neural-networks

November 25, 2024

```
[1]: import pandas as pd
     from sklearn.metrics import accuracy_score, confusion_matrix
     from sklearn.model_selection import train_test_split, cross_val_score
     # Create your first MLP in Keras
     from keras.models import Sequential
     from keras.layers import Dense
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
[2]: data = pd.read_csv("/content/Alphabets_data.csv")
     data
[2]:
            letter
                    xbox
                           ybox
                                  width
                                         height
                                                  onpix xbar
                                                                 ybar
                                                                        x2bar
                                                                                y2bar
     0
                 Τ
                        2
                               8
                                       3
                                               5
                                                       1
                                                              8
                                                                    13
                                                                            0
                                                                                    6
     1
                 Ι
                        5
                              12
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                                               7
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     2
                 D
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                 G
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     19995
                 D
                        2
                               2
                                       3
                                               3
                                                       2
                                                                                    6
     19996
                 С
                        7
                              10
                                      8
                                               8
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                 Т
                                               7
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     19997
                        6
                               9
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                                                                            3
                                                                    11
     19998
                 S
                        2
                               3
                                       4
                                               2
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     19999
                 Α
                        4
                                                                            3
                                                                                    1
             xybar
                    x2ybar
                             xy2bar
                                      xedge
                                              xedgey
                                                       yedge
                                                              yedgex
     0
                 6
                         10
                                   8
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     1
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     19995
                 6
                                   4
                                                    8
     19996
                12
                          9
                                  13
                                           2
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                                                            3
                                                                     7
```

19997	11	9	5	2	12	2	4
19998	10	6	8	1	9	5	8
19999	8	1	8	2	7	2	8

[20000 rows x 17 columns]

[5]: data.info()

<class 'pandas.core.frame.DataFrame'> RangeIndex: 20000 entries, 0 to 19999 Data columns (total 17 columns):

#	Column	Non-Nu	ıll Count	Dtype
0	letter	20000	non-null	object
1	xbox	20000	non-null	int64
2	ybox	20000	non-null	int64
3	width	20000	non-null	int64
4	height	20000	non-null	int64
5	onpix	20000	non-null	int64
6	xbar	20000	non-null	int64
7	ybar	20000	non-null	int64
8	x2bar	20000	non-null	int64
9	y2bar	20000	non-null	int64
10	xybar	20000	non-null	int64
11	x2ybar	20000	non-null	int64
12	xy2bar	20000	non-null	int64
13	xedge	20000	non-null	int64
14	xedgey	20000	non-null	int64
15	yedge	20000	non-null	int64
16	yedgex	20000	non-null	int64
dtyp	es: int6	4(16),	object(1)	

memory usage: 2.6+ MB

[7]: data.corr(numeric_only=True)

[7]: xbox width height onpix xbar ybar ybox xbox 1.000000 0.757793 0.851514 0.672764 0.619097 -0.032595 0.045545 0.757793 1.000000 0.671912 0.823207 0.555067 0.045690 -0.040925 ybox width 1.000000 0.660215 0.765716 0.851514 0.671912 0.061959 0.024832 height 0.672764 0.823207 0.660215 1.000000 0.644366 0.042844 -0.020072 onpix 0.619097 0.555067 0.765716 0.644366 1.000000 0.139159 -0.028822 xbar -0.032595 0.045690 0.061959 0.042844 0.139159 1.000000 -0.356580 0.045545 -0.040925 0.024832 -0.020072 -0.028822 -0.356580 1.000000 ybar x2bar 0.014306 -0.025019 -0.098611 0.082383 -0.011985 -0.053306 -0.118625 y2bar 0.052086 0.096478 0.057074 0.059032 -0.065557 -0.122851 -0.049658 0.012458 -0.069776 0.085963 0.178318 xybar 0.148056 0.159954 0.115018 x2ybar 0.011694 -0.011991 -0.072941 -0.341957 0.600397 0.035464 -0.054648

```
xy2bar -0.046333 -0.007568 -0.045009
                                         0.026386 -0.038858 -0.032115 -0.271649
                                                   0.627507
                                                             0.144325 -0.036722
    xedge
            0.489155 0.274431
                                0.557251
                                         0.265243
    xedgey
            0.098180 -0.001336
                                0.045658
                                         0.025359
                                                   0.017649 -0.253339 0.555060
    yedge
            0.273504 0.230883
                                0.260285
                                         0.297545
                                                   0.492653
                                                             0.127056 -0.078008
    yedgex -0.105147 -0.042741 -0.118273 -0.018853 -0.062969
                                                             0.248816 -0.207900
                                                                xedge
               x2bar
                         y2bar
                                   xybar
                                            x2ybar
                                                     xy2bar
                                                                         xedgey
            0.014306
                      0.052086
    xbox
                                0.148056 0.035464 -0.046333
                                                             0.489155
                                                                      0.098180
           -0.025019
                      0.096478
                                0.159954 -0.054648 -0.007568
                                                             0.274431 -0.001336
    ybox
    width
           -0.098611
                      0.057074
                                0.115018 0.011694 -0.045009
                                                             0.557251
                                                                       0.045658
    height 0.082383
                      0.059032
                                0.012458 -0.011991
                                                   0.026386
                                                             0.265243
                                                                       0.025359
    onpix -0.011985 -0.065557 -0.069776 -0.072941 -0.038858
                                                             0.627507
                                                                       0.017649
    xbar
           -0.053306 -0.122851
                                0.085963 -0.341957 -0.032115
                                                             0.144325 -0.253339
    ybar
           -0.118625 -0.049658
                                0.082020
                                                             0.142132 -0.084820
    x2bar
            1.000000 -0.188431 -0.317780
                                         0.042545
    y2bar
           -0.188431 1.000000
                                0.132000 -0.060116
                                                   0.119048 -0.384018 -0.052545
                                         0.057988 -0.106759 -0.175676 0.029419
           -0.317780
                      0.132000
                                1.000000
    xybar
    x2ybar 0.042545 -0.060116
                                0.057988
                                         1.000000
                                                   0.063214
                                                             0.053566
                                                                       0.527239
    xy2bar
           0.082020 0.119048 -0.106759
                                         0.063214
                                                   1.000000 -0.008753 -0.184927
            0.142132 -0.384018 -0.175676
                                         0.053566 -0.008753
                                                             1.000000
                                                                       0.002849
    xedge
    xedgey -0.084820 -0.052545 0.029419
                                         0.527239 -0.184927
                                                             0.002849
                                                                       1.000000
            0.006546 0.277540 -0.087019 -0.226251
                                                   0.049695
                                                             0.108411 -0.064402
    yedge
    yedgex 0.182902 -0.061335 -0.114223 -0.236518 0.245808 -0.049789 -0.187591
               yedge
                        yedgex
    xbox
            0.273504 -0.105147
            0.230883 -0.042741
    ybox
    width
            0.260285 -0.118273
    height
            0.297545 -0.018853
    onpix
            0.492653 -0.062969
    xbar
            0.127056 0.248816
    ybar
           -0.078008 -0.207900
            0.006546 0.182902
    x2bar
    y2bar
            0.277540 -0.061335
           -0.087019 -0.114223
    xybar
    x2ybar -0.226251 -0.236518
    xy2bar 0.049695 0.245808
    xedge
            0.108411 -0.049789
    xedgey -0.064402 -0.187591
    yedge
            1.000000
                      0.143588
    yedgex 0.143588
                      1.000000
[8]: data.isnull().sum()
[8]: letter
              0
              0
    xbox
```

0

ybox

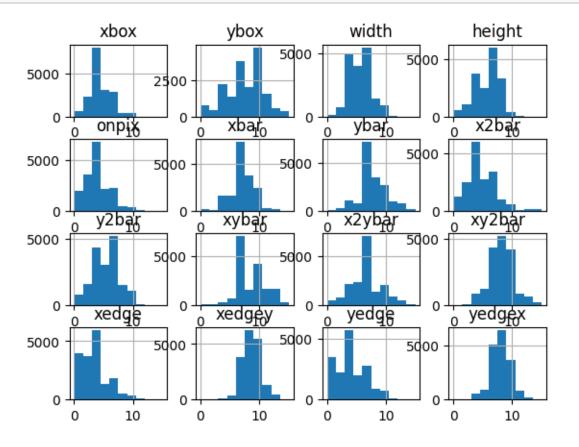
```
width
     height
                0
      onpix
                0
      xbar
                0
     ybar
                0
      x2bar
                0
     y2bar
                0
                0
      xybar
      x2ybar
                0
      xy2bar
                0
      xedge
                0
      xedgey
                0
      yedge
                0
      yedgex
                0
      dtype: int64
 [9]: duplicates = data.duplicated()
      duplicates.value_counts()
 [9]: False
              18668
     True
                1332
      Name: count, dtype: int64
[11]: data1 = data.drop_duplicates(keep=False)
[12]: duplicates = data1.duplicated()
      duplicates.value_counts()
[12]: False
              17823
      Name: count, dtype: int64
[13]: data1.info()
     <class 'pandas.core.frame.DataFrame'>
     Index: 17823 entries, 0 to 19999
     Data columns (total 17 columns):
          Column Non-Null Count Dtype
                  _____
      0
          letter 17823 non-null object
                  17823 non-null int64
      1
          xbox
      2
          ybox
                  17823 non-null int64
      3
          width
                  17823 non-null int64
      4
          height 17823 non-null int64
      5
                  17823 non-null int64
          onpix
          xbar
      6
                  17823 non-null int64
          ybar
                  17823 non-null int64
          x2bar
                  17823 non-null int64
```

0

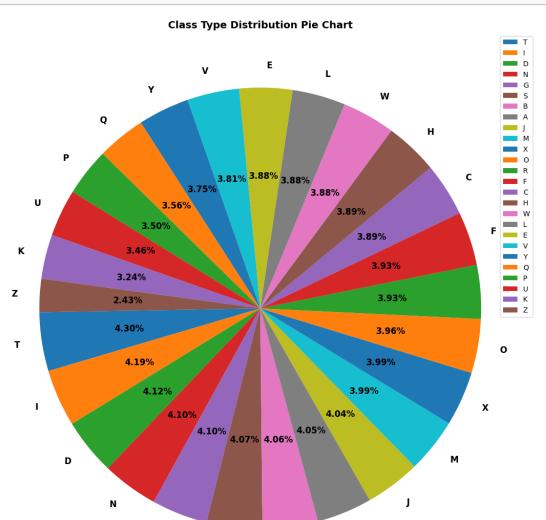
```
y2bar
            17823 non-null int64
 9
 10
    xybar
            17823 non-null int64
    x2ybar 17823 non-null int64
 11
12 xy2bar
            17823 non-null int64
    xedge
            17823 non-null int64
 13
 14
    xedgey
            17823 non-null int64
            17823 non-null int64
 15
    yedge
    yedgex 17823 non-null
                           int64
dtypes: int64(16), object(1)
memory usage: 2.4+ MB
```

[14]: data2= data1.copy()

[15]: data2.hist() plt.rcParams.update({'figure.figsize':(25,12), 'figure.dpi':100})



plt.title("Class Type Distribution Pie Chart", fontsize=14, fontweight='bold')
plt.show()

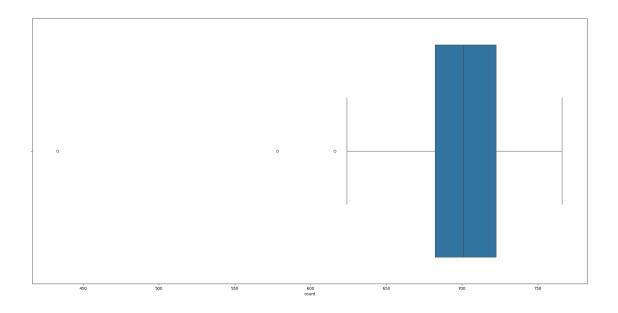


В



s

G





```
[21]: from sklearn import preprocessing
      # how to understand word labels.
      label_encoder = preprocessing.LabelEncoder()
      # Encode labels in column 'letters'.
      data1['letter'] = label_encoder.fit_transform(data1['letter'])
      data1['letter'].unique()
     <ipython-input-21-8732cd2a233d>:7: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-
     docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
       data1['letter'] = label_encoder.fit_transform(data1['letter'])
[21]: array([19, 8, 3, 13, 6, 18, 1, 0, 9, 12, 23, 14, 17, 5, 2, 7, 22,
             11, 4, 21, 24, 16, 15, 20, 10, 25])
[22]: x = data1.drop('letter',
                     axis = 1)
      y = data1['letter']
[23]: x
```

```
[23]:
                     ybox width height onpix xbar
                                                                  x2bar
                                                                          y2bar
              xbox
                                                            ybar
                                                                                   xybar \
      0
                  2
                        8
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                                                                        0
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      1
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                        12
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                        11
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      3
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                        11
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      4
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      19994
                  5
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                                                                        4
      19995
                  2
                         2
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      19996
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      19997
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      19999
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              x2ybar
                        xy2bar
                                 xedge
                                         xedgey
                                                         yedgex
                                                  yedge
      0
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                    3
      1
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                                     2
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                                                               10
      2
                    3
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      3
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                    7
      19994
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      19995
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                                                       3
      19996
                    9
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                                                                7
                            13
                                                       2
      19997
                    9
                             5
                                     2
                                              12
                                                                4
      19999
                    1
                             8
                                     2
                                              7
                                                       2
                                                                8
      [17823 rows x 16 columns]
[24]: y
[24]: 0
                 19
                  8
      1
                  3
      2
      3
                 13
                  6
                 . .
      19994
                 19
      19995
                  3
      19996
                  2
      19997
                 19
      19999
                  0
      Name: letter, Length: 17823, dtype: int64
[25]: from sklearn.model_selection import train_test_split
      x_train, x_test, y_train, y_test = train_test_split(x, y, test_size = 0.20)
```

```
[26]: model = Sequential()
      model.add(Dense(128, input_dim=16, activation='relu'))
      model.add(Dense(32, activation='relu'))
      model.add(Dense(26, activation='softmax'))
     /usr/local/lib/python3.10/dist-packages/keras/src/layers/core/dense.py:87:
     UserWarning: Do not pass an `input_shape`/`input_dim` argument to a layer. When
     using Sequential models, prefer using an `Input(shape)` object as the first
     layer in the model instead.
       super().__init__(activity_regularizer=activity_regularizer, **kwargs)
[27]: model.compile(loss='sparse categorical crossentropy', optimizer='adam', |
       →metrics=['accuracy'])
[28]: history=model.fit(x_train,y_train, validation_split=0.33, epochs=100,_u
       ⇔batch_size= 128 )
     Epoch 1/100
     75/75
                       2s 5ms/step -
     accuracy: 0.0886 - loss: 3.4697 - val_accuracy: 0.2346 - val_loss: 2.6565
     Epoch 2/100
     75/75
                       Os 3ms/step -
     accuracy: 0.3198 - loss: 2.3936 - val_accuracy: 0.5161 - val_loss: 1.8203
     Epoch 3/100
     75/75
                       Os 3ms/step -
     accuracy: 0.5515 - loss: 1.6615 - val_accuracy: 0.5986 - val_loss: 1.4521
     Epoch 4/100
     75/75
                       Os 3ms/step -
     accuracy: 0.6300 - loss: 1.3609 - val_accuracy: 0.6458 - val_loss: 1.2812
     Epoch 5/100
     75/75
                       Os 3ms/step -
     accuracy: 0.6701 - loss: 1.2203 - val_accuracy: 0.6864 - val_loss: 1.1562
     Epoch 6/100
     75/75
                       Os 3ms/step -
     accuracy: 0.6986 - loss: 1.1146 - val_accuracy: 0.7133 - val_loss: 1.0689
     Epoch 7/100
     75/75
                       Os 3ms/step -
     accuracy: 0.7251 - loss: 1.0096 - val_accuracy: 0.7159 - val_loss: 1.0166
     Epoch 8/100
     75/75
                       Os 3ms/step -
     accuracy: 0.7316 - loss: 0.9584 - val_accuracy: 0.7331 - val_loss: 0.9413
     Epoch 9/100
     75/75
                       Os 3ms/step -
     accuracy: 0.7577 - loss: 0.8926 - val_accuracy: 0.7418 - val_loss: 0.9150
     Epoch 10/100
                       Os 3ms/step -
     accuracy: 0.7614 - loss: 0.8408 - val_accuracy: 0.7486 - val_loss: 0.8731
     Epoch 11/100
```

```
75/75
                 Os 3ms/step -
accuracy: 0.7702 - loss: 0.8100 - val_accuracy: 0.7563 - val_loss: 0.8342
Epoch 12/100
75/75
                 Os 3ms/step -
accuracy: 0.7737 - loss: 0.7956 - val accuracy: 0.7743 - val loss: 0.8000
Epoch 13/100
75/75
                 Os 3ms/step -
accuracy: 0.7878 - loss: 0.7568 - val_accuracy: 0.7818 - val_loss: 0.7813
Epoch 14/100
75/75
                 Os 3ms/step -
accuracy: 0.7883 - loss: 0.7413 - val accuracy: 0.7860 - val loss: 0.7465
Epoch 15/100
75/75
                 Os 3ms/step -
accuracy: 0.7967 - loss: 0.7137 - val_accuracy: 0.7879 - val_loss: 0.7367
Epoch 16/100
75/75
                 Os 3ms/step -
accuracy: 0.8083 - loss: 0.6776 - val_accuracy: 0.7928 - val_loss: 0.7201
Epoch 17/100
75/75
                 Os 3ms/step -
accuracy: 0.8096 - loss: 0.6640 - val_accuracy: 0.7952 - val_loss: 0.6963
Epoch 18/100
75/75
                 Os 3ms/step -
accuracy: 0.8226 - loss: 0.6442 - val_accuracy: 0.8130 - val_loss: 0.6621
Epoch 19/100
75/75
                 Os 3ms/step -
accuracy: 0.8229 - loss: 0.6280 - val accuracy: 0.8098 - val loss: 0.6614
Epoch 20/100
75/75
                 Os 3ms/step -
accuracy: 0.8177 - loss: 0.6321 - val_accuracy: 0.8124 - val_loss: 0.6480
Epoch 21/100
75/75
                 0s 4ms/step -
accuracy: 0.8333 - loss: 0.6004 - val_accuracy: 0.8147 - val_loss: 0.6303
Epoch 22/100
75/75
                 Os 3ms/step -
accuracy: 0.8333 - loss: 0.5822 - val accuracy: 0.8185 - val loss: 0.6339
Epoch 23/100
75/75
                 1s 6ms/step -
accuracy: 0.8365 - loss: 0.5645 - val_accuracy: 0.8179 - val_loss: 0.6197
Epoch 24/100
                 0s 4ms/step -
75/75
accuracy: 0.8383 - loss: 0.5664 - val_accuracy: 0.8255 - val_loss: 0.5976
Epoch 25/100
75/75
                 Os 5ms/step -
accuracy: 0.8486 - loss: 0.5378 - val_accuracy: 0.8351 - val_loss: 0.5807
Epoch 26/100
                 1s 8ms/step -
accuracy: 0.8468 - loss: 0.5273 - val_accuracy: 0.8264 - val_loss: 0.5953
Epoch 27/100
```

```
75/75
                 1s 6ms/step -
accuracy: 0.8484 - loss: 0.5256 - val_accuracy: 0.8319 - val_loss: 0.5914
Epoch 28/100
75/75
                 1s 6ms/step -
accuracy: 0.8485 - loss: 0.5208 - val accuracy: 0.8383 - val loss: 0.5488
Epoch 29/100
75/75
                 1s 6ms/step -
accuracy: 0.8497 - loss: 0.5077 - val_accuracy: 0.8455 - val_loss: 0.5555
Epoch 30/100
75/75
                 Os 5ms/step -
accuracy: 0.8602 - loss: 0.4932 - val_accuracy: 0.8389 - val_loss: 0.5591
Epoch 31/100
75/75
                 Os 3ms/step -
accuracy: 0.8579 - loss: 0.4871 - val_accuracy: 0.8430 - val_loss: 0.5330
Epoch 32/100
75/75
                 Os 3ms/step -
accuracy: 0.8611 - loss: 0.4761 - val_accuracy: 0.8513 - val_loss: 0.5082
Epoch 33/100
75/75
                 Os 3ms/step -
accuracy: 0.8567 - loss: 0.4778 - val_accuracy: 0.8479 - val_loss: 0.5130
Epoch 34/100
75/75
                 Os 3ms/step -
accuracy: 0.8731 - loss: 0.4421 - val_accuracy: 0.8483 - val_loss: 0.5012
Epoch 35/100
75/75
                 Os 3ms/step -
accuracy: 0.8713 - loss: 0.4596 - val accuracy: 0.8587 - val loss: 0.4872
Epoch 36/100
75/75
                 Os 3ms/step -
accuracy: 0.8734 - loss: 0.4333 - val_accuracy: 0.8538 - val_loss: 0.4844
Epoch 37/100
                 Os 3ms/step -
75/75
accuracy: 0.8724 - loss: 0.4272 - val_accuracy: 0.8498 - val_loss: 0.4945
Epoch 38/100
75/75
                 Os 3ms/step -
accuracy: 0.8773 - loss: 0.4283 - val accuracy: 0.8551 - val loss: 0.4923
Epoch 39/100
75/75
                 Os 3ms/step -
accuracy: 0.8711 - loss: 0.4355 - val_accuracy: 0.8561 - val_loss: 0.4776
Epoch 40/100
                 0s 3ms/step -
75/75
accuracy: 0.8795 - loss: 0.3983 - val_accuracy: 0.8627 - val_loss: 0.4581
Epoch 41/100
75/75
                 Os 3ms/step -
accuracy: 0.8839 - loss: 0.3914 - val_accuracy: 0.8657 - val_loss: 0.4475
Epoch 42/100
                 Os 3ms/step -
accuracy: 0.8878 - loss: 0.3816 - val_accuracy: 0.8638 - val_loss: 0.4657
Epoch 43/100
```

```
75/75
                 Os 3ms/step -
accuracy: 0.8863 - loss: 0.3919 - val_accuracy: 0.8685 - val_loss: 0.4388
Epoch 44/100
75/75
                 Os 3ms/step -
accuracy: 0.8883 - loss: 0.3815 - val accuracy: 0.8691 - val loss: 0.4423
Epoch 45/100
75/75
                 Os 3ms/step -
accuracy: 0.8881 - loss: 0.3756 - val_accuracy: 0.8672 - val_loss: 0.4406
Epoch 46/100
75/75
                 Os 3ms/step -
accuracy: 0.9012 - loss: 0.3639 - val accuracy: 0.8693 - val loss: 0.4424
Epoch 47/100
75/75
                 Os 3ms/step -
accuracy: 0.8917 - loss: 0.3711 - val_accuracy: 0.8719 - val_loss: 0.4227
Epoch 48/100
75/75
                 Os 4ms/step -
accuracy: 0.8932 - loss: 0.3627 - val_accuracy: 0.8674 - val_loss: 0.4329
Epoch 49/100
75/75
                 Os 4ms/step -
accuracy: 0.8879 - loss: 0.3696 - val_accuracy: 0.8787 - val_loss: 0.4084
Epoch 50/100
75/75
                 1s 4ms/step -
accuracy: 0.8937 - loss: 0.3514 - val_accuracy: 0.8759 - val_loss: 0.4096
Epoch 51/100
75/75
                 1s 6ms/step -
accuracy: 0.8965 - loss: 0.3441 - val accuracy: 0.8780 - val loss: 0.4083
Epoch 52/100
75/75
                 1s 4ms/step -
accuracy: 0.8987 - loss: 0.3457 - val_accuracy: 0.8729 - val_loss: 0.4286
Epoch 53/100
                 Os 3ms/step -
75/75
accuracy: 0.8981 - loss: 0.3345 - val_accuracy: 0.8802 - val_loss: 0.3912
Epoch 54/100
75/75
                 Os 3ms/step -
accuracy: 0.9027 - loss: 0.3292 - val accuracy: 0.8789 - val loss: 0.3945
Epoch 55/100
                 0s 4ms/step -
accuracy: 0.9015 - loss: 0.3247 - val_accuracy: 0.8787 - val_loss: 0.3927
Epoch 56/100
75/75
                 Os 3ms/step -
accuracy: 0.9131 - loss: 0.3026 - val_accuracy: 0.8808 - val_loss: 0.3968
Epoch 57/100
75/75
                 1s 7ms/step -
accuracy: 0.9045 - loss: 0.3263 - val_accuracy: 0.8789 - val_loss: 0.3871
Epoch 58/100
                 Os 3ms/step -
accuracy: 0.9079 - loss: 0.3063 - val_accuracy: 0.8816 - val_loss: 0.3862
Epoch 59/100
```

```
75/75
                 1s 8ms/step -
accuracy: 0.9133 - loss: 0.3127 - val_accuracy: 0.8840 - val_loss: 0.3808
Epoch 60/100
75/75
                 1s 7ms/step -
accuracy: 0.9074 - loss: 0.3120 - val accuracy: 0.8780 - val loss: 0.3923
Epoch 61/100
75/75
                 Os 5ms/step -
accuracy: 0.9098 - loss: 0.2943 - val_accuracy: 0.8859 - val_loss: 0.3725
Epoch 62/100
75/75
                 1s 6ms/step -
accuracy: 0.9110 - loss: 0.3058 - val accuracy: 0.8821 - val loss: 0.3783
Epoch 63/100
75/75
                 Os 4ms/step -
accuracy: 0.9083 - loss: 0.3009 - val_accuracy: 0.8901 - val_loss: 0.3716
Epoch 64/100
75/75
                 1s 5ms/step -
accuracy: 0.9071 - loss: 0.2981 - val_accuracy: 0.8802 - val_loss: 0.3799
Epoch 65/100
75/75
                 1s 6ms/step -
accuracy: 0.9145 - loss: 0.2974 - val accuracy: 0.8948 - val loss: 0.3610
Epoch 66/100
75/75
                 1s 5ms/step -
accuracy: 0.9167 - loss: 0.2742 - val_accuracy: 0.8829 - val_loss: 0.3776
Epoch 67/100
75/75
                 1s 7ms/step -
accuracy: 0.9129 - loss: 0.2835 - val accuracy: 0.8853 - val loss: 0.3650
Epoch 68/100
75/75
                 1s 7ms/step -
accuracy: 0.9193 - loss: 0.2740 - val_accuracy: 0.8874 - val_loss: 0.3698
Epoch 69/100
75/75
                 1s 9ms/step -
accuracy: 0.9166 - loss: 0.2807 - val_accuracy: 0.8838 - val_loss: 0.3697
Epoch 70/100
75/75
                 1s 7ms/step -
accuracy: 0.9218 - loss: 0.2618 - val accuracy: 0.8889 - val loss: 0.3585
Epoch 71/100
75/75
                 1s 7ms/step -
accuracy: 0.9189 - loss: 0.2684 - val_accuracy: 0.8921 - val_loss: 0.3518
Epoch 72/100
                 1s 8ms/step -
75/75
accuracy: 0.9237 - loss: 0.2625 - val_accuracy: 0.8933 - val_loss: 0.3489
Epoch 73/100
75/75
                 1s 6ms/step -
accuracy: 0.9248 - loss: 0.2605 - val_accuracy: 0.8878 - val_loss: 0.3730
Epoch 74/100
                 Os 3ms/step -
accuracy: 0.9185 - loss: 0.2590 - val_accuracy: 0.8867 - val_loss: 0.3715
Epoch 75/100
```

```
75/75
                 Os 3ms/step -
accuracy: 0.9138 - loss: 0.2706 - val_accuracy: 0.8999 - val_loss: 0.3417
Epoch 76/100
75/75
                 Os 3ms/step -
accuracy: 0.9212 - loss: 0.2603 - val accuracy: 0.8929 - val loss: 0.3472
Epoch 77/100
75/75
                 Os 3ms/step -
accuracy: 0.9271 - loss: 0.2430 - val_accuracy: 0.8965 - val_loss: 0.3397
Epoch 78/100
75/75
                 Os 3ms/step -
accuracy: 0.9241 - loss: 0.2471 - val accuracy: 0.8946 - val loss: 0.3380
Epoch 79/100
75/75
                 Os 5ms/step -
accuracy: 0.9267 - loss: 0.2408 - val_accuracy: 0.8931 - val_loss: 0.3402
Epoch 80/100
75/75
                 1s 10ms/step -
accuracy: 0.9288 - loss: 0.2335 - val_accuracy: 0.8893 - val_loss: 0.3636
Epoch 81/100
75/75
                 1s 9ms/step -
accuracy: 0.9320 - loss: 0.2357 - val accuracy: 0.8957 - val loss: 0.3510
Epoch 82/100
75/75
                 1s 8ms/step -
accuracy: 0.9231 - loss: 0.2485 - val_accuracy: 0.8942 - val_loss: 0.3352
Epoch 83/100
75/75
                 2s 11ms/step -
accuracy: 0.9321 - loss: 0.2359 - val_accuracy: 0.8984 - val_loss: 0.3327
Epoch 84/100
75/75
                 1s 10ms/step -
accuracy: 0.9331 - loss: 0.2262 - val_accuracy: 0.8923 - val_loss: 0.3415
Epoch 85/100
75/75
                 1s 7ms/step -
accuracy: 0.9297 - loss: 0.2377 - val_accuracy: 0.8959 - val_loss: 0.3309
Epoch 86/100
75/75
                 1s 6ms/step -
accuracy: 0.9229 - loss: 0.2477 - val accuracy: 0.8974 - val loss: 0.3276
Epoch 87/100
75/75
                 0s 4ms/step -
accuracy: 0.9289 - loss: 0.2298 - val_accuracy: 0.8963 - val_loss: 0.3429
Epoch 88/100
                 0s 3ms/step -
75/75
accuracy: 0.9329 - loss: 0.2254 - val_accuracy: 0.9014 - val_loss: 0.3169
Epoch 89/100
75/75
                 Os 3ms/step -
accuracy: 0.9314 - loss: 0.2232 - val_accuracy: 0.8950 - val_loss: 0.3286
Epoch 90/100
                 Os 3ms/step -
accuracy: 0.9336 - loss: 0.2199 - val_accuracy: 0.8978 - val_loss: 0.3258
Epoch 91/100
```

```
accuracy: 0.9288 - loss: 0.2275 - val_accuracy: 0.8978 - val_loss: 0.3372
     Epoch 92/100
     75/75
                       Os 3ms/step -
     accuracy: 0.9335 - loss: 0.2218 - val accuracy: 0.9012 - val loss: 0.3204
     Epoch 93/100
     75/75
                       Os 3ms/step -
     accuracy: 0.9394 - loss: 0.2098 - val_accuracy: 0.9003 - val_loss: 0.3193
     Epoch 94/100
     75/75
                       Os 3ms/step -
     accuracy: 0.9348 - loss: 0.2258 - val accuracy: 0.8931 - val loss: 0.3345
     Epoch 95/100
     75/75
                       Os 3ms/step -
     accuracy: 0.9304 - loss: 0.2197 - val_accuracy: 0.8980 - val_loss: 0.3285
     Epoch 96/100
     75/75
                       Os 3ms/step -
     accuracy: 0.9315 - loss: 0.2169 - val_accuracy: 0.8995 - val_loss: 0.3248
     Epoch 97/100
     75/75
                       0s 3ms/step -
     accuracy: 0.9313 - loss: 0.2167 - val_accuracy: 0.9086 - val_loss: 0.3098
     Epoch 98/100
     75/75
                       Os 4ms/step -
     accuracy: 0.9363 - loss: 0.2087 - val_accuracy: 0.8959 - val_loss: 0.3266
     Epoch 99/100
     75/75
                       Os 3ms/step -
     accuracy: 0.9343 - loss: 0.2144 - val accuracy: 0.9065 - val loss: 0.2998
     Epoch 100/100
     75/75
                       Os 3ms/step -
     accuracy: 0.9412 - loss: 0.1941 - val_accuracy: 0.8980 - val_loss: 0.3212
[29]: scores = model.evaluate(x, y)
     print("%s: %.2f%%" % (model.metrics_names[1], scores[1]*100))
     557/557
                         1s 1ms/step -
     accuracy: 0.9186 - loss: 0.2621
     compile_metrics: 91.84%
[30]: from sklearn.metrics import classification_report, confusion_matrix
      # Define expected unique labels (based on your previous output)
      expected_labels = ['T', 'I', 'D', 'N', 'G', 'S', 'B', 'A', 'J', 'M', 'X', 'O', _
       □ 'R', 'F', 'C', 'H', 'W', 'L', 'P', 'E', 'V', 'Y', 'Q', 'U', 'K', 'Z']
      # Evaluate the model
      y_pred_prob = model.predict(x_test)
      y_pred_encoded = np.argmax(y_pred_prob, axis=1)
```

75/75

Os 3ms/step -

```
# Convert predicted labels back to original letters using inverse_transform
y_test_pred = label_encoder.inverse_transform(y_test)
y_pred_letters = label_encoder.inverse_transform(y_pred_encoded)
```

Classification Report:

	precision	recall	f1-score	support
T	0.94	0.95	0.94	140
I	0.93	0.89	0.91	129
D	0.94	0.87	0.90	138
N	0.90	0.79	0.84	142
G	0.78	0.83	0.80	129
S	0.95	0.85	0.90	149
В	0.75	0.86	0.80	125
A	0.81	0.83	0.82	152
J	0.90	0.84	0.87	73
M	0.88	0.96	0.92	114
X	0.90	0.92	0.91	132
0	0.87	0.94	0.91	127
R	0.85	0.98	0.91	135
F	0.96	0.88	0.92	133
C	0.88	0.86	0.87	164
Н	0.94	0.90	0.92	157
W	0.94	0.95	0.94	173
L	0.89	0.92	0.90	143
Р	0.87	0.92	0.90	157
E	0.92	0.96	0.94	129
V	0.87	0.95	0.91	153
Y	0.95	0.96	0.95	128
Q	0.97	0.93	0.95	147
U	0.90	0.89	0.90	117
K	0.98	0.92	0.95	142
Z	0.97	0.81	0.88	137
accuracy			0.90	3565
macro avg	0.90	0.90	0.90	3565
weighted avg	0.90	0.90	0.90	3565

```
[32]: model = Sequential()
      model.add(Dense(128, activation='relu'))
      model.add(Dense(64, activation='relu'))
      model.add(Dense(32, activation='relu'))
      model.add(Dense(len(label_encoder.classes_), activation='softmax'))
      # Compile the model
      model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', __
       ⇔metrics=['accuracy'])
      # Train the model
      model.fit(x_train,y_train, epochs=50, batch_size=32, validation_split=0.2)
     Epoch 1/50
     357/357
                         6s 11ms/step -
     accuracy: 0.2466 - loss: 2.7189 - val_accuracy: 0.6280 - val_loss: 1.3616
     Epoch 2/50
     357/357
                         3s 5ms/step -
     accuracy: 0.6493 - loss: 1.2684 - val_accuracy: 0.6701 - val_loss: 1.1147
     Epoch 3/50
     357/357
                         3s 5ms/step -
     accuracy: 0.7172 - loss: 0.9907 - val_accuracy: 0.7051 - val_loss: 0.9362
     Epoch 4/50
     357/357
                         1s 2ms/step -
     accuracy: 0.7569 - loss: 0.8457 - val_accuracy: 0.7542 - val_loss: 0.7985
     Epoch 5/50
     357/357
                         1s 2ms/step -
     accuracy: 0.7752 - loss: 0.7418 - val_accuracy: 0.7700 - val_loss: 0.7231
     Epoch 6/50
     357/357
                         1s 2ms/step -
     accuracy: 0.7994 - loss: 0.6789 - val_accuracy: 0.8040 - val_loss: 0.6397
     Epoch 7/50
     357/357
                         1s 2ms/step -
     accuracy: 0.8130 - loss: 0.6095 - val accuracy: 0.8012 - val loss: 0.6300
     Epoch 8/50
     357/357
                         1s 2ms/step -
     accuracy: 0.8358 - loss: 0.5385 - val_accuracy: 0.8170 - val_loss: 0.5705
     Epoch 9/50
     357/357
                         1s 4ms/step -
     accuracy: 0.8421 - loss: 0.5145 - val_accuracy: 0.8215 - val_loss: 0.5574
     Epoch 10/50
     357/357
                         1s 4ms/step -
     accuracy: 0.8514 - loss: 0.4645 - val_accuracy: 0.8415 - val_loss: 0.5198
     Epoch 11/50
     357/357
                         1s 3ms/step -
     accuracy: 0.8662 - loss: 0.4455 - val_accuracy: 0.8538 - val_loss: 0.4617
     Epoch 12/50
```

```
357/357
                   1s 2ms/step -
accuracy: 0.8693 - loss: 0.4196 - val_accuracy: 0.8520 - val_loss: 0.4578
Epoch 13/50
357/357
                   1s 2ms/step -
accuracy: 0.8823 - loss: 0.3794 - val_accuracy: 0.8661 - val_loss: 0.4232
Epoch 14/50
357/357
                   1s 2ms/step -
accuracy: 0.8823 - loss: 0.3777 - val_accuracy: 0.8640 - val_loss: 0.4093
Epoch 15/50
357/357
                   1s 2ms/step -
accuracy: 0.8916 - loss: 0.3568 - val accuracy: 0.8657 - val loss: 0.4245
Epoch 16/50
357/357
                    1s 2ms/step -
accuracy: 0.9006 - loss: 0.3187 - val_accuracy: 0.8836 - val_loss: 0.3704
Epoch 17/50
357/357
                   1s 2ms/step -
accuracy: 0.9082 - loss: 0.2990 - val_accuracy: 0.8692 - val_loss: 0.3937
Epoch 18/50
357/357
                   1s 2ms/step -
accuracy: 0.9101 - loss: 0.2907 - val_accuracy: 0.8917 - val_loss: 0.3334
Epoch 19/50
357/357
                   1s 2ms/step -
accuracy: 0.9152 - loss: 0.2615 - val_accuracy: 0.8948 - val_loss: 0.3323
Epoch 20/50
357/357
                   1s 2ms/step -
accuracy: 0.9145 - loss: 0.2643 - val accuracy: 0.8899 - val loss: 0.3492
Epoch 21/50
357/357
                   1s 4ms/step -
accuracy: 0.9174 - loss: 0.2522 - val_accuracy: 0.8734 - val_loss: 0.3618
Epoch 22/50
357/357
                   1s 4ms/step -
accuracy: 0.9248 - loss: 0.2409 - val_accuracy: 0.8973 - val_loss: 0.3215
Epoch 23/50
357/357
                   2s 2ms/step -
accuracy: 0.9264 - loss: 0.2268 - val accuracy: 0.8994 - val loss: 0.3065
Epoch 24/50
                   1s 2ms/step -
accuracy: 0.9253 - loss: 0.2311 - val_accuracy: 0.8931 - val_loss: 0.3156
Epoch 25/50
357/357
                   1s 2ms/step -
accuracy: 0.9292 - loss: 0.2183 - val_accuracy: 0.8938 - val_loss: 0.3359
Epoch 26/50
357/357
                   1s 2ms/step -
accuracy: 0.9342 - loss: 0.2042 - val_accuracy: 0.9071 - val_loss: 0.2782
Epoch 27/50
                   1s 2ms/step -
357/357
accuracy: 0.9270 - loss: 0.2091 - val_accuracy: 0.8941 - val_loss: 0.3263
Epoch 28/50
```

```
357/357
                   1s 2ms/step -
accuracy: 0.9285 - loss: 0.2171 - val_accuracy: 0.9165 - val_loss: 0.2855
Epoch 29/50
357/357
                   1s 2ms/step -
accuracy: 0.9320 - loss: 0.1989 - val_accuracy: 0.9092 - val_loss: 0.2712
Epoch 30/50
357/357
                   1s 2ms/step -
accuracy: 0.9391 - loss: 0.1855 - val_accuracy: 0.9022 - val_loss: 0.3023
Epoch 31/50
357/357
                   1s 2ms/step -
accuracy: 0.9357 - loss: 0.1888 - val accuracy: 0.9148 - val loss: 0.2608
Epoch 32/50
357/357
                   2s 4ms/step -
accuracy: 0.9409 - loss: 0.1766 - val_accuracy: 0.9109 - val_loss: 0.2763
Epoch 33/50
357/357
                   2s 2ms/step -
accuracy: 0.9454 - loss: 0.1713 - val_accuracy: 0.9225 - val_loss: 0.2364
Epoch 34/50
357/357
                   1s 2ms/step -
accuracy: 0.9492 - loss: 0.1568 - val_accuracy: 0.9102 - val_loss: 0.2730
Epoch 35/50
357/357
                   1s 2ms/step -
accuracy: 0.9399 - loss: 0.1746 - val_accuracy: 0.9158 - val_loss: 0.2441
Epoch 36/50
357/357
                   1s 2ms/step -
accuracy: 0.9460 - loss: 0.1537 - val accuracy: 0.9043 - val loss: 0.2726
Epoch 37/50
357/357
                   1s 2ms/step -
accuracy: 0.9441 - loss: 0.1593 - val_accuracy: 0.9194 - val_loss: 0.2503
Epoch 38/50
357/357
                   1s 2ms/step -
accuracy: 0.9512 - loss: 0.1476 - val_accuracy: 0.9208 - val_loss: 0.2303
Epoch 39/50
357/357
                   1s 2ms/step -
accuracy: 0.9464 - loss: 0.1576 - val accuracy: 0.9116 - val loss: 0.2924
Epoch 40/50
                   1s 2ms/step -
accuracy: 0.9491 - loss: 0.1457 - val_accuracy: 0.9183 - val_loss: 0.2478
Epoch 41/50
357/357
                   1s 2ms/step -
accuracy: 0.9567 - loss: 0.1385 - val_accuracy: 0.9134 - val_loss: 0.2493
Epoch 42/50
357/357
                   2s 4ms/step -
accuracy: 0.9491 - loss: 0.1527 - val_accuracy: 0.9197 - val_loss: 0.2439
Epoch 43/50
                   2s 2ms/step -
357/357
accuracy: 0.9506 - loss: 0.1381 - val_accuracy: 0.9194 - val_loss: 0.2358
Epoch 44/50
```

```
357/357
                         1s 2ms/step -
     accuracy: 0.9541 - loss: 0.1300 - val_accuracy: 0.9225 - val_loss: 0.2430
     Epoch 45/50
     357/357
                         1s 2ms/step -
     accuracy: 0.9512 - loss: 0.1426 - val_accuracy: 0.9274 - val_loss: 0.2199
     Epoch 46/50
                         1s 2ms/step -
     357/357
     accuracy: 0.9572 - loss: 0.1240 - val_accuracy: 0.9316 - val_loss: 0.2266
     Epoch 47/50
     357/357
                         1s 2ms/step -
     accuracy: 0.9521 - loss: 0.1349 - val accuracy: 0.9173 - val loss: 0.2578
     Epoch 48/50
     357/357
                         1s 2ms/step -
     accuracy: 0.9538 - loss: 0.1399 - val accuracy: 0.9218 - val loss: 0.2358
     Epoch 49/50
     357/357
                         1s 2ms/step -
     accuracy: 0.9584 - loss: 0.1193 - val_accuracy: 0.9243 - val_loss: 0.2434
     Epoch 50/50
     357/357
                         1s 2ms/step -
     accuracy: 0.9612 - loss: 0.1131 - val_accuracy: 0.9197 - val_loss: 0.2600
[32]: <keras.src.callbacks.history.History at 0x7be622632e00>
[33]: # Define expected unique labels (based on your previous output)
      expected_labels = ['T', 'I', 'D', 'N', 'G', 'S', 'B', 'A', 'J', 'M', 'X', 'O', _
       ⇔'R', 'F', 'C', 'H', 'W', 'L', 'P', 'E', 'V', 'Y', 'Q', 'U', 'K', 'Z']
      # Evaluate the model
      y_pred_prob = model.predict(x_test)
      y_pred_encoded = np.argmax(y_pred_prob, axis=1)
      # Convert predicted labels back to original letters using inverse transform
      y_test_pred = label_encoder.inverse_transform(y_test)
      y_pred_letters = label_encoder.inverse_transform(y_pred_encoded)
     112/112
                         Os 2ms/step
 []:
 []: # Generate classification report
      report = classification_report(y_test_pred, y_pred_letters,_
       starget_names=expected_labels)
      # Print classification report
      print("Classification Report:")
      print(report)
```

[37]: !pip install scikeras

```
Collecting scikeras
  Downloading scikeras-0.13.0-py3-none-any.whl.metadata (3.1 kB)
Requirement already satisfied: keras>=3.2.0 in /usr/local/lib/python3.10/dist-
packages (from scikeras) (3.5.0)
Requirement already satisfied: scikit-learn>=1.4.2 in
/usr/local/lib/python3.10/dist-packages (from scikeras) (1.5.2)
Requirement already satisfied: absl-py in /usr/local/lib/python3.10/dist-
packages (from keras>=3.2.0->scikeras) (1.4.0)
Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages
(from keras >= 3.2.0 -> scikeras) (1.26.4)
Requirement already satisfied: rich in /usr/local/lib/python3.10/dist-packages
(from keras >= 3.2.0 -> scikeras) (13.9.4)
Requirement already satisfied: namex in /usr/local/lib/python3.10/dist-packages
(from keras >= 3.2.0 -> scikeras) (0.0.8)
Requirement already satisfied: h5py in /usr/local/lib/python3.10/dist-packages
(from keras>=3.2.0->scikeras) (3.12.1)
Requirement already satisfied: optree in /usr/local/lib/python3.10/dist-packages
(from keras>=3.2.0->scikeras) (0.13.1)
Requirement already satisfied: ml-dtypes in /usr/local/lib/python3.10/dist-
packages (from keras>=3.2.0->scikeras) (0.4.1)
Requirement already satisfied: packaging in /usr/local/lib/python3.10/dist-
packages (from keras>=3.2.0->scikeras) (24.2)
Requirement already satisfied: scipy>=1.6.0 in /usr/local/lib/python3.10/dist-
packages (from scikit-learn>=1.4.2->scikeras) (1.13.1)
Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.10/dist-
packages (from scikit-learn>=1.4.2->scikeras) (1.4.2)
Requirement already satisfied: threadpoolctl>=3.1.0 in
/usr/local/lib/python3.10/dist-packages (from scikit-learn>=1.4.2->scikeras)
(3.5.0)
Requirement already satisfied: typing-extensions>=4.5.0 in
/usr/local/lib/python3.10/dist-packages (from optree->keras>=3.2.0->scikeras)
(4.12.2)
Requirement already satisfied: markdown-it-py>=2.2.0 in
/usr/local/lib/python3.10/dist-packages (from rich->keras>=3.2.0->scikeras)
(3.0.0)
Requirement already satisfied: pygments<3.0.0,>=2.13.0 in
/usr/local/lib/python3.10/dist-packages (from rich->keras>=3.2.0->scikeras)
(2.18.0)
Requirement already satisfied: mdurl~=0.1 in /usr/local/lib/python3.10/dist-
packages (from markdown-it-py>=2.2.0->rich->keras>=3.2.0->scikeras) (0.1.2)
Downloading scikeras-0.13.0-py3-none-any.whl (26 kB)
Installing collected packages: scikeras
Successfully installed scikeras-0.13.0
```

```
[38]: #from tensorflow.keras.wrappers.scikit learn import KerasClassifier # Removeu
       ⇔this line
      from scikeras.wrappers import KerasClassifier # Import from scikeras instead
      from sklearn.metrics import accuracy score, make scorer
      from sklearn.model_selection import train_test_split, GridSearchCV
[40]: def create_model(optimizer='adam', activation='relu', neurons=64,__
       ⇔hidden_layers=1):
          model = Sequential()
          model.add(Dense(neurons, activation=activation, input_shape=(x_train.
       ⇔shape[1],)))
          model.add(Dense(32, activation='relu'))
          for _ in range(hidden_layers - 1):
              model.add(Dense(neurons, activation=activation))
          model.add(Dense(26, activation='softmax'))
          model.compile(optimizer=optimizer, loss='sparse_categorical_crossentropy', u
       →metrics=['accuracy'])
          return model
      # Create KerasClassifier based on the create_model function
      model = KerasClassifier(build_fn=create_model, verbose=0)
      # Define hyperparameters grid for GridSearchCV
      param_grid = {
          'model__optimizer': ['adam', 'rmsprop'],
          'model__activation': ['relu', 'tanh'],
          'model neurons': [32, 64, 128],
          'model_hidden_layers': [1, 2]
      }
      # Define accuracy as the scoring metric for GridSearchCV
      scoring = {'accuracy': make_scorer(accuracy_score)}
      # Perform GridSearchCV for hyperparameter tuning
      grid_search = GridSearchCV(estimator=model, param_grid=param_grid,__
       ⇔scoring=scoring, refit='accuracy', cv=3)
      grid_result = grid_search.fit(x_train, y_train)
```

```
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[41]: # Print best hyperparameters and corresponding accuracy
      print("Best Accuracy: {:.4f}".format(grid_result.best_score_))
      print("Best Parameters:", grid_result.best_params_)
      # Evaluate the best model on the test set
      best model = grid result.best estimator
      y_pred = best_model.predict(x_test)
      test_accuracy = accuracy_score(y_test, y_pred)
      print("Test Accuracy: {:.4f}".format(test_accuracy))
     Best Accuracy: 0.6725
     Best Parameters: {'model__activation': 'tanh', 'model__hidden_layers': 2,
     'model__neurons': 128, 'model__optimizer': 'adam'}
     Test Accuracy: 0.7086
[42]: print(classification_report(y_test, y_pred))
```

p:	recision	recall	f1-score	support
0	0.94	0.84	0.89	140
1	0.48	0.84	0.61	129

2	0.75	0.66	0.70	138
3	0.67	0.71	0.69	142
4	0.57	0.64	0.60	129
5	0.71	0.74	0.73	149
6	0.48	0.48	0.48	125
7	0.75	0.51	0.61	152
8	0.76	0.73	0.74	73
9	0.84	0.77	0.80	114
10	0.54	0.81	0.65	132
11	0.86	0.80	0.83	127
12	0.69	0.92	0.79	135
13	0.93	0.59	0.72	133
14	0.72	0.68	0.70	164
15	0.78	0.82	0.80	157
16	0.87	0.55	0.67	173
17	0.54	0.77	0.64	143
18	0.55	0.26	0.35	157
19	0.73	0.79	0.76	129
20	0.89	0.85	0.87	153
21	0.84	0.88	0.86	128
22	0.73	0.90	0.81	147
23	0.73	0.54	0.62	117
24	0.86	0.75	0.80	142
25	0.64	0.66	0.65	137
accuracy			0.71	3565
macro avg	0.73	0.71	0.71	3565
weighted avg	0.73	0.71	0.71	3565

[]: