 || Max -> 2.5864775780511495 Harf -> Z
['A' 4 9 6 6 2 9 5 3 1 8 1 8 2 7 2 8]
İşlenen Veri -> 4000 || Max -> 2.8176227104122775 Harf -> A
```

#### In [21]: predictions.head(20)

#### Out[21]: Real\_Val Prediction Accuracy

	Real_Val	Prediction	Accuracy
0	U	В	False
1	N	1	False
2	V	1	False
3	1	1	True
4	N	N	True
5	Н	Н	True
6	Е	1	False
7	Υ	Υ	True
8	G	0	False
9	Е	Е	True
10	N	1	False
11	В	В	True
12	G	G	True
13	L	1	False
14	Е	Q	False
15	G	Е	False
16	М	М	True

17	D	D	True
18	Υ	F	False
19	R	R	True

```
In [25]: # Toplam tahmin başarı oranı
print("Başarı Yüzdesi -> ", predictions.Accuracy.values.sum() / len(predictions))
Başarı Yüzdesi -> 0.52
```

# Karmaşıklık Matrisi

```
In [23]: # Confusin Matrix
    confusion_mtx = confusion_matrix(predictions["Real_Val"], predictions["Prediction"])
    # plot the confusion matrix
    f,ax = plt.subplots(figsize=(20, 20))
    sns.heatmap(confusion_mtx, annot=True, linewidths=0.01,cmap="Greens",linecolor="gray", f.
    plt.xlabel("Predicted Label")
    plt.ylabel("True Label")
    plt.title("Confusion Matrix")
    plt.show()
```

Confusion Matrix																										
4	116.0	3.0	0.0	4.0	0.0	0.0	2.0	6.0	3.0	0.0	2.0	0.0	0.0	1.0	9.0	0.0	0.0	4.0	1.0	0.0	0.0	0.0	0.0	5.0	0.0	0.0
a [	0.0		0.0	2.0	0.0	0.0	4.0	5.0	12.0	0.0	1.0	0.0	1.0	2.0	3.0	0.0	0.0	11.0	1.0	0.0	1.0	0.0	0.0	5.0	0.0	0.0
o	0.0	4.0		0.0	2.0	1.0	7.0	1.0	12.0	0.0	3.0	0.0	0.0	0.0	11.0	0.0	3.0	0.0	0.0	0.0	6.0	2.0	0.0	4.0	0.0	1.0
٥	0.0	19.0	0.0	60.0	1.0	0.0	3.0	5.0	28.0	0.0	0.0	0.0	1.0	1.0	19.0	0.0	0.0	9.0	1.0	0.0	0.0	0.0	0.0	20.0	0.0	0.0
ш	0.0	2.0	3.0	0.0	50.0	0.0	4.0	6.0	29.0	0.0	5.0	0.0	0.0	0.0	5.0	1.0	6.0	5.0	1.0	0.0	2.0	1.0	0.0	27.0	0.0	5.0
ш	0.0	3.0	3.0	4.0	1.0		2.0	3.0	19.0	0.0	0.0	0.0	0.0	7.0	3.0	14.0	0.0	1.0	0.0	0.0	1.0	0.0	1.0	3.0	0.0	1.0
O	0.0	6.0	2.0	1.0	4.0	0.0		2.0	8.0	0.0	2.0	0.0	1.0	2.0	16.0	0.0	4.0	9.0	1.0	0.0	2.0	0.0	5.0	7.0	0.0	3.0
Ξ	0.0	9.0	1.0	7.0	2.0	0.0	1.0	67.0	23.0	0.0	1.0	0.0	2.0	2.0	5.0	0.0	2.0	10.0	2.0	0.0	6.0	1.0	2.0	7.0	1.0	0.0
-	0.0	7.0	1.0	4.0	1.0	0.0	0.0	1.0	134.0	0.0	0.0	0.0	0.0	0.0	1.0	2.0	3.0	7.0	0.0	0.0	1.0	0.0	0.0	0.0	0.0	3.0
-	0.0	11.0	0.0	3.0	0.0	1.0	0.0	6.0	13.0	95.0	0.0	2.0	3.0	0.0	3.0	0.0	0.0	4.0	2.0	0.0	0.0	0.0	0.0	1.0	0.0	4.0
×	2.0	4.0	1.0	7.0	1.0	1.0	4.0	14.0	6.0	0.0	54.0	0.0	6.0	0.0	14.0	0.0	1.0	13.0	0.0	0.0	0.0	0.0	0.0	18.0	0.0	0.0
_	19.0	4.0	1.0	0.0	2.0	0.0	3.0	5.0	21.0	4.0	3.0	75.0	0.0	3.0	4.0	0.0	6.0	3.0	0.0	0.0	2.0	0.0	0.0	1.0	0.0	1.0
abel M	8.0	5.0	1.0	1.0	0.0	2.0	1.0	7.0	11.0	0.0	3.0	0.0	87.0	4.0	6.0	0.0	0.0	2.0	0.0	0.0	1.0	0.0	1.0	3.0	0.0	1.0
True Label N M	0.0	8.0	0.0	2.0	0.0	1.0	1.0	5.0	50.0	4.0	0.0	0.0	3.0	59.0	6.0	3.0	1.0	4.0	0.0	0.0	12.0	3.0	2.0	1.0	1.0	0.0
0	0.0	1.0	0.0	0.0	1.0	0.0	5.0	3.0	14.0	0.0	0.0	0.0	0.0	2.0	92.0	1.0	4.0	6.0	1.0	0.0	3.0	0.0	1.0	5.0	0.0	0.0
۵	0.0	5.0	1.0	5.0	1.0	6.0	2.0	0.0	5.0	0.0	0.0	0.0	0.0	1.0	7.0	114.0	1.0	1.0	3.0	0.0	3.0	0.0	9.0	1.0	3.0	0.0
σ	1.0	6.0	0.0	2.0	1.0	0.0	13.0	8.0	5.0	0.0	2.0	0.0	3.0	1.0	37.0	0.0	65.0	6.0	0.0	0.0	9.0	4.0	1.0	3.0	1.0	0.0
œ	2.0	18.0	0.0	7.0	0.0	0.0	5.0	9.0	3.0	0.0	0.0	0.0	3.0	5.0	16.0	0.0	1.0	86.0	2.0	0.0	1.0	0.0	0.0	3.0	0.0	0.0
s	3.0	13.0	4.0	0.0	1.0	1.0	7.0	9.0	30.0	1.0	0.0	0.0	0.0	2.0	22.0	1.0	3.0	9.0	34.0	0.0	4.0	2.0	0.0	9.0	0.0	6.0
-	0.0	1.0	1.0	2.0	3.0	14.0	4.0	3.0	33.0	0.0	2.0	0.0	0.0	0.0	1.0	1.0	1.0	2.0	1.0	37.0	3.0	22.0	0.0	5.0	15.0	0.0
٥	0.0	6.0	3.0	1.0	1.0	2.0	3.0	7.0	10.0	0.0	1.0	0.0	5.0	0.0	11.0	0.0	5.0	1.0	0.0	0.0	105.0	3.0	1.0	1.0	0.0	2.0
>	0.0	0.0	0.0	1.0	0.0	1.0	2.0	0.0	3.0	0.0	0.0	0.0	0.0	3.0	4.0	4.0	1.0	1.0	0.0	0.0	7.0	95.0	9.0	1.0	4.0	0.0
^	0.0	2.0	1.0	0.0	0.0	3.0	0.0	0.0	12.0	0.0	0.0	0.0	0.0	1.0	5.0	0.0	0.0	0.0	0.0	0.0	1.0	12.0	100.0	1.0	1.0	0.0
×	2.0	9.0	1.0	1.0	3.0	2.0	3.0	4.0	39.0	0.0	4.0	0.0	0.0	2.0	9.0	0.0	2.0	0.0	0.0	0.0	5.0	0.0	0.0	70.0	0.0	3.0
>	0.0	1.0	0.0	3.0	0.0	8.0	1.0	3.0	0.0	0.0	0.0	0.0	0.0	0.0	3.0	10.0	4.0	0.0	0.0	0.0	7.0	16.0	7.0	0.0	82.0	0.0
Z	0.0	9.0	1.0	1.0	5.0	4.0	2.0	3.0	52.0	0.0	0.0	0.0	0.0	0.0	2.0	0.0	2.0	5.0	2.0	0.0	8.0	1.0	1.0	6.0	0.0	54.0
l	A	В	С	D	Е	F	G	Н	ı	J	К	L	M Predicte	N ed Label	0	Р	Q	R	s	Т	U	V	w	Х	Υ	Z

In []: