atsiskaitymas

August 5, 2022

```
[1]: import os
     import matplotlib.pyplot as plt
     import pandas as pd
     import numpy as np
     import seaborn as sns
     from mpl_toolkits.mplot3d import axes3d
     import imageio
     import glob
     import itertools
     import tqdm
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.model_selection import train_test_split
     from sklearn.metrics import accuracy_score
     from sklearn import tree
     import tensorflow as tf
[2]: os.getcwd()
[2]: 'C:\\Users\\ariju\\Documents\\pamokos\\programavimas\\atsiskaitymas'
[3]: os.listdir()
[3]: ['.ipynb_checkpoints',
      '2022-05-24 18_50_25-Greenshot.png',
      'accuracy_entropy.jpg',
      'accuracy_mean_s_e.jpg',
      'Arijus Skaisgirys',
      'atsiskaitymas.ipynb',
      'backup',
      'bandymas.xlsx',
      'CarBuyers.csv',
      'Car_sales.csv',
      'cheapest_cars.png',
      'check.csv',
      'chromium',
      'data',
      'expensive_cars.png',
      'gifas.gif',
```

```
'kainu_pasiskirstymas.png',
      'loss_entropy.jpg',
      'loss_mean_s_e.jpg',
      'mfp',
      'mfp_naujas',
      'mobile_de',
      'most_popular_cars.png',
      'pasiskirstymasmfu.png',
      'patikrint.csv',
      'power_transmission.png',
      'price-predictor.dot',
      'price_power.png',
      'price_predicttion_tree_new.dot',
      'skelbimai',
      'usedCarsFinal.csv']
[4]: data = pd.read_csv('CarBuyers.csv')
[5]:
     data
[5]:
          Manufacturer
                            Model
                                               Transmission
                                                                   Power
                                        Price
     0
                  Ford
                            Focus
                                    30.619322
                                                   5.966102
                                                               94.033898
     1
                  Ford
                          Fiesta
                                    18.532143
                                                   5.714286
                                                               68.571429
     2
            Volkswagen
                             Golf
                                    31.242154
                                                   6.164835
                                                               89.461538
     3
                             Clio
               Renault
                                    22.100000
                                                   5.615385
                                                               75.576923
                                                              126.111111
     4
                   BMW
                             320i
                                    47.848370
                                                   6.44444
     6087
            Land-Rover Defender
                                   108.747195
                                                   7.853659
                                                              207.609756
     6088
                Toyota
                             RAV4
                                    43.548516
                                                   1.354839
                                                              137.774193
     6089
            Alfa-Romeo
                           Spider
                                    55.200000
                                                   6.000000
                                                              163.500000
     6090
                 Honda
                         Shuttle
                                    30.081000
                                                   4.000000
                                                              110.000000
     6091
            Mitsubishi
                            Space
                                    23.165158
                                                   3.947368
                                                               82.157895
             Engine CC
                           Fuel
                                   Male
                                         Female Unknown
                                                            Total
     0
           1497.169492
                        petrol
                                 814172
                                         422731
                                                 56,487
                                                          1293390
     1
           1166.142857
                                 554879
                                         631666
                                                 54,057
                        petrol
                                                          1240602
     2
           1537.406593
                        petrol
                                 483216
                                         310604
                                                 47,563
                                                           841383
     3
           1219.653846
                                 241287
                                         312556
                                                 28,004
                                                           581847
                        petrol
     4
           1995.777778
                        petrol
                                 408016
                                         115843
                                                 29,125
                                                           552984
                                    •••
     6087
           2304.975610
                        diesel
                                  1,012
                                            150
                                                      80
                                                            1,242
     6088
           2261.193548
                        petrol
                                    670
                                            482
                                                      66
                                                            1,218
     6089
           2696.500000
                        petrol
                                    790
                                            247
                                                      81
                                                            1,118
     6090 2254.000000
                                    639
                                            416
                                                      49
                                                            1,104
                        petrol
     6091 1817.315789 petrol
                                    721
                                            251
                                                      40
                                                            1,012
     [6092 rows x 11 columns]
```

```
[6]: data.dtypes
[6]: Manufacturer
                      object
    Model
                      object
    Price
                     float64
    Transmission
                     float64
    Power
                     float64
    Engine CC
                     float64
    Fuel
                      object
    Male
                      object
    Female
                      object
    Unknown
                      object
    Total
                      object
     dtype: object
[7]: class String_to_float:
         def __init__(self, y):
             self.y = y
         def comma_to_dot(self, x): # pakeicia , i . ir padaro floatu
             x = x.replace(',', '.')
             return float(x)
         def column_to_float(self): # lenteles stulpelio nerius pakeicia i float ,⊔
      ⇔jei yra kablelis skaiciui
             for i in self.y:
                 data[i] = data[i].map(self.comma_to_dot)
     string_to_float = String_to_float(['Male', 'Female', 'Unknown', 'Total'])
     string_to_float.column_to_float()
[8]: data.dtypes
[8]: Manufacturer
                      object
```

Model object Price float64 float64 Transmission Power float64 Engine CC float64 Fuel object Male float64 Female float64 Unknown float64 Total float64 dtype: object

[9]: data [9]: Manufacturer Model Transmission Power Price 0 Ford Focus 30.619322 5.966102 94.033898 1 Ford 18.532143 5.714286 Fiesta 68.571429 2 Volkswagen Golf 31.242154 6.164835 89.461538 Renault Clio 3 22.100000 5.615385 75.576923 4 320i BMW 47.848370 6.44444 126.111111 6087 Land-Rover Defender 108.747195 7.853659 207.609756 6088 Toyota RAV4 43.548516 1.354839 137.774193 Alfa-Romeo 6089 Spider 55.200000 6.000000 163.500000 6090 Shuttle Honda 30.081000 4.000000 110.000000 6091 Mitsubishi Space 3.947368 23.165158 82.157895 Engine CC Fuel Male Female Unknown Total 0 1497.169492 petrol 814172.000 422731.0 56.487 1293390.000 631666.0 1 554879.000 54.057 1166.142857 petrol 1240602.000 2 1537.406593 petrol 483216.000 310604.0 47.563 841383.000 3 241287.000 312556.0 28.004 581847.000 1219.653846 petrol 4 1995.777778 petrol 408016.000 115843.0 29.125 552984.000 ••• ••• ••• ••• ••• 80.000 1.242 6087 2304.975610 diesel 1.012 150.0 6088 2261.193548 petrol 670.000 482.0 66.000 1.218 6089 2696.500000 petrol 790.000 247.0 81.000 1.118 6090 2254.000000 petrol 639.000 416.0 49.000 1.104 6091 1817.315789 721.000 251.0 40.000 1.012 petrol [6092 rows x 11 columns] [10]: data = data.groupby(['Manufacturer', 'Model', 'Price', 'Transmission', 'Power', data [10]: Manufacturer Model Transmission Engine CC Price Power 0 Abarth 500C 21.105625 5.000000 60.625000 1039.500000 1 Alfa-Romeo 145 17.427143 5.000000 87.714286 1696.428571 2 Alfa-Romeo 146 18.198429 5.000000 87.714286 1696.428571 3 Alfa-Romeo 147 25.343387 5.193548 95.096774 1802.774194 4 Alfa-Romeo 155 23.909333 5.000000 100.666667 2016.777778 . . 497 Volvo **V**50 36.524191 5.176471 113.352941 2106.500000 498 Volvo V60 50.484130 7.304348 150.608696 1969.000000 499 Volvo V70 50.961091 6.181818 135.290909 1971.163636 500 Volvo XC70 58.469074 6.44444 148.518518 2174.037037 501 Volvo XC90 85.829462 8.000000 213.153846 1969.000000

```
7.124
                            653.070
                                     570.000
                                              12.413
     0
          petrol
     1
          petrol
                 1754.267
                           6311.549
                                    2186.000
                                              44.034
     2
          petrol
                1745.222
                           3780.461
                                    2969.000
                                              58.128
     3
          petrol
                  149.499
                             83.501
                                    2215.074
                                             246.278
     4
          petrol
                  883.005
                           4422.000 1191.000
                                              34.473
     497
         petrol
                  154.491
                            449.406 1879.292
                                             227.447
     498 diesel
                    7.831
                            880.290
                                    720.000
                                              10.720
     499
         diesel
                  703.323
                            587.631 1478.078
                                             928.874
     500 diesel
                  839.749
                           2172.313
                                    5608.000
                                              83.607
     501 petrol
                  164.061
                           781.948 1618.694 258.012
     [502 rows x 11 columns]
→populiariausi auto pagal firma
[12]: data_1 = data.iloc[:, [0, -1]]
     data 1
[12]:
         Manufacturer
                        Total
     0
              Abarth
                       12.413
                       44.034
     1
           Alfa-Romeo
     2
           Alfa-Romeo
                       58.128
     3
           Alfa-Romeo
                      246.278
     4
           Alfa-Romeo
                       34.473
     497
               Volvo
                      227.447
     498
               Volvo
                       10.720
     499
               Volvo 928.874
     500
               Volvo
                       83.607
     501
                      258.012
               Volvo
     [502 rows x 2 columns]
[13]: data_1 = data_1.groupby(['Manufacturer'], as_index=False).sum()
     data_1 = data_1.sort_values(['Total'], ascending=False)
     data 1
[13]:
         Manufacturer
                             Total
     17
                Ford 8.027580e+07
     36
              Peugeot
                      2.688234e+07
           Volkswagen
     50
                      2.510736e+07
     38
              Renault
                      2.048640e+07
     34
              Nissan 1.548595e+07
     4
                 \mathtt{BMW}
                      1.193343e+07
```

Fuel

Male

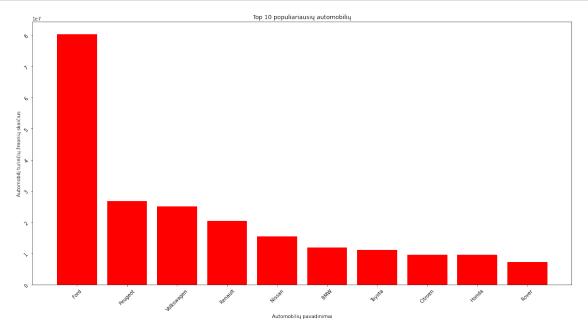
Female

Unknown

Total

```
48
          Toyota
                  1.107564e+07
8
         Citroen
                   9.626473e+06
18
           Honda
                   9.613328e+06
39
           Rover
                   7.327669e+06
30
        Mercedes
                   5.082870e+06
24
      Land-Rover
                   4.321683e+06
3
                   4.112435e+06
            Audi
42
           Skoda
                   1.933191e+06
16
            Fiat
                   1.446149e+06
41
            Seat
                   8.891286e+05
29
           Mazda
                   8.885507e+05
31
            Mini
                   7.779585e+05
40
            Saab
                   7.186725e+05
51
           Volvo
                   5.240522e+05
19
         Hyundai
                   1.102948e+05
21
              Kia
                   1.076330e+05
20
          Jaguar
                   1.030064e+05
46
          Suzuki
                   4.087961e+03
32
      Mitsubishi
                   2.160711e+03
45
          Subaru
                   1.340194e+03
27
               MG
                   1.057708e+03
1
      Alfa-Romeo
                   1.049063e+03
37
         Porsche
                   1.038770e+03
10
          Daewoo
                   9.820730e+02
7
        Chrysler
                   8.360680e+02
6
       Chevrolet
                   5.222720e+02
43
            Smart
                   3.899990e+02
11
        Daihatsu
                   3.269050e+02
13
          Datsun
                   2.824070e+02
22
                   2.745240e+02
            Lada
49
         Triumph
                   2.547010e+02
35
             Opel
                   2.063590e+02
26
           Lotus
                   1.246440e+02
33
          Morris
                   8.636600e+01
2
    Aston-Martin
                   8.573300e+01
44
       Ssangyong
                   6.581300e+01
23
          Lancia
                   6.417200e+01
5
         Bentley
                   6.102100e+01
14
           Dodge
                   5.765000e+01
25
           Lexus
                   3.583700e+01
12
         Daimler
                   3.434000e+01
9
           Dacia
                   1.495600e+01
0
          Abarth
                   1.241300e+01
47
              TVR
                   1.203100e+01
15
         Ferrari
                   1.121100e+01
                  3.598000e+00
28
        Maserati
```

```
fig, ax = plt.subplots(figsize=(20, 10))
plt.bar(data_1_1['Manufacturer'], data_1_1['Total'], color='r')
ax.set_xlabel('Automobilių pavadinimai')
ax.set_ylabel('Automobilį turinčių žmonių skaičius')
ax.tick_params(labelrotation=45)
ax.set_title('Top 10 populiariausių automobilių')
plt.savefig('most_popular_cars', dpi=100)
```



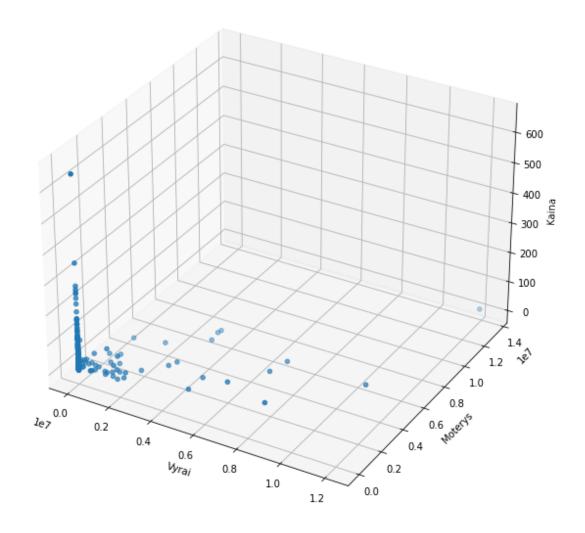
```
[16]: data_2 = data.iloc[:, [2, -4, -3]]
    data_2 = data_2.sort_values(['Price'])
    data_2
```

```
[16]:
                Price
                           Male
                                   Female
      397
             3.332200
                         60.665 1152.319
             3.781000 1525.677
                                 1621.767
      111
      213
             4.536000
                          1.295
                                  361.000
      396
             4.626333
                                 1846.300
                        730.901
      171
             4.968000 1671.380
                                 1473.000
      . .
                          •••
      262 264.675000
                        143.077
                                 2461.713
      14
           273.755000
                         24.625
                                 1989.000
      55
           285.247000
                         52.397 5032.024
```

```
15 360.755000 25.375 2035.000
347 646.605000 999.378 2693.000
```

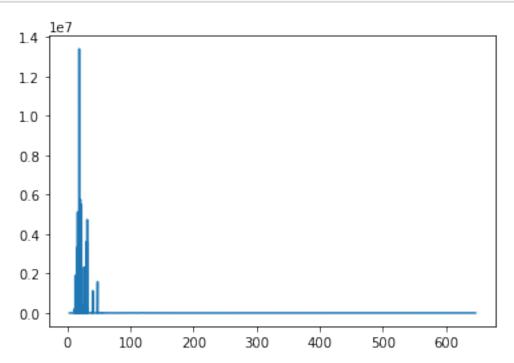
[502 rows x 3 columns]

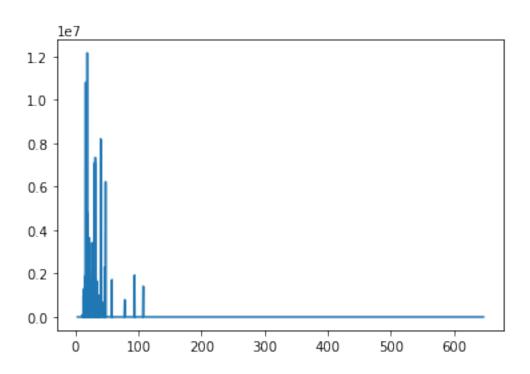
```
[17]: fig, ax = plt.subplots(figsize=(10, 10), subplot_kw={"projection": "3d"})
    ax.scatter(data_2['Male'], data_2['Female'], data_2['Price'])
    ax.set_xlabel('Vyrai')
    ax.set_ylabel('Moterys')
    ax.set_zlabel('Kaina')
    plt.savefig('kainu_pasiskirstymas', dpi=100)
```



[18]: # padaryt kad butu 2 grafikai ir jie greta sudeti butu. Tada braizytu kaina ir kiek vyru/moteru isigije uz ta kaina masinas

```
[19]: plt.plot(data_2['Price'], data_2['Female'])
   plt.show()
   plt.plot(data_2['Price'], data_2['Male'])
   plt.show()
```





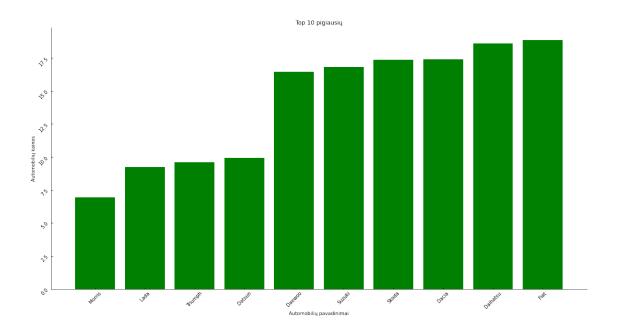
```
[20]: data_2['Female'].max()
[20]: 13391971.0
[21]: data_2['Male'].max()
[21]: 12150919.0
[22]: data_2['Price'].max()
[22]: 646.605
 [1]: class Graphics_plot:
          def __init__(self, data_x, data_y1, data_y2, x_name, y1_name, y2_name):
              self.data_x = data_x
              self.data_y1 = data_y1
              self.data_y2 = data_y2
              self.x_name = x_name
              self.y1_name = y1_name
              self.y2_name = y2_name
          def limits_y_axis(self): # suvienodina masteli y asiu
              y1_list = []
              y2_list = []
              for i in self.data_y1:
```

```
y1_list.append(i)
      for j in self.data_y2:
          y2_list.append(j)
       if np.array(y2_list).max() >= np.array(y1_list).max():
           y_lim_max = np.array(y2_list).max() + np.array(y2_list).mean()/10
       else:
           y_lim_max = np.array(y1_list).max() + np.array(y1_list).mean()/10
       if np.array(y2_list).min() <= np.array(y1_list).min():</pre>
           y_lim_min = np.array(y2_list).min() - np.array(y2_list).mean()/10
       else:
           y_lim_min = np.array(y1_list).min() - np.array(y1_list).mean()/10
      return y_lim_min, ', ', y_lim_max
  def limits_x_axis(self): # suvienodina masteli x asiu
      x1 list = []
      x2_list = []
      for i in self.data_x1:
          x1_list.append(i)
      for j in self.data_x2:
           x2_list.append(j)
       if np.array(x2_list).max() >= np.array(x1_list).max():
           x_lim_max = np.array(x2_list).max() + np.array(x2_list).mean()/10
      else:
           x lim max = np.array(x1 list).max() + np.array(x1 list).mean()/10
       if np.array(x2_list).min() <= np.array(x1_list).min():</pre>
           x_lim_min = np.array(x2_list).min() - np.array(x2_list).mean()/10
       else:
           x_lim_min = np.array(x1_list).min() - np.array(x1_list).mean()/10
      return x_lim_min, ', ', x_lim_max
  def plot(self, name):
      path = r"C:
→\Users\ariju\Documents\pamokos\programavimas\atsiskaitymas\mfp_naujas"
      if not os.path.exists(path):
           os.mkdir(path)
      fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(10, 4))
      ax1.scatter(self.data x, self.data y1, linewidth=3, color='b', alpha=1)
      ax2.scatter(self.data_x, self.data_y2, linewidth=3, marker='d',__
⇔color='r', alpha=1)
      ax1.spines["right"].set_visible(False)
      ax1.spines["top"].set_visible(False)
      ax2.spines["right"].set_visible(False)
      ax2.spines["top"].set_visible(False)
      ax1.tick_params(axis="y", direction="in")
      ax1.tick_params(axis="x", direction="in")
      ax2.tick_params(axis="y", direction="in")
      ax2.tick_params(axis="x", direction="in")
```

```
ax1.set_xlabel (self.x_name, size=10, font="Arial")
        ax1.set_ylabel(self.y1_name, size=10, font="Arial")
        ax2.set_xlabel (self.x_name, size=10, font="Arial")
        ax2.set_ylabel(self.y2_name, size=10, font="Arial")
        ax1.set_xlim(3.3, 647)
        ax1.set_ylim(60, 2700)
        ax2.set_xlim(3.3, 647)
        ax2.set_ylim(60, 2700)
        # ax1.set xlim(self.limits x axis)
        # ax1.set_ylim(self.limits_y_axis)
        # ax2.set xlim(self.limits y axis)
        # ax2.set_ylim(self.limits_y_axis)
        plt.savefig(os.path.join(path, f'{name}.png'), dpi=200)
        plt.close()
def separate_graphs(data, x, y1, y2, x_name, y1_name, y2_name): # x, y1, y2_{\sqcup}
 ⇒stulpeliu pavadinimai, data - duomenu dataframe'o pavadinimas, c_n -⊔
 ⇔stulpelio numeris is kurio imamas eiluciu skaicius
    NUM = 1
    lim = int(data.shape[0])
    for i in tqdm.tqdm(range(0, lim + 1)):
        graphics_plot = Graphics_plot(x.iloc[0:i],
                                       y1.iloc[0:i],
                                       y2.iloc[0:i],
                                       x_name,
                                       y1_name,
                                       y2_name
        name = '{num:02d}'.format(num=NUM)
        NIJM += 1
        graphics_plot.plot(name)
separate_graphs(data_2,
                data_2['Price'],
                data_2['Male'],
                data_2['Female'],
                'Kaina * 10^3 [EUR]',
                'Vyrai',
                'Moterys'
```

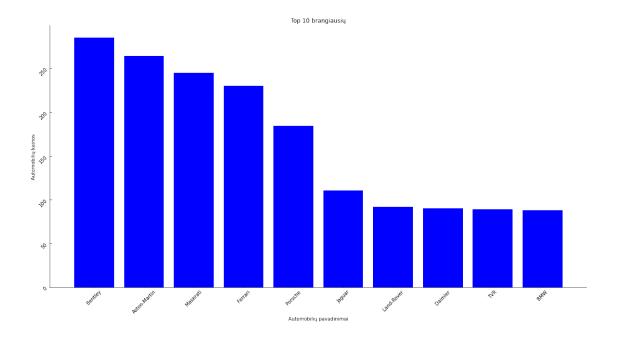
```
images = []
     for filename in tqdm.tqdm(filenames):
         image = imageio.imread(filename)
         images.append(image)
     imageio.mimsave(r'C:
      →\Users\ariju\Documents\pamokos\programavimas\atsiskaitymas\gifas.gif', u
       →images, fps=30, duration=0.5)
→top 10 pagal piguma ir palygint su top10 perkamiausiu
[26]: data_3 = data.iloc[:, [0, 2]]
     data_3 = data_3.groupby(['Manufacturer'], as_index=False).mean()
     data_3 = data_3.sort_values(['Price'])
     data 3
[26]:
         Manufacturer
                           Price
     33
              Morris
                        6.938000
     22
                Lada
                        9.230933
     49
              Triumph
                        9.593667
     13
               Datsun
                        9.933169
     10
               Daewoo
                       16.442375
     46
               Suzuki
                       16.796190
     42
               Skoda
                       17.340498
     9
               Dacia
                       17.398333
     11
             Daihatsu
                       18.593025
     16
                Fiat
                       18.852969
     39
               Rover
                       20.042730
     43
               Smart
                       20.549341
     0
                       21.105625
               Abarth
     41
                Seat
                       22.185178
     38
              Renault
                       23.042933
     35
                 Opel
                       23.083011
            Chevrolet
     6
                       23.324335
     21
                 Kia
                       24.828111
     8
              Citroen
                       25.429163
     19
              Hyundai
                       25.909890
     29
               Mazda
                       25.957820
     36
              Peugeot
                       26.496910
     23
               Lancia
                       28.596276
     32
           Mitsubishi
                       28.718452
     18
               Honda
                       29.707111
     48
               Toyota
                       29.987245
     17
                Ford
                       31.488367
     27
                  MG
                       31.712500
     1
           Alfa-Romeo
                       33.566362
```

```
31
                  Mini
                         34.143247
      50
            Volkswagen
                         37.015100
      14
                 Dodge
                         37.415083
      45
                Subaru
                         37.476506
      40
                  Saab
                         38.441615
      34
                Nissan
                         41.709880
      7
              Chrysler
                         42.210109
      51
                 Volvo
                         42.714411
      44
             Ssangyong
                         47.138946
      30
              Mercedes
                         66.462009
      25
                 Lexus
                         66.559947
      26
                 Lotus
                         73.608167
      3
                  Audi
                         74.302643
      4
                   BMW
                         87.993440
      47
                   TVR
                         89.443500
      12
               Daimler
                         90.406667
      24
            Land-Rover
                         91.933370
      20
                Jaguar
                        110.875494
      37
               Porsche
                        184.446424
      15
               Ferrari
                        230.242625
      28
              Maserati
                        245.168000
      2
          Aston-Martin 264.444333
      5
               Bentley
                        285.247000
[27]: data_3_10 = data_3.iloc[0:10, :]
      fig, ax = plt.subplots(figsize=(20, 10))
      plt.bar(data_3_10['Manufacturer'], data_3_10['Price'], color='g')
      ax.set_xlabel('Automobiliu pavadinimai')
      ax.set_ylabel('Automobilium kainos')
      ax.tick_params(labelrotation=45, direction="in")
      ax.spines["right"].set_visible(False)
      ax.spines["top"].set_visible(False)
      ax.set_title('Top 10 pigiausiu')
      plt.savefig('cheapest_cars', dpi=200)
```



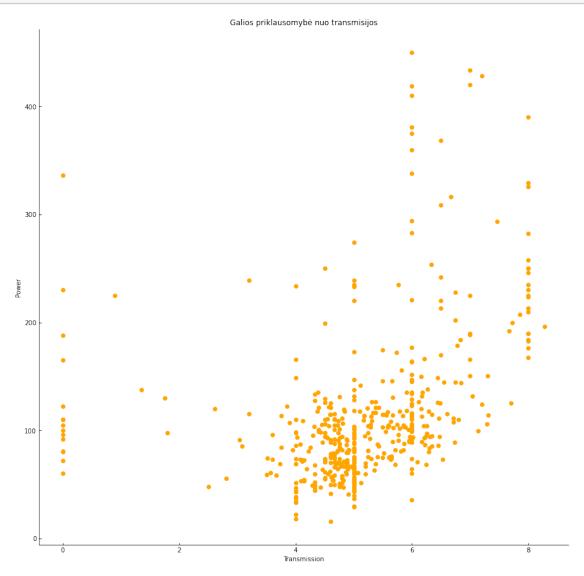
```
[28]: data_3_exp = data_3.iloc[-10:, :].sort_values(['Price'], ascending=False)

fig, ax = plt.subplots(figsize=(20, 10))
  plt.bar(data_3_exp['Manufacturer'], data_3_exp['Price'], color='b')
  ax.set_xlabel('Automobiliu pavadinimai')
  ax.set_ylabel('Automobiliu kainos')
  ax.tick_params(labelrotation=45, direction="in")
  ax.spines["right"].set_visible(False)
  ax.spines["top"].set_visible(False)
  ax.set_title('Top 10 brangiausiu')
  plt.savefig('expensive_cars', dpi=200)
```



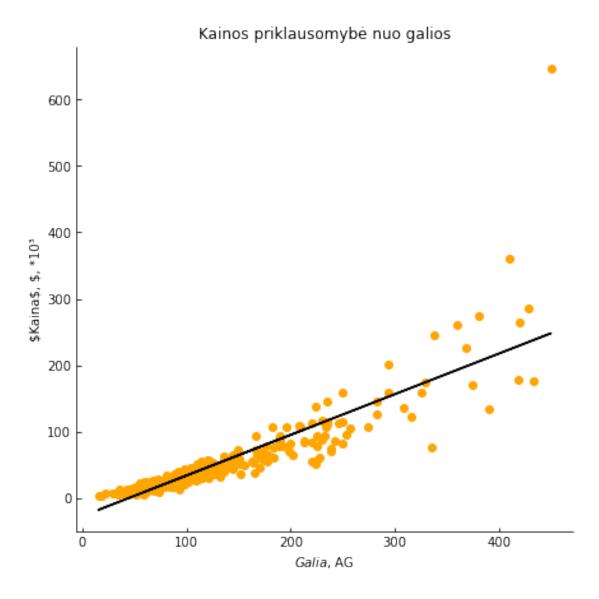
```
############### Transmission nuo Power priklausomybe
[30]: data_4 = data.iloc[:, [3, 4]]
      data_4
[30]:
           Transmission
                              Power
      0
               5.000000
                          60.625000
      1
               5.000000
                          87.714286
      2
               5.000000
                          87.714286
      3
               5.193548
                          95.096774
      4
               5.000000
                         100.666667
      497
               5.176471
                         113.352941
      498
               7.304348
                         150.608696
      499
               6.181818
                         135.290909
      500
               6.44444
                         148.518518
      501
               8.000000
                         213.153846
      [502 rows x 2 columns]
[31]: fig, ax = plt.subplots(figsize=(15, 15))
      plt.scatter(data_4['Transmission'], data_4['Power'], color='orange')
      ax.set_xlabel('Transmission')
      ax.set_ylabel('Power')
      ax.tick_params(direction="in")
      ax.spines["right"].set_visible(False)
      ax.spines["top"].set_visible(False)
```

```
ax.set_title('Galios priklausomybė nuo transmisijos')
plt.savefig('power_transmission', dpi=200)
```



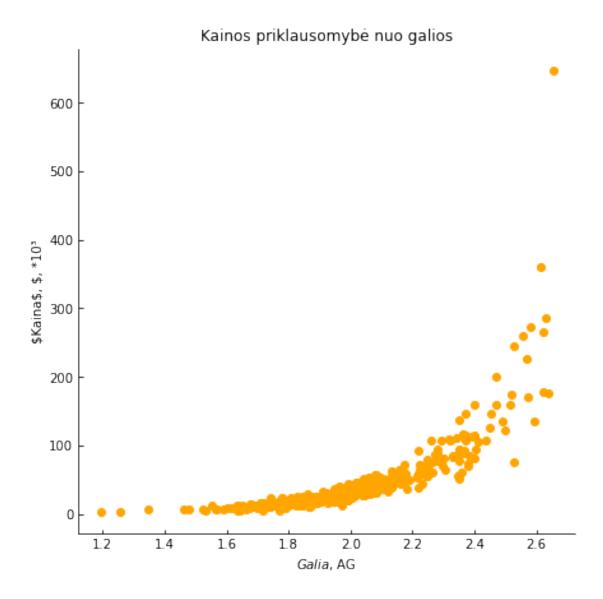
```
[33]: Price Power
0 21.105625 60.625000
1 17.427143 87.714286
2 18.198429 87.714286
3 25.343387 95.096774
```

```
4
          23.909333 100.666667
      . .
      497 36.524191 113.352941
      498 50.484130 150.608696
      499 50.961091 135.290909
      500 58.469074 148.518518
      501 85.829462 213.153846
      [502 rows x 2 columns]
[34]: fig, ax = plt.subplots(figsize=(7, 7))
      plt.scatter(data_5['Power'], data_5['Price'], color='orange')
      ax.set_xlabel('$Galia$, AG')
      ax.set_ylabel('$Kaina$, $, *103')
      ax.tick_params(direction="in")
      ax.spines["right"].set_visible(False)
      ax.spines["top"].set_visible(False)
      ax.set_title('Kainos priklausomybė nuo galios')
      function = np.poly1d(np.polyfit(data_5['Power'], data_5['Price'], 1))
      ax.plot(data_5['Power'], function(data_5['Power']), color="black")
      plt.savefig('price_power', dpi=200)
```



```
fig, ax = plt.subplots(figsize=(7, 7))
plt.scatter(np.log10(data_5['Power']), data_5['Price'], color='orange')
ax.set_xlabel('$Galia$, AG')
ax.set_ylabel('$Kaina$, $, *10³')
ax.tick_params(direction="in")
ax.spines["right"].set_visible(False)
ax.spines["top"].set_visible(False)
ax.set_title('Kainos priklausomybė nuo galios')
```

[35]: Text(0.5, 1.0, 'Kainos priklausomybė nuo galios')



36]:	data											
[36]:		Manufacturer	Model	Price	Transmission	Power	Engine CC	\				
	0	Abarth	500C	21.105625	5.000000	60.625000	1039.500000					
	1	Alfa-Romeo	145	17.427143	5.000000	87.714286	1696.428571					
	2	Alfa-Romeo	146	18.198429	5.000000	87.714286	1696.428571					
	3	Alfa-Romeo	147	25.343387	5.193548	95.096774	1802.774194					
	4	Alfa-Romeo	155	23.909333	5.000000	100.666667	2016.777778					
		•••	•••	•••	•••							
	497	Volvo	V50	36.524191	5.176471	113.352941	2106.500000					
	498	Volvo	V60	50.484130	7.304348	150.608696	1969.000000					
	499	Volvo	V70	50.961091	6.181818	135.290909	1971.163636					
	500	Volvo	XC70	58.469074	6.44444	148.518518	2174.037037					

```
8.000000 213.153846 1969.000000
      501
                Volvo XC90 85.829462
            Fuel
                       Male
                               Female
                                       Unknown
                                                   Total
      0
           petrol
                      7.124
                              653.070
                                        570.000
                                                  12.413
                             6311.549
                                                  44.034
      1
          petrol
                  1754.267
                                      2186.000
      2
          petrol
                  1745.222
                            3780.461
                                      2969.000
                                                  58.128
      3
          petrol
                    149.499
                               83.501
                                       2215.074
                                                 246.278
      4
          petrol
                   883.005
                            4422.000
                                       1191.000
                                                  34.473
                     •••
      497
          petrol
                    154.491
                              449.406 1879.292
                                                 227.447
      498
          diesel
                      7.831
                              880.290
                                       720.000
                                                  10.720
      499
          diesel
                   703.323
                              587.631 1478.078
                                                 928.874
      500
          diesel
                   839.749
                            2172.313
                                      5608.000
                                                  83.607
      501 petrol
                   164.061
                             781.948 1618.694
                                                 258.012
      [502 rows x 11 columns]
[37]: predict_data = data.iloc[:, [2, 4, 5]]
      predict_data
[37]:
                                    Engine CC
              Price
                           Power
                                 1039.500000
      0
           21.105625
                       60.625000
           17.427143
      1
                       87.714286
                                  1696.428571
      2
           18.198429
                      87.714286
                                  1696.428571
      3
           25.343387
                       95.096774
                                 1802.774194
      4
           23.909333 100.666667
                                  2016.777778
      497
          36.524191
                     113.352941
                                 2106.500000
      498 50.484130
                     150.608696
                                 1969.000000
      499
          50.961091
                     135.290909
                                 1971.163636
      500 58.469074
                     148.518518
                                 2174.037037
      501 85.829462 213.153846 1969.000000
      [502 rows x 3 columns]
[38]: predict_data['Price'].mean()
[38]: 41.50487473874703
[39]: predict_data = predict_data.round()
[40]: predict_data.loc[predict_data['Price'] <= 40 , 'Price'] = 0 # maziau lygu 40k -
      predict_data.loc[predict_data['Price'] > 40 , 'Price'] = 1 # daugiau uz 40k - 1
[41]: predict_data = predict_data.astype(int)
```

```
[42]: predict_data = predict_data.sort_values(['Price'], ascending=True)
[43]: predict_data
[43]:
          Price Power Engine CC
              0
                    61
                             1040
     323
              0
                    67
                             1699
     322
              0
                    56
                             1607
     320
              0
                    73
                             1242
     319
              0
                    80
                             1536
                   106
     253
                             1545
     82
              1
                   123
                             1832
     251
                             2972
              1
                   136
     258
              1
                   144
                             2418
     501
              1
                   213
                             1969
     [502 rows x 3 columns]
[44]: X_input = predict_data.drop(columns=['Price']).values
     Y_output = predict_data['Price'].values
     X_input_train, X_input_test, Y_output_train, Y_output_test =_
      strain_test_split(X_input, Y_output, test_size=0.2)
     model = DecisionTreeClassifier()
     model.fit(X_input_train, Y_output_train)
     predictions = model.predict(X_input_test)
     score = accuracy_score(Y_output_test, predictions)
     score
[44]: 0.881188118812
[45]: # X_input = predict_data.drop(columns=['Price'])
      # Y output = predict data['Price']
     # model = DecisionTreeClassifier()
     # model.fit(X_input.values, Y_output.values)
     # # Y_output = Y_output.astype(str)
     # tree.export_graphviz(model,
     #
                            out_file='price-predictor.dot',
     #
                            feature_names=['Power', 'Engine CC'],
     #
                            class_names=sorted(Y_output.unique()),
```

```
# label='all',
# rounded=True,
# filled=True
# )

[46]: X_input = predict_data.drop(columns=['Price'])
Y_output = predict_data['Price']

model = DecisionTreeClassifier()
model.fit(X_input.values, Y_output.values)
```

if predictions == 0:

print('Automobilis kainuoja ne daugiau 40.000€')

else:

print('Automobilis kainuoja daugiau nei 40.000€')

Automobilis kainuoja daugiau nei 40.000€

predictions = model.predict([[132, 1850]])

```
[47]: data
```

[47]:		Manufacture:	r Model	Price	e Transmi	ssion	Power	Engine CC	\
	0	Abart				00000	60.625000	1039.500000	•
	1	Alfa-Rome				00000	87.714286	1696.428571	
	2	Alfa-Rome				00000	87.714286	1696.428571	
	3	Alfa-Rome	o 147			93548	95.096774	1802.774194	
	4	Alfa-Rome	155	23.90933	3 5.0	00000	100.666667	2016.777778	
			•••	•••	•••	•		•	
	497	Volve	o V50	36.524193	1 5.1	76471	113.352941	2106.500000	
	498	Volve	o V60	50.484130	7.3	04348	150.608696	1969.000000	
	499	Volve	o V70	50.961093	1 6.1	81818	135.290909	1971.163636	
	500	Volve	xC70	58.469074	4 6.4	44444	148.518518	2174.037037	
	501	Volve	xC90	85.829462	2 8.0	00000	213.153846	1969.000000	
		Fuel	Male	Female	Unknown	Tota	al		
	0	petrol	7.124	653.070	570.000	12.43	13		
	1	petrol 17	54.267	6311.549	2186.000	44.03	34		
	2	petrol 17	45.222	3780.461	2969.000	58.12	28		
	3	petrol 1	49.499	83.501	2215.074	246.27	78		
	4	petrol 8	33.005	4422.000	1191.000	34.47	73		
			•••						
	497	petrol 1	54.491	449.406	1879.292	227.44	17		
	498	diesel	7.831	880.290	720.000	10.72	20		
	499	diesel 70	03.323	587.631	1478.078	928.87	74		
	500	diesel 8	39.749	2172.313	5608.000	83.60	07		
	501	petrol 1	64.061	781.948	1618.694	258.03	12		

[502 rows x 11 columns]

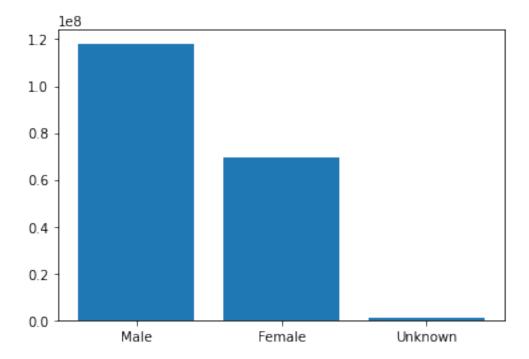
```
[48]: data_gender = data.iloc[:, [7, 8, 9]]
data_gender = data_gender.sum()
data_gender
```

[48]: Male 1.181561e+08
Female 6.936266e+07
Unknown 1.198816e+06
dtype: float64

[49]: data_gender= pd.DataFrame({'Gender':['Male', 'Female', 'Unknown']})
data_gender['Count']= data_gender['Gender'].apply(lambda x: data[x].sum())
data_gender

[49]: Gender Count
0 Male 1.181561e+08
1 Female 6.936266e+07
2 Unknown 1.198816e+06

[50]: plt.bar(data_gender['Gender'], data_gender['Count'])
plt.savefig('pasiskirstymasmfu', dpi=200)



```
[52]: data audi = pd.read_csv('skelbimai/audi.csv', usecols=[0, 1, 2, 4, 5, 7, 8])
     data_bmw = pd.read_csv('skelbimai/bmw.csv', usecols=[0, 1, 2, 4, 5, 7, 8])
     data_ford = pd.read_csv('skelbimai/ford.csv', usecols=[0, 1, 2, 4, 5, 7, 8])
     data_hyundai = pd.read_csv('skelbimai/hyundi.csv', usecols=[0, 1, 2, 4, 5, 7, __
     data_mercedes = pd.read_csv('skelbimai/merc.csv', usecols=[0, 1, 2, 4, 5, 7, 8])
     data_skoda = pd.read_csv('skelbimai/skoda.csv', usecols=[0, 1, 2, 4, 5, 7, 8])
     data_toyota = pd.read_csv('skelbimai/toyota.csv', usecols=[0, 1, 2, 4, 5, 7, 8])
     data_vauxhall = pd.read_csv('skelbimai/vauxhall.csv', usecols=[0, 1, 2, 4, 5, __
       →7, 8])
     data vw = pd.read csv('skelbimai/vw.csv', usecols=[0, 1, 2, 4, 5, 7, 8])
[53]: data_list = [data_audi, data_bmw, data_ford, data_hyundai, data_mercedes,_

data_skoda, data_toyota, data_vauxhall, data_vw]

     new_names list = ['Audi', 'Bmw', 'Ford', 'Hyundai', 'Mercedes', 'Skoda', __

¬'Toyota', 'Vauxhall', 'Volkswagen']
[54]: | # Gallon Per Hour (GPH) = (specific fuel consumption x HP)/Fuel Specific Weight
      # diesel --> power = (7.2 * 60 / mpg) / 0.045 * 0.7457
      \# petrol \longrightarrow power = (6.1 * 60 / mpg) / 0.55 * 0.7457
[55]: class New data cleaning:
         def __init__(self, data_name):
              self.data_name = data_name
         def rename_column_values(self, col_title, new_col_values_name):
              self.data_name.loc[self.data_name[col_title] != new_col_values_name,_u
       Gol_title] = new_col_values_name
         def miles to km(self, col title):
              self.data_name[col_title] = self.data_name[col_title] * 1.61
             self.data_name[col_title] = self.data_name[col_title].round()
         def pounds_to_eur(self, col_title):
             self.data_name[col_title] = self.data_name[col_title] * 1.17
             self.data_name[col_title] = self.data_name[col_title].round()
         def mpg_to_power(self, col_title_with_fuel, col_title_with_mpg):
              self.data_name.loc[self.data_name[col_title_with_fuel] == 'Petrol',u
       col_title_with_mpg] = ((6.1 * (60 / self.data_name[col_title_with_mpg])) / 0.
       4045) * 0.7457
              self.data_name.loc[self.data_name[col_title_with_fuel] == 'Diesel',u
       col_title_with_mpg] = ((7.2 * (60 / self.data_name[col_title_with_mpg])) / 0.
       →055) * 0.7457
```

```
self.data_name.drop(self.data_name.loc[self.
       data_name[col_title_with_fuel] == 'Hybrid'].index, inplace=True)
             self.data_name.drop(self.data_name.loc[self.
       data name[col title with fuel] == 'Other'].index, inplace=True)
             self.data_name.drop(self.data_name.loc[self.
       data_name[col_title_with_fuel] == 'Electric'].index, inplace=True)
             self.data_name[col_title_with_mpg] = self.data_name[col_title_with_mpg].
       →round()
     for i in tqdm.tqdm(range(0, 9)):
         new_data_cleaning = New_data_cleaning(data_list[i])
         new_data_cleaning.rename_column_values('model', new_names_list[i])
         new_data_cleaning.miles_to_km('mileage')
         new_data_cleaning.pounds_to_eur('price')
         new_data_cleaning.mpg_to_power('fuelType', 'mpg')
     100%|
         | 9/9 [00:00<00:00, 108.38it/s]
[56]: data_cars = pd.concat(data_list)
     data cars
[56]:
                 model year
                                price
                                        mileage fuelType
                                                           mpg engineSize
                       2017 14625.0
                                        25333.0
                                                 Petrol 109.0
     0
                  Audi
                                                                       1.4
     1
                  Audi 2016 19305.0
                                        58287.0
                                                 Diesel
                                                          91.0
                                                                       2.0
                  Audi 2016 12870.0
                                        48213.0 Petrol 109.0
                                                                       1.4
     2
     3
                  Audi 2017 19656.0
                                        41783.0 Diesel 87.0
                                                                       2.0
                  Audi 2019 20241.0
                                        3217.0 Petrol 122.0
                                                                       1.0
                               7008.0 119140.0 Diesel
                                                         99.0
                                                                       2.0
     15152 Volkswagen 2012
     15153 Volkswagen 2008
                               2105.0 141844.0
                                                 Petrol 131.0
                                                                       1.2
                               1860.0 112700.0 Petrol 144.0
                                                                       1.4
     15154 Volkswagen 2009
     15155 Volkswagen 2006
                               1462.0 133153.0 Petrol 131.0
                                                                       1.2
     15156 Volkswagen 2007
                               2685.0 119140.0 Petrol 131.0
                                                                       1.2
     [95856 rows x 7 columns]
[57]: data_cars = data_cars.rename(columns={'model': 'Manufacturer',
                                           'year': 'Year',
                                           'price': 'Price',
                                           'mileage': 'Distance km',
                                           'fuelType': 'Fuel',
                                           'mpg': 'Power',
                                           'engineSize': 'Engine_size'
                                          })
     data_cars
```

```
[57]:
            Manufacturer Year
                                  Price Distance_km
                                                         Fuel Power
                                                                      Engine_size
                    Audi 2017
                                                               109.0
      0
                                14625.0
                                             25333.0 Petrol
                                                                              1.4
      1
                    Audi
                          2016 19305.0
                                             58287.0
                                                      Diesel
                                                                91.0
                                                                              2.0
      2
                    Audi
                          2016
                                12870.0
                                             48213.0
                                                      Petrol 109.0
                                                                              1.4
      3
                    Audi
                          2017
                                             41783.0 Diesel
                                                                87.0
                                                                              2.0
                                19656.0
      4
                    Audi
                          2019
                                20241.0
                                              3217.0 Petrol
                                                               122.0
                                                                              1.0
                                                                •••
                                               •••
      15152
              Volkswagen
                         2012
                                 7008.0
                                             119140.0 Diesel
                                                                99.0
                                                                              2.0
                         2008
                                 2105.0
                                                              131.0
      15153
              Volkswagen
                                             141844.0 Petrol
                                                                              1.2
      15154
              Volkswagen
                          2009
                                 1860.0
                                             112700.0
                                                      Petrol
                                                               144.0
                                                                              1.4
              Volkswagen
                          2006
                                                               131.0
                                                                              1.2
      15155
                                 1462.0
                                             133153.0
                                                      Petrol
      15156
              Volkswagen 2007
                                 2685.0
                                            119140.0 Petrol
                                                               131.0
                                                                              1.2
      [95856 rows x 7 columns]
[58]: data.drop(data.loc[data['Fuel'] == ('automatic' or 'Automatic' or 'other' or<sub>||</sub>
       data
[58]:
          Manufacturer Model
                                  Price Transmission
                                                             Power
                                                                      Engine CC \
                Abarth 500C
                              21.105625
                                             5.000000
                                                         60.625000
                                                                    1039.500000
      1
            Alfa-Romeo
                         145
                              17.427143
                                             5.000000
                                                         87.714286
                                                                    1696.428571
      2
            Alfa-Romeo
                         146
                              18.198429
                                             5.000000
                                                         87.714286
                                                                    1696.428571
      3
            Alfa-Romeo
                         147
                              25.343387
                                             5.193548
                                                         95.096774
                                                                    1802.774194
      4
            Alfa-Romeo
                         155
                              23.909333
                                             5.000000
                                                        100.666667
                                                                    2016.777778
      . .
                                                                    2106.500000
      497
                 Volvo
                         V50
                              36.524191
                                             5.176471
                                                       113.352941
      498
                 Volvo
                         V60
                              50.484130
                                             7.304348
                                                       150.608696
                                                                    1969.000000
      499
                 Volvo
                         V70
                              50.961091
                                             6.181818
                                                       135.290909
                                                                    1971.163636
      500
                 Volvo
                        XC70
                              58.469074
                                             6.44444
                                                       148.518518
                                                                    2174.037037
      501
                        XC90
                 Volvo
                              85.829462
                                             8.000000 213.153846
                                                                    1969.000000
             Fuel
                       Male
                               Female
                                        Unknown
                                                   Total
                      7.124
                                        570.000
                                                   12.413
      0
           petrol
                              653.070
                                                   44.034
      1
           petrol 1754.267
                             6311.549
                                       2186.000
      2
                             3780.461
                                       2969.000
           petrol
                  1745.222
                                                  58.128
      3
           petrol
                    149.499
                               83.501
                                       2215.074
                                                 246.278
      4
           petrol
                    883.005
                             4422.000
                                       1191.000
                                                  34.473
      . .
              •••
      497
                                       1879.292
          petrol
                    154.491
                              449.406
                                                 227.447
      498
          diesel
                      7.831
                              880.290
                                        720.000
                                                  10.720
      499
           diesel
                    703.323
                              587.631
                                       1478.078
                                                 928.874
      500
           diesel
                    839.749
                             2172.313
                                       5608.000
                                                  83.607
      501
           petrol
                    164.061
                              781.948
                                       1618.694
                                                 258.012
```

[497 rows x 11 columns]

```
[59]: data_add = data.drop(columns=['Model', 'Transmission', 'Male', 'Female', |

    'Unknown', 'Total'])
[60]: data_add = data_add.rename(columns={'Engine CC': 'Engine_size'})
      data_add['Engine_size'] = data_add['Engine_size'] * 0.06101 * 0.01639
      data_add['Price'] = data_add['Price'] * 1000
      data add['Price'] = data add['Price'].round()
      data_add['Power'] = data_add['Power'].round()
      data_add['Engine_size'] = data_add['Engine_size'].round()
      data_add['Fuel'] = data_add['Fuel'].str.capitalize()
[61]: data_add
[61]:
          Manufacturer
                          Price Power Engine_size
                                                       Fuel
                Abarth 21106.0
                                 61.0
                                                1.0 Petrol
                                 88.0
      1
           Alfa-Romeo 17427.0
                                                2.0 Petrol
      2
           Alfa-Romeo 18198.0
                                 88.0
                                                2.0 Petrol
      3
           Alfa-Romeo 25343.0
                                 95.0
                                                2.0 Petrol
      4
           Alfa-Romeo 23909.0 101.0
                                                2.0 Petrol
      497
                 Volvo
                       36524.0 113.0
                                                2.0 Petrol
      498
                Volvo 50484.0 151.0
                                                2.0 Diesel
      499
                Volvo 50961.0 135.0
                                                2.0 Diesel
      500
                Volvo 58469.0 149.0
                                                2.0 Diesel
      501
                Volvo 85829.0 213.0
                                                2.0 Petrol
      [497 rows x 5 columns]
[62]: class Manage_cols():
          def __init__(self, df):
              self.df = df
          def add_new_col(self, new_col_location, new_col_name, new_col_value):
              self.df.insert(loc=new_col_location,
                             column=new_col_name,
                             value=new_col_value
          def swap_columns(self, col_1, col_2):
              col_list = list(self.df.columns)
              x, y = col_list.index(col_1), col_list.index(col_2)
              col_list[y], col_list[x] = col_list[x], col_list[y]
              self.df = self.df[col list]
              return self.df
      manage_cols = Manage_cols(data_add)
      manage_cols.add_new_col(1, 'Year', 2022)
```

```
manage_cols.add_new_col(3, 'Distance_km', 0)
      data_add = manage_cols.swap_columns('Power', 'Fuel')
      data_add = manage_cols.swap_columns('Engine_size', 'Power')
[63]: data_add
[63]:
                                      Distance_km
                                                      Fuel Power
                                                                    Engine_size
          Manufacturer
                        Year
                                Price
                                                              61.0
      0
                Abarth
                        2022
                              21106.0
                                                    Petrol
                                                                            1.0
      1
            Alfa-Romeo
                        2022
                              17427.0
                                                    Petrol
                                                              88.0
                                                                            2.0
      2
            Alfa-Romeo
                        2022
                              18198.0
                                                    Petrol
                                                              88.0
                                                                            2.0
      3
                        2022
                              25343.0
                                                              95.0
            Alfa-Romeo
                                                  0
                                                    Petrol
                                                                            2.0
      4
            Alfa-Romeo
                        2022
                              23909.0
                                                    Petrol 101.0
                                                                            2.0
      497
                 Volvo
                        2022
                              36524.0
                                                    Petrol
                                                            113.0
                                                                            2.0
                                                 0
      498
                 Volvo
                        2022
                              50484.0
                                                 0
                                                    Diesel
                                                            151.0
                                                                            2.0
      499
                 Volvo
                              50961.0
                                                    Diesel
                                                            135.0
                                                                            2.0
                        2022
      500
                 Volvo
                        2022
                              58469.0
                                                    Diesel 149.0
                                                                            2.0
      501
                 Volvo
                        2022
                              85829.0
                                                 0 Petrol 213.0
                                                                            2.0
      [497 rows x 7 columns]
     data_cars
[64]:
[64]:
            Manufacturer Year
                                  Price
                                         Distance_km
                                                        Fuel
                                                              Power
                                                                      Engine size
                    Audi
                          2017
                                14625.0
                                             25333.0
                                                      Petrol
                                                               109.0
                                                                              1.4
      0
                    Audi 2016
                                19305.0
                                             58287.0
                                                      Diesel
                                                                91.0
                                                                              2.0
      1
      2
                    Audi 2016
                                12870.0
                                             48213.0
                                                      Petrol 109.0
                                                                              1.4
      3
                    Audi
                          2017
                                19656.0
                                             41783.0
                                                      Diesel
                                                                87.0
                                                                              2.0
      4
                    Audi
                          2019
                                                              122.0
                                20241.0
                                              3217.0
                                                      Petrol
                                                                              1.0
                                               ...
                                                                ...
      15152
              Volkswagen
                         2012
                                 7008.0
                                            119140.0 Diesel
                                                                99.0
                                                                              2.0
                                                                              1.2
      15153
              Volkswagen
                         2008
                                 2105.0
                                            141844.0
                                                      Petrol
                                                              131.0
      15154
              Volkswagen 2009
                                 1860.0
                                            112700.0 Petrol
                                                              144.0
                                                                              1.4
      15155
              Volkswagen 2006
                                 1462.0
                                            133153.0
                                                      Petrol
                                                              131.0
                                                                              1.2
              Volkswagen 2007
                                                                              1.2
      15156
                                 2685.0
                                            119140.0 Petrol
                                                              131.0
      [95856 rows x 7 columns]
[65]: data_cars = pd.concat([data_add, data_cars])
      data_cars
[65]:
            Manufacturer Year
                                  Price Distance km
                                                        Fuel Power
                                                                      Engine size
                  Abarth 2022 21106.0
                                                 0.0 Petrol
                                                                61.0
                                                                              1.0
      0
      1
              Alfa-Romeo 2022 17427.0
                                                 0.0 Petrol
                                                                88.0
                                                                              2.0
      2
              Alfa-Romeo 2022
                                18198.0
                                                 0.0 Petrol
                                                                88.0
                                                                              2.0
                                                                95.0
      3
              Alfa-Romeo 2022
                                25343.0
                                                 0.0 Petrol
                                                                              2.0
              Alfa-Romeo 2022
                                23909.0
                                                 0.0 Petrol
                                                              101.0
                                                                              2.0
```

```
15152
                                 7008.0
                                                                99.0
                                                                               2.0
              Volkswagen
                          2012
                                             119140.0 Diesel
                                                                               1.2
      15153
              Volkswagen
                          2008
                                 2105.0
                                             141844.0 Petrol 131.0
              Volkswagen
                          2009
                                  1860.0
                                             112700.0 Petrol 144.0
                                                                               1.4
      15154
      15155
              Volkswagen 2006
                                 1462.0
                                             133153.0 Petrol 131.0
                                                                               1.2
              Volkswagen 2007
                                 2685.0
                                                                               1.2
      15156
                                             119140.0 Petrol 131.0
      [96353 rows x 7 columns]
[66]: car_names = []
      for i in data_cars['Manufacturer']:
          if i not in car_names:
              car_names.append(i)
[67]: # car names = []
      # [car_names.append(i) for i in data_cars['Manufacturer'] if i not in car_names]
[68]: car_names
[68]: ['Abarth',
       'Alfa-Romeo',
       'Aston-Martin',
       'Audi',
       'BMW',
       'Bentley',
       'Chevrolet',
       'Chrysler',
       'Citroen',
       'Dacia',
       'Daewoo',
       'Daihatsu',
       'Daimler',
       'Datsun',
       'Dodge',
       'Ferrari',
       'Fiat',
       'Ford',
       'Honda',
       'Hyundai',
       'Jaguar',
       'Kia',
       'Lada',
       'Lancia',
       'Land-Rover',
       'Lexus',
       'Lotus',
       'MG',
```

```
'Maserati',
       'Mazda',
       'Mercedes',
       'Mini',
       'Mitsubishi',
       'Morris',
       'Nissan',
       'Opel',
       'Peugeot',
       'Porsche',
       'Renault',
       'Rover',
       'Saab',
       'Seat',
       'Skoda',
       'Smart',
       'Ssangyong',
       'Subaru',
       'Suzuki',
       'TVR',
       'Toyota',
       'Triumph',
       'Volkswagen',
       'Volvo',
       'Bmw',
       'Vauxhall']
[69]: ############ neuroninis tinklas
[70]: data_cars = data_cars.drop(columns=['Manufacturer'])
[71]: data_cars.loc[data_cars['Fuel'] == 'Petrol', 'Fuel'] = 0 # petrol = 0
      data_cars.loc[data_cars['Fuel'] == 'Diesel', 'Fuel'] = 1 # diesel = 1
[72]: # manufacturer_name_to_id = {}
      # idx = 0
      # for i in car_names:
           manufacturer_name_to_id.update({i: idx})
      #
            idx += 1
[73]: # manufacturer_name_to_id
[74]: data_cars.dtypes
[74]: Year
                       int64
                     float64
     Price
      Distance_km
                     float64
```

```
Fuel
                      object
      Power
                     float64
      Engine_size
                     float64
      dtype: object
[75]: data_cars['Year'] = data_cars['Year'].astype('float')
      data cars['Fuel'] = data cars['Fuel'].astype('float')
[76]: # data_cars['Year'] = data_cars['Year'] / 2022
      # data_cars['Price'] = data_cars['Price'] / 646605
      # data_cars['Distance_km'] = data_cars['Distance_km'] / 520030
      # data_cars['Engine_size'] = data_cars['Engine_size'] / 6.6
      # data_cars['Power'] = data_cars['Power'] / 20217
[77]: data_cars
[77]:
               Year
                       Price Distance_km Fuel Power
                                                        Engine_size
      0
             2022.0 21106.0
                                      0.0
                                            0.0
                                                  61.0
                                                                 1.0
             2022.0 17427.0
                                      0.0
                                                  88.0
                                                                 2.0
      1
                                            0.0
      2
             2022.0 18198.0
                                      0.0
                                            0.0
                                                  88.0
                                                                 2.0
      3
             2022.0
                     25343.0
                                      0.0
                                            0.0
                                                  95.0
                                                                 2.0
      4
             2022.0
                     23909.0
                                      0.0
                                            0.0 101.0
                                                                 2.0
      15152 2012.0
                      7008.0
                                 119140.0
                                                                 2.0
                                            1.0
                                                  99.0
      15153 2008.0
                      2105.0
                                 141844.0
                                            0.0 131.0
                                                                 1.2
      15154 2009.0
                      1860.0
                                 112700.0
                                            0.0 144.0
                                                                 1.4
                                                                 1.2
      15155 2006.0
                      1462.0
                                 133153.0
                                            0.0 131.0
      15156 2007.0
                      2685.0
                                            0.0 131.0
                                                                 1.2
                                 119140.0
      [96353 rows x 6 columns]
[78]: print(data_cars['Price'].max())
      print(data_cars['Price'].mean())
      print(data_cars['Price'].min())
     646605.0
     19680.81250194597
     526.0
[79]: \# (0-5] k \longrightarrow 0
      # (5-10] k --> 1
      # (10-15] k --> 2
      # (15-20] k --> 3
      # (20-25] k --> 4
      # (25-35] k --> 5
      # (35<=] k --> 6
```

```
[80]: def price_to_index(start, end, idx):
          data_cars.loc[(data_cars['Price'] > start) & (data_cars['Price'] <= end),__
       →'Price'] = idx
      # price_to_index(0, 5000, 0)
      # price_to_index(5000, 10000, 1)
      # price_to_index(10000, 15000, 2)
      # price_to_index(15000, 20000, 3)
      # price_to_index(20000, 25000, 4)
      # price_to_index(25000, 35000, 5)
      # price_to_index(35000, 5000000, 6)
[81]: price_to_index(0, 20000, 0)
      price_to_index(20000, 200000000, 1)
[82]: data_cars
[82]:
               Year Price Distance_km Fuel Power Engine_size
             2022.0
                       1.0
      0
                                    0.0
                                          0.0
                                                 61.0
                                                               1.0
             2022.0
                                                               2.0
      1
                       0.0
                                    0.0
                                          0.0
                                                88.0
      2
             2022.0
                       0.0
                                    0.0
                                          0.0
                                                88.0
                                                               2.0
      3
             2022.0
                                    0.0
                                                               2.0
                       1.0
                                          0.0
                                                95.0
      4
                                    0.0
                                                               2.0
             2022.0
                       1.0
                                          0.0 101.0
                                                               2.0
      15152 2012.0
                       0.0
                               119140.0
                                          1.0
                                                99.0
      15153 2008.0
                       0.0
                               141844.0
                                          0.0 131.0
                                                               1.2
      15154 2009.0
                       0.0
                               112700.0
                                          0.0 144.0
                                                               1.4
      15155 2006.0
                       0.0
                               133153.0
                                          0.0 131.0
                                                               1.2
      15156 2007.0
                       0.0
                               119140.0
                                          0.0 131.0
                                                               1.2
      [96353 rows x 6 columns]
[83]: tf.convert_to_tensor(data_cars)
[83]: <tf.Tensor: shape=(96353, 6), dtype=float64, numpy=
      array([[2.02200e+03, 1.00000e+00, 0.00000e+00, 0.00000e+00, 6.10000e+01,
              1.00000e+00],
             [2.02200e+03, 0.00000e+00, 0.00000e+00, 0.00000e+00, 8.80000e+01,
              2.00000e+00],
             [2.02200e+03, 0.00000e+00, 0.00000e+00, 0.00000e+00, 8.80000e+01,
              2.00000e+00],
             [2.00900e+03, 0.00000e+00, 1.12700e+05, 0.00000e+00, 1.44000e+02,
              1.40000e+00],
             [2.00600e+03, 0.00000e+00, 1.33153e+05, 0.00000e+00, 1.31000e+02,
             [2.00700e+03, 0.00000e+00, 1.19140e+05, 0.00000e+00, 1.31000e+02,
```

```
[84]: normalizer = tf.keras.layers.Normalization(axis=-1)
      normalizer.adapt(data_cars.drop(columns=['Price']))
[85]: normalizer(data_cars.drop(columns=['Price']).iloc[:3])
[85]: <tf.Tensor: shape=(3, 5), dtype=float32, numpy=
      array([[ 2.2651272 , -1.081996  , -0.86067873, -0.75479716, -1.1706171 ],
             [ 2.2651272 , -1.081996 , -0.86067873, -0.38674808, 0.5978566 ],
             [ 2.2651272 , -1.081996 , -0.86067873, -0.38674808, 0.5978566 ]],
            dtype=float32)>
[86]: x_data = data_cars.drop(columns=['Price'])
      y_data = data_cars['Price']
      x_train, x_test, y_train, y_test = train_test_split(x_data, y_data, test_size=0.
[87]: data_cars.shape
[87]: (96353, 6)
[88]: n = 50
      model = tf.keras.Sequential([
          normalizer.
          tf.keras.layers.Dense(units=round((n+1)/2), activation='relu'),
          tf.keras.layers.Dropout(rate=0.2),
          tf.keras.layers.Dense(units=round((n+1)/4), activation='relu'),
          tf.keras.layers.Dropout(rate=0.2),
          tf.keras.layers.Dense(units=1, activation='relu')
      ])
      model.compile(loss=tf.keras.losses.BinaryCrossentropy(from_logits=True),
                    optimizer='adam',
                    metrics=['accuracy'])
      val_accuracy_list = []
      while len(val_accuracy_list) < 10:</pre>
          history = model.fit(x_train,
                              y_train,
                              batch_size=50,
                              epochs=100,
                              validation_data=(x_test, y_test))
          val_accuracy_list.append(history.history['val_accuracy'])
```

1.20000e+00]])>

```
Epoch 1/100
accuracy: 0.8475 - val_loss: 0.5567 - val_accuracy: 0.8710
Epoch 2/100
accuracy: 0.8684 - val_loss: 0.5549 - val_accuracy: 0.8735
1349/1349 [============ ] - 1s 1ms/step - loss: 0.5591 -
accuracy: 0.8707 - val_loss: 0.5531 - val_accuracy: 0.8768
Epoch 4/100
accuracy: 0.8741 - val_loss: 0.5509 - val_accuracy: 0.8773
Epoch 5/100
1349/1349 [============= - - 1s 1ms/step - loss: 0.5554 -
accuracy: 0.8756 - val_loss: 0.5496 - val_accuracy: 0.8783
Epoch 6/100
accuracy: 0.8774 - val_loss: 0.5478 - val_accuracy: 0.8790
Epoch 7/100
accuracy: 0.8787 - val_loss: 0.5466 - val_accuracy: 0.8804
Epoch 8/100
accuracy: 0.8791 - val_loss: 0.5456 - val_accuracy: 0.8809
Epoch 9/100
accuracy: 0.8812 - val_loss: 0.5453 - val_accuracy: 0.8788
Epoch 10/100
accuracy: 0.8816 - val_loss: 0.5444 - val_accuracy: 0.8794
Epoch 11/100
accuracy: 0.8814 - val_loss: 0.5436 - val_accuracy: 0.8826
Epoch 12/100
accuracy: 0.8817 - val_loss: 0.5429 - val_accuracy: 0.8841
Epoch 13/100
accuracy: 0.8820 - val_loss: 0.5429 - val_accuracy: 0.8800
Epoch 14/100
accuracy: 0.8824 - val_loss: 0.5429 - val_accuracy: 0.8820
Epoch 15/100
accuracy: 0.8823 - val_loss: 0.5420 - val_accuracy: 0.8837
Epoch 16/100
accuracy: 0.8839 - val_loss: 0.5415 - val_accuracy: 0.8836
```

```
Epoch 17/100
accuracy: 0.8830 - val_loss: 0.5412 - val_accuracy: 0.8832
Epoch 18/100
accuracy: 0.8828 - val_loss: 0.5412 - val_accuracy: 0.8847
Epoch 19/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5446 -
accuracy: 0.8824 - val_loss: 0.5419 - val_accuracy: 0.8806
Epoch 20/100
accuracy: 0.8819 - val_loss: 0.5407 - val_accuracy: 0.8854
Epoch 21/100
1349/1349 [============== - - 1s 1ms/step - loss: 0.5441 -
accuracy: 0.8834 - val_loss: 0.5406 - val_accuracy: 0.8823
Epoch 22/100
accuracy: 0.8833 - val_loss: 0.5409 - val_accuracy: 0.8805
Epoch 23/100
1349/1349 [============ ] - 1s 1ms/step - loss: 0.5435 -
accuracy: 0.8834 - val_loss: 0.5413 - val_accuracy: 0.8811
Epoch 24/100
accuracy: 0.8835 - val_loss: 0.5402 - val_accuracy: 0.8867
Epoch 25/100
accuracy: 0.8824 - val_loss: 0.5400 - val_accuracy: 0.8822
Epoch 26/100
accuracy: 0.8824 - val_loss: 0.5394 - val_accuracy: 0.8856
Epoch 27/100
accuracy: 0.8843 - val_loss: 0.5398 - val_accuracy: 0.8829
Epoch 28/100
accuracy: 0.8835 - val_loss: 0.5396 - val_accuracy: 0.8842
Epoch 29/100
accuracy: 0.8844 - val_loss: 0.5392 - val_accuracy: 0.8842
Epoch 30/100
accuracy: 0.8842 - val_loss: 0.5386 - val_accuracy: 0.8859
Epoch 31/100
accuracy: 0.8833 - val_loss: 0.5392 - val_accuracy: 0.8832
Epoch 32/100
accuracy: 0.8838 - val_loss: 0.5394 - val_accuracy: 0.8878
```

```
Epoch 33/100
accuracy: 0.8834 - val_loss: 0.5389 - val_accuracy: 0.8841
Epoch 34/100
accuracy: 0.8825 - val_loss: 0.5391 - val_accuracy: 0.8855
Epoch 35/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5423 -
accuracy: 0.8840 - val_loss: 0.5388 - val_accuracy: 0.8840
Epoch 36/100
accuracy: 0.8822 - val_loss: 0.5390 - val_accuracy: 0.8835
Epoch 37/100
accuracy: 0.8838 - val_loss: 0.5380 - val_accuracy: 0.8852
Epoch 38/100
1349/1349 [============= ] - 1s 1ms/step - loss: 0.5420 -
accuracy: 0.8836 - val_loss: 0.5388 - val_accuracy: 0.8831
Epoch 39/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5423 -
accuracy: 0.8842 - val_loss: 0.5392 - val_accuracy: 0.8875
Epoch 40/100
accuracy: 0.8835 - val_loss: 0.5387 - val_accuracy: 0.8837
Epoch 41/100
accuracy: 0.8841 - val_loss: 0.5380 - val_accuracy: 0.8842
Epoch 42/100
accuracy: 0.8841 - val_loss: 0.5387 - val_accuracy: 0.8860
Epoch 43/100
accuracy: 0.8834 - val_loss: 0.5386 - val_accuracy: 0.8852
Epoch 44/100
accuracy: 0.8838 - val_loss: 0.5379 - val_accuracy: 0.8852
Epoch 45/100
accuracy: 0.8848 - val_loss: 0.5381 - val_accuracy: 0.8849
Epoch 46/100
accuracy: 0.8834 - val_loss: 0.5379 - val_accuracy: 0.8848
Epoch 47/100
accuracy: 0.8842 - val_loss: 0.5385 - val_accuracy: 0.8821
Epoch 48/100
accuracy: 0.8838 - val_loss: 0.5380 - val_accuracy: 0.8827
```

```
Epoch 49/100
accuracy: 0.8825 - val_loss: 0.5381 - val_accuracy: 0.8841
Epoch 50/100
accuracy: 0.8853 - val_loss: 0.5400 - val_accuracy: 0.8812
Epoch 51/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5414 -
accuracy: 0.8850 - val_loss: 0.5376 - val_accuracy: 0.8870
Epoch 52/100
accuracy: 0.8845 - val_loss: 0.5371 - val_accuracy: 0.8868
Epoch 53/100
accuracy: 0.8845 - val_loss: 0.5391 - val_accuracy: 0.8842
Epoch 54/100
accuracy: 0.8847 - val_loss: 0.5388 - val_accuracy: 0.8832
Epoch 55/100
1349/1349 [============= - - 2s 1ms/step - loss: 0.5411 -
accuracy: 0.8851 - val_loss: 0.5368 - val_accuracy: 0.8862
Epoch 56/100
accuracy: 0.8852 - val_loss: 0.5377 - val_accuracy: 0.8832
Epoch 57/100
accuracy: 0.8839 - val_loss: 0.5382 - val_accuracy: 0.8829
Epoch 58/100
accuracy: 0.8839 - val_loss: 0.5385 - val_accuracy: 0.8859
Epoch 59/100
accuracy: 0.8853 - val_loss: 0.5376 - val_accuracy: 0.8842
Epoch 60/100
accuracy: 0.8842 - val_loss: 0.5385 - val_accuracy: 0.8863
Epoch 61/100
accuracy: 0.8850 - val_loss: 0.5371 - val_accuracy: 0.8857
Epoch 62/100
accuracy: 0.8858 - val_loss: 0.5371 - val_accuracy: 0.8860
Epoch 63/100
accuracy: 0.8850 - val_loss: 0.5382 - val_accuracy: 0.8859
Epoch 64/100
accuracy: 0.8844 - val_loss: 0.5379 - val_accuracy: 0.8848
```

```
Epoch 65/100
accuracy: 0.8839 - val_loss: 0.5381 - val_accuracy: 0.8852
Epoch 66/100
accuracy: 0.8835 - val_loss: 0.5364 - val_accuracy: 0.8866
Epoch 67/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5415 -
accuracy: 0.8837 - val_loss: 0.5382 - val_accuracy: 0.8838
Epoch 68/100
accuracy: 0.8851 - val_loss: 0.5397 - val_accuracy: 0.8804
Epoch 69/100
accuracy: 0.8848 - val_loss: 0.5372 - val_accuracy: 0.8863
Epoch 70/100
accuracy: 0.8854 - val_loss: 0.5369 - val_accuracy: 0.8860
Epoch 71/100
accuracy: 0.8853 - val_loss: 0.5368 - val_accuracy: 0.8874
Epoch 72/100
accuracy: 0.8840 - val_loss: 0.5371 - val_accuracy: 0.8852
Epoch 73/100
accuracy: 0.8852 - val_loss: 0.5372 - val_accuracy: 0.8856
Epoch 74/100
accuracy: 0.8838 - val_loss: 0.5364 - val_accuracy: 0.8864
Epoch 75/100
accuracy: 0.8834 - val_loss: 0.5368 - val_accuracy: 0.8859
Epoch 76/100
accuracy: 0.8851 - val_loss: 0.5364 - val_accuracy: 0.8871
Epoch 77/100
accuracy: 0.8847 - val_loss: 0.5386 - val_accuracy: 0.8827
Epoch 78/100
accuracy: 0.8847 - val_loss: 0.5368 - val_accuracy: 0.8856
Epoch 79/100
accuracy: 0.8841 - val_loss: 0.5384 - val_accuracy: 0.8817
Epoch 80/100
accuracy: 0.8855 - val_loss: 0.5365 - val_accuracy: 0.8863
```

```
Epoch 81/100
accuracy: 0.8848 - val_loss: 0.5382 - val_accuracy: 0.8840
Epoch 82/100
accuracy: 0.8856 - val_loss: 0.5366 - val_accuracy: 0.8858
Epoch 83/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5411 -
accuracy: 0.8844 - val_loss: 0.5369 - val_accuracy: 0.8864
Epoch 84/100
accuracy: 0.8848 - val_loss: 0.5370 - val_accuracy: 0.8859
Epoch 85/100
accuracy: 0.8836 - val_loss: 0.5364 - val_accuracy: 0.8858
Epoch 86/100
accuracy: 0.8844 - val_loss: 0.5368 - val_accuracy: 0.8868
Epoch 87/100
accuracy: 0.8839 - val_loss: 0.5367 - val_accuracy: 0.8864
Epoch 88/100
accuracy: 0.8836 - val_loss: 0.5372 - val_accuracy: 0.8867
Epoch 89/100
accuracy: 0.8852 - val_loss: 0.5368 - val_accuracy: 0.8870
Epoch 90/100
accuracy: 0.8835 - val_loss: 0.5378 - val_accuracy: 0.8873
Epoch 91/100
accuracy: 0.8853 - val_loss: 0.5371 - val_accuracy: 0.8851
Epoch 92/100
accuracy: 0.8864 - val_loss: 0.5386 - val_accuracy: 0.8866
Epoch 93/100
accuracy: 0.8855 - val_loss: 0.5367 - val_accuracy: 0.8870
Epoch 94/100
accuracy: 0.8855 - val_loss: 0.5367 - val_accuracy: 0.8858
Epoch 95/100
accuracy: 0.8849 - val_loss: 0.5370 - val_accuracy: 0.8867
Epoch 96/100
accuracy: 0.8847 - val_loss: 0.5366 - val_accuracy: 0.8875
```

```
Epoch 97/100
accuracy: 0.8852 - val_loss: 0.5380 - val_accuracy: 0.8836
Epoch 98/100
accuracy: 0.8846 - val_loss: 0.5373 - val_accuracy: 0.8863
Epoch 99/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5405 -
accuracy: 0.8850 - val_loss: 0.5361 - val_accuracy: 0.8878
Epoch 100/100
accuracy: 0.8856 - val_loss: 0.5387 - val_accuracy: 0.8842
Epoch 1/100
accuracy: 0.8845 - val_loss: 0.5363 - val_accuracy: 0.8868
Epoch 2/100
accuracy: 0.8851 - val_loss: 0.5375 - val_accuracy: 0.8874
Epoch 3/100
accuracy: 0.8859 - val_loss: 0.5364 - val_accuracy: 0.8867
Epoch 4/100
accuracy: 0.8847 - val_loss: 0.5370 - val_accuracy: 0.8866
Epoch 5/100
accuracy: 0.8838 - val_loss: 0.5371 - val_accuracy: 0.8846
Epoch 6/100
accuracy: 0.8859 - val_loss: 0.5381 - val_accuracy: 0.8837
Epoch 7/100
accuracy: 0.8845 - val_loss: 0.5368 - val_accuracy: 0.8869
Epoch 8/100
accuracy: 0.8848 - val_loss: 0.5365 - val_accuracy: 0.8850
Epoch 9/100
accuracy: 0.8857 - val_loss: 0.5369 - val_accuracy: 0.8844
Epoch 10/100
accuracy: 0.8849 - val_loss: 0.5367 - val_accuracy: 0.8860
Epoch 11/100
accuracy: 0.8833 - val_loss: 0.5376 - val_accuracy: 0.8882
Epoch 12/100
accuracy: 0.8862 - val_loss: 0.5367 - val_accuracy: 0.8868
```

```
Epoch 13/100
accuracy: 0.8864 - val_loss: 0.5373 - val_accuracy: 0.8843
Epoch 14/100
accuracy: 0.8857 - val_loss: 0.5384 - val_accuracy: 0.8844
Epoch 15/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5404 -
accuracy: 0.8852 - val_loss: 0.5370 - val_accuracy: 0.8865
Epoch 16/100
accuracy: 0.8856 - val_loss: 0.5389 - val_accuracy: 0.8847
Epoch 17/100
accuracy: 0.8849 - val_loss: 0.5370 - val_accuracy: 0.8843
Epoch 18/100
accuracy: 0.8845 - val_loss: 0.5371 - val_accuracy: 0.8869
Epoch 19/100
accuracy: 0.8852 - val_loss: 0.5369 - val_accuracy: 0.8863
Epoch 20/100
accuracy: 0.8853 - val_loss: 0.5360 - val_accuracy: 0.8876
Epoch 21/100
accuracy: 0.8852 - val_loss: 0.5370 - val_accuracy: 0.8843
Epoch 22/100
accuracy: 0.8852 - val_loss: 0.5372 - val_accuracy: 0.8837
Epoch 23/100
accuracy: 0.8851 - val_loss: 0.5377 - val_accuracy: 0.8840
Epoch 24/100
accuracy: 0.8859 - val_loss: 0.5363 - val_accuracy: 0.8860
Epoch 25/100
accuracy: 0.8858 - val_loss: 0.5368 - val_accuracy: 0.8866
Epoch 26/100
accuracy: 0.8850 - val_loss: 0.5375 - val_accuracy: 0.8828
Epoch 27/100
accuracy: 0.8857 - val_loss: 0.5363 - val_accuracy: 0.8876
Epoch 28/100
accuracy: 0.8856 - val_loss: 0.5378 - val_accuracy: 0.8830
```

```
Epoch 29/100
accuracy: 0.8845 - val_loss: 0.5371 - val_accuracy: 0.8856
Epoch 30/100
accuracy: 0.8848 - val_loss: 0.5365 - val_accuracy: 0.8872
Epoch 31/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5401 -
accuracy: 0.8858 - val_loss: 0.5363 - val_accuracy: 0.8858
Epoch 32/100
accuracy: 0.8852 - val_loss: 0.5375 - val_accuracy: 0.8845
Epoch 33/100
accuracy: 0.8851 - val_loss: 0.5372 - val_accuracy: 0.8835
Epoch 34/100
accuracy: 0.8853 - val_loss: 0.5378 - val_accuracy: 0.8819
Epoch 35/100
accuracy: 0.8846 - val_loss: 0.5367 - val_accuracy: 0.8861
Epoch 36/100
accuracy: 0.8844 - val_loss: 0.5368 - val_accuracy: 0.8859
Epoch 37/100
accuracy: 0.8853 - val_loss: 0.5362 - val_accuracy: 0.8854
Epoch 38/100
accuracy: 0.8853 - val_loss: 0.5366 - val_accuracy: 0.8853
Epoch 39/100
accuracy: 0.8848 - val_loss: 0.5367 - val_accuracy: 0.8853
Epoch 40/100
accuracy: 0.8857 - val_loss: 0.5370 - val_accuracy: 0.8867
Epoch 41/100
accuracy: 0.8853 - val_loss: 0.5372 - val_accuracy: 0.8853
Epoch 42/100
accuracy: 0.8865 - val_loss: 0.5371 - val_accuracy: 0.8857
Epoch 43/100
accuracy: 0.8845 - val_loss: 0.5369 - val_accuracy: 0.8863
Epoch 44/100
accuracy: 0.8848 - val_loss: 0.5361 - val_accuracy: 0.8881
```

```
Epoch 45/100
accuracy: 0.8862 - val_loss: 0.5370 - val_accuracy: 0.8830
Epoch 46/100
accuracy: 0.8856 - val_loss: 0.5361 - val_accuracy: 0.8856
Epoch 47/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5399 -
accuracy: 0.8866 - val_loss: 0.5360 - val_accuracy: 0.8884
Epoch 48/100
accuracy: 0.8865 - val_loss: 0.5360 - val_accuracy: 0.8873
Epoch 49/100
accuracy: 0.8859 - val_loss: 0.5360 - val_accuracy: 0.8872
Epoch 50/100
accuracy: 0.8861 - val_loss: 0.5364 - val_accuracy: 0.8880
Epoch 51/100
accuracy: 0.8868 - val_loss: 0.5366 - val_accuracy: 0.8855
Epoch 52/100
accuracy: 0.8860 - val_loss: 0.5365 - val_accuracy: 0.8872
Epoch 53/100
accuracy: 0.8859 - val_loss: 0.5362 - val_accuracy: 0.8868
Epoch 54/100
accuracy: 0.8864 - val_loss: 0.5372 - val_accuracy: 0.8838
Epoch 55/100
accuracy: 0.8862 - val_loss: 0.5365 - val_accuracy: 0.8848
Epoch 56/100
accuracy: 0.8857 - val_loss: 0.5359 - val_accuracy: 0.8875
Epoch 57/100
accuracy: 0.8864 - val_loss: 0.5369 - val_accuracy: 0.8874
Epoch 58/100
accuracy: 0.8861 - val_loss: 0.5367 - val_accuracy: 0.8863
Epoch 59/100
accuracy: 0.8855 - val_loss: 0.5365 - val_accuracy: 0.8867
Epoch 60/100
accuracy: 0.8860 - val_loss: 0.5372 - val_accuracy: 0.8854
```

```
Epoch 61/100
accuracy: 0.8857 - val_loss: 0.5354 - val_accuracy: 0.8883
Epoch 62/100
accuracy: 0.8867 - val_loss: 0.5384 - val_accuracy: 0.8843
Epoch 63/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5398 -
accuracy: 0.8866 - val_loss: 0.5353 - val_accuracy: 0.8878
Epoch 64/100
accuracy: 0.8855 - val_loss: 0.5356 - val_accuracy: 0.8875
Epoch 65/100
accuracy: 0.8864 - val_loss: 0.5358 - val_accuracy: 0.8885
Epoch 66/100
accuracy: 0.8853 - val_loss: 0.5363 - val_accuracy: 0.8852
Epoch 67/100
accuracy: 0.8856 - val_loss: 0.5363 - val_accuracy: 0.8849
Epoch 68/100
accuracy: 0.8861 - val_loss: 0.5368 - val_accuracy: 0.8867
Epoch 69/100
accuracy: 0.8871 - val_loss: 0.5355 - val_accuracy: 0.8892
Epoch 70/100
accuracy: 0.8871 - val_loss: 0.5372 - val_accuracy: 0.8859
Epoch 71/100
accuracy: 0.8857 - val_loss: 0.5364 - val_accuracy: 0.8848
Epoch 72/100
accuracy: 0.8873 - val_loss: 0.5350 - val_accuracy: 0.8890
Epoch 73/100
accuracy: 0.8865 - val_loss: 0.5347 - val_accuracy: 0.8895
Epoch 74/100
accuracy: 0.8872 - val_loss: 0.5351 - val_accuracy: 0.8890
Epoch 75/100
accuracy: 0.8872 - val_loss: 0.5355 - val_accuracy: 0.8886
Epoch 76/100
accuracy: 0.8880 - val_loss: 0.5359 - val_accuracy: 0.8856
```

```
Epoch 77/100
accuracy: 0.8871 - val_loss: 0.5359 - val_accuracy: 0.8872
Epoch 78/100
accuracy: 0.8876 - val_loss: 0.5361 - val_accuracy: 0.8884
Epoch 79/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5397 -
accuracy: 0.8869 - val_loss: 0.5361 - val_accuracy: 0.8865
Epoch 80/100
accuracy: 0.8877 - val_loss: 0.5354 - val_accuracy: 0.8899
Epoch 81/100
accuracy: 0.8883 - val_loss: 0.5352 - val_accuracy: 0.8903
Epoch 82/100
accuracy: 0.8868 - val_loss: 0.5356 - val_accuracy: 0.8886
Epoch 83/100
accuracy: 0.8876 - val_loss: 0.5357 - val_accuracy: 0.8884
Epoch 84/100
accuracy: 0.8877 - val_loss: 0.5353 - val_accuracy: 0.8893
Epoch 85/100
accuracy: 0.8871 - val_loss: 0.5367 - val_accuracy: 0.8845
Epoch 86/100
accuracy: 0.8871 - val_loss: 0.5361 - val_accuracy: 0.8888
Epoch 87/100
accuracy: 0.8854 - val_loss: 0.5356 - val_accuracy: 0.8899
Epoch 88/100
accuracy: 0.8870 - val_loss: 0.5359 - val_accuracy: 0.8874
Epoch 89/100
accuracy: 0.8870 - val_loss: 0.5366 - val_accuracy: 0.8889
Epoch 90/100
accuracy: 0.8865 - val_loss: 0.5350 - val_accuracy: 0.8882
Epoch 91/100
accuracy: 0.8868 - val_loss: 0.5360 - val_accuracy: 0.8882
Epoch 92/100
accuracy: 0.8878 - val_loss: 0.5351 - val_accuracy: 0.8891
```

```
Epoch 93/100
accuracy: 0.8871 - val_loss: 0.5359 - val_accuracy: 0.8881
Epoch 94/100
accuracy: 0.8871 - val_loss: 0.5359 - val_accuracy: 0.8877
Epoch 95/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5389 -
accuracy: 0.8889 - val_loss: 0.5360 - val_accuracy: 0.8862
Epoch 96/100
accuracy: 0.8878 - val_loss: 0.5347 - val_accuracy: 0.8896
Epoch 97/100
accuracy: 0.8887 - val_loss: 0.5367 - val_accuracy: 0.8888
Epoch 98/100
accuracy: 0.8871 - val_loss: 0.5356 - val_accuracy: 0.8897
Epoch 99/100
accuracy: 0.8862 - val_loss: 0.5351 - val_accuracy: 0.8882
Epoch 100/100
accuracy: 0.8873 - val_loss: 0.5351 - val_accuracy: 0.8897
Epoch 1/100
accuracy: 0.8876 - val_loss: 0.5354 - val_accuracy: 0.8884
Epoch 2/100
accuracy: 0.8864 - val_loss: 0.5355 - val_accuracy: 0.8877
Epoch 3/100
accuracy: 0.8868 - val_loss: 0.5353 - val_accuracy: 0.8888
Epoch 4/100
accuracy: 0.8860 - val_loss: 0.5355 - val_accuracy: 0.8899
Epoch 5/100
accuracy: 0.8864 - val_loss: 0.5347 - val_accuracy: 0.8889
Epoch 6/100
accuracy: 0.8867 - val_loss: 0.5344 - val_accuracy: 0.8896
Epoch 7/100
accuracy: 0.8884 - val_loss: 0.5362 - val_accuracy: 0.8912
Epoch 8/100
accuracy: 0.8882 - val_loss: 0.5344 - val_accuracy: 0.8911
```

```
Epoch 9/100
accuracy: 0.8879 - val_loss: 0.5344 - val_accuracy: 0.8907
Epoch 10/100
accuracy: 0.8878 - val_loss: 0.5352 - val_accuracy: 0.8890
Epoch 11/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8878 - val_loss: 0.5363 - val_accuracy: 0.8884
Epoch 12/100
accuracy: 0.8877 - val_loss: 0.5344 - val_accuracy: 0.8890
Epoch 13/100
accuracy: 0.8887 - val_loss: 0.5348 - val_accuracy: 0.8912
Epoch 14/100
accuracy: 0.8874 - val_loss: 0.5347 - val_accuracy: 0.8904
Epoch 15/100
accuracy: 0.8886 - val_loss: 0.5362 - val_accuracy: 0.8869
Epoch 16/100
accuracy: 0.8881 - val_loss: 0.5344 - val_accuracy: 0.8911
Epoch 17/100
accuracy: 0.8878 - val_loss: 0.5347 - val_accuracy: 0.8917
Epoch 18/100
accuracy: 0.8890 - val_loss: 0.5357 - val_accuracy: 0.8883
Epoch 19/100
accuracy: 0.8879 - val_loss: 0.5341 - val_accuracy: 0.8904
Epoch 20/100
accuracy: 0.8885 - val_loss: 0.5348 - val_accuracy: 0.8894
Epoch 21/100
accuracy: 0.8883 - val_loss: 0.5352 - val_accuracy: 0.8893
Epoch 22/100
accuracy: 0.8874 - val_loss: 0.5354 - val_accuracy: 0.8916
Epoch 23/100
accuracy: 0.8877 - val_loss: 0.5349 - val_accuracy: 0.8907
Epoch 24/100
accuracy: 0.8895 - val_loss: 0.5345 - val_accuracy: 0.8909
```

```
Epoch 25/100
accuracy: 0.8872 - val_loss: 0.5358 - val_accuracy: 0.8898
Epoch 26/100
accuracy: 0.8868 - val_loss: 0.5348 - val_accuracy: 0.8887
Epoch 27/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5394 -
accuracy: 0.8867 - val_loss: 0.5357 - val_accuracy: 0.8904
Epoch 28/100
accuracy: 0.8879 - val_loss: 0.5346 - val_accuracy: 0.8907
Epoch 29/100
accuracy: 0.8887 - val_loss: 0.5343 - val_accuracy: 0.8915
Epoch 30/100
accuracy: 0.8888 - val_loss: 0.5350 - val_accuracy: 0.8904
Epoch 31/100
accuracy: 0.8876 - val_loss: 0.5341 - val_accuracy: 0.8911
Epoch 32/100
accuracy: 0.8874 - val_loss: 0.5345 - val_accuracy: 0.8904
Epoch 33/100
accuracy: 0.8873 - val_loss: 0.5345 - val_accuracy: 0.8903
Epoch 34/100
accuracy: 0.8877 - val_loss: 0.5367 - val_accuracy: 0.8921
Epoch 35/100
accuracy: 0.8875 - val_loss: 0.5344 - val_accuracy: 0.8918
Epoch 36/100
accuracy: 0.8888 - val_loss: 0.5355 - val_accuracy: 0.8909
Epoch 37/100
accuracy: 0.8876 - val_loss: 0.5358 - val_accuracy: 0.8906
Epoch 38/100
accuracy: 0.8891 - val_loss: 0.5350 - val_accuracy: 0.8917
Epoch 39/100
accuracy: 0.8871 - val_loss: 0.5354 - val_accuracy: 0.8913
Epoch 40/100
accuracy: 0.8876 - val_loss: 0.5355 - val_accuracy: 0.8910
```

```
Epoch 41/100
accuracy: 0.8876 - val_loss: 0.5341 - val_accuracy: 0.8911
Epoch 42/100
accuracy: 0.8880 - val_loss: 0.5343 - val_accuracy: 0.8896
Epoch 43/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5381 -
accuracy: 0.8891 - val_loss: 0.5346 - val_accuracy: 0.8919
Epoch 44/100
accuracy: 0.8885 - val_loss: 0.5349 - val_accuracy: 0.8922
Epoch 45/100
accuracy: 0.8884 - val_loss: 0.5345 - val_accuracy: 0.8889
Epoch 46/100
accuracy: 0.8880 - val_loss: 0.5350 - val_accuracy: 0.8892
Epoch 47/100
accuracy: 0.8880 - val_loss: 0.5347 - val_accuracy: 0.8908
Epoch 48/100
accuracy: 0.8882 - val_loss: 0.5341 - val_accuracy: 0.8920
Epoch 49/100
accuracy: 0.8888 - val_loss: 0.5342 - val_accuracy: 0.8900
Epoch 50/100
accuracy: 0.8873 - val_loss: 0.5351 - val_accuracy: 0.8909
Epoch 51/100
accuracy: 0.8881 - val_loss: 0.5347 - val_accuracy: 0.8901
Epoch 52/100
accuracy: 0.8868 - val_loss: 0.5350 - val_accuracy: 0.8890
Epoch 53/100
accuracy: 0.8870 - val_loss: 0.5347 - val_accuracy: 0.8911
Epoch 54/100
accuracy: 0.8874 - val_loss: 0.5345 - val_accuracy: 0.8909
Epoch 55/100
accuracy: 0.8891 - val_loss: 0.5340 - val_accuracy: 0.8924
Epoch 56/100
accuracy: 0.8871 - val_loss: 0.5345 - val_accuracy: 0.8910
```

```
Epoch 57/100
accuracy: 0.8884 - val_loss: 0.5343 - val_accuracy: 0.8917
Epoch 58/100
accuracy: 0.8876 - val_loss: 0.5364 - val_accuracy: 0.8875
Epoch 59/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5399 -
accuracy: 0.8868 - val_loss: 0.5354 - val_accuracy: 0.8897
Epoch 60/100
accuracy: 0.8879 - val_loss: 0.5344 - val_accuracy: 0.8920
Epoch 61/100
accuracy: 0.8886 - val_loss: 0.5345 - val_accuracy: 0.8894
Epoch 62/100
accuracy: 0.8881 - val_loss: 0.5342 - val_accuracy: 0.8928
Epoch 63/100
accuracy: 0.8893 - val_loss: 0.5341 - val_accuracy: 0.8931
Epoch 64/100
accuracy: 0.8886 - val_loss: 0.5344 - val_accuracy: 0.8905
Epoch 65/100
accuracy: 0.8885 - val_loss: 0.5360 - val_accuracy: 0.8898
Epoch 66/100
accuracy: 0.8891 - val_loss: 0.5356 - val_accuracy: 0.8925
Epoch 67/100
accuracy: 0.8899 - val_loss: 0.5365 - val_accuracy: 0.8887
Epoch 68/100
accuracy: 0.8896 - val_loss: 0.5357 - val_accuracy: 0.8932
Epoch 69/100
accuracy: 0.8889 - val_loss: 0.5347 - val_accuracy: 0.8927
Epoch 70/100
accuracy: 0.8884 - val_loss: 0.5350 - val_accuracy: 0.8892
Epoch 71/100
accuracy: 0.8877 - val_loss: 0.5350 - val_accuracy: 0.8892
Epoch 72/100
accuracy: 0.8883 - val_loss: 0.5342 - val_accuracy: 0.8907
```

```
Epoch 73/100
accuracy: 0.8885 - val_loss: 0.5342 - val_accuracy: 0.8909
Epoch 74/100
accuracy: 0.8892 - val_loss: 0.5344 - val_accuracy: 0.8921
Epoch 75/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5394 -
accuracy: 0.8884 - val_loss: 0.5346 - val_accuracy: 0.8921
Epoch 76/100
accuracy: 0.8884 - val_loss: 0.5345 - val_accuracy: 0.8913
Epoch 77/100
accuracy: 0.8883 - val_loss: 0.5341 - val_accuracy: 0.8927
Epoch 78/100
accuracy: 0.8875 - val_loss: 0.5346 - val_accuracy: 0.8898
Epoch 79/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5401 -
accuracy: 0.8875 - val_loss: 0.5354 - val_accuracy: 0.8897
Epoch 80/100
accuracy: 0.8882 - val_loss: 0.5365 - val_accuracy: 0.8873
Epoch 81/100
accuracy: 0.8870 - val_loss: 0.5344 - val_accuracy: 0.8903
Epoch 82/100
accuracy: 0.8882 - val_loss: 0.5358 - val_accuracy: 0.8910
Epoch 83/100
accuracy: 0.8875 - val_loss: 0.5346 - val_accuracy: 0.8914
Epoch 84/100
accuracy: 0.8872 - val_loss: 0.5344 - val_accuracy: 0.8898
Epoch 85/100
accuracy: 0.8874 - val_loss: 0.5346 - val_accuracy: 0.8910
Epoch 86/100
accuracy: 0.8890 - val_loss: 0.5346 - val_accuracy: 0.8886
Epoch 87/100
accuracy: 0.8886 - val_loss: 0.5344 - val_accuracy: 0.8905
Epoch 88/100
accuracy: 0.8872 - val_loss: 0.5361 - val_accuracy: 0.8882
```

```
Epoch 89/100
accuracy: 0.8876 - val_loss: 0.5356 - val_accuracy: 0.8876
Epoch 90/100
accuracy: 0.8887 - val_loss: 0.5358 - val_accuracy: 0.8879
Epoch 91/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5398 -
accuracy: 0.8882 - val_loss: 0.5355 - val_accuracy: 0.8896
Epoch 92/100
accuracy: 0.8873 - val_loss: 0.5345 - val_accuracy: 0.8906
Epoch 93/100
accuracy: 0.8890 - val_loss: 0.5352 - val_accuracy: 0.8884
Epoch 94/100
accuracy: 0.8887 - val_loss: 0.5343 - val_accuracy: 0.8902
Epoch 95/100
accuracy: 0.8884 - val_loss: 0.5345 - val_accuracy: 0.8916
Epoch 96/100
accuracy: 0.8892 - val_loss: 0.5349 - val_accuracy: 0.8912
Epoch 97/100
accuracy: 0.8888 - val_loss: 0.5349 - val_accuracy: 0.8895
Epoch 98/100
accuracy: 0.8896 - val_loss: 0.5357 - val_accuracy: 0.8900
Epoch 99/100
accuracy: 0.8886 - val_loss: 0.5344 - val_accuracy: 0.8893
Epoch 100/100
accuracy: 0.8891 - val_loss: 0.5343 - val_accuracy: 0.8922
Epoch 1/100
accuracy: 0.8888 - val_loss: 0.5377 - val_accuracy: 0.8893
Epoch 2/100
accuracy: 0.8875 - val_loss: 0.5350 - val_accuracy: 0.8914
Epoch 3/100
accuracy: 0.8881 - val_loss: 0.5343 - val_accuracy: 0.8903
Epoch 4/100
accuracy: 0.8881 - val_loss: 0.5353 - val_accuracy: 0.8895
```

```
Epoch 5/100
accuracy: 0.8885 - val_loss: 0.5342 - val_accuracy: 0.8917
Epoch 6/100
accuracy: 0.8894 - val_loss: 0.5351 - val_accuracy: 0.8915
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5396 -
accuracy: 0.8882 - val_loss: 0.5342 - val_accuracy: 0.8902
Epoch 8/100
accuracy: 0.8879 - val_loss: 0.5342 - val_accuracy: 0.8902
Epoch 9/100
accuracy: 0.8884 - val_loss: 0.5350 - val_accuracy: 0.8889
Epoch 10/100
accuracy: 0.8881 - val_loss: 0.5349 - val_accuracy: 0.8901
Epoch 11/100
accuracy: 0.8884 - val_loss: 0.5367 - val_accuracy: 0.8879
Epoch 12/100
accuracy: 0.8867 - val_loss: 0.5360 - val_accuracy: 0.8878
Epoch 13/100
accuracy: 0.8871 - val_loss: 0.5349 - val_accuracy: 0.8907
Epoch 14/100
accuracy: 0.8887 - val_loss: 0.5348 - val_accuracy: 0.8901
Epoch 15/100
accuracy: 0.8877 - val_loss: 0.5346 - val_accuracy: 0.8893
Epoch 16/100
accuracy: 0.8881 - val_loss: 0.5355 - val_accuracy: 0.8924
Epoch 17/100
accuracy: 0.8899 - val_loss: 0.5349 - val_accuracy: 0.8906
Epoch 18/100
accuracy: 0.8898 - val_loss: 0.5358 - val_accuracy: 0.8916
Epoch 19/100
accuracy: 0.8902 - val_loss: 0.5358 - val_accuracy: 0.8938
Epoch 20/100
accuracy: 0.8902 - val_loss: 0.5350 - val_accuracy: 0.8928
```

```
Epoch 21/100
accuracy: 0.8891 - val_loss: 0.5357 - val_accuracy: 0.8929
Epoch 22/100
accuracy: 0.8902 - val_loss: 0.5353 - val_accuracy: 0.8905
Epoch 23/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5382 -
accuracy: 0.8897 - val_loss: 0.5353 - val_accuracy: 0.8904
Epoch 24/100
accuracy: 0.8903 - val_loss: 0.5363 - val_accuracy: 0.8922
Epoch 25/100
accuracy: 0.8893 - val_loss: 0.5366 - val_accuracy: 0.8909
Epoch 26/100
accuracy: 0.8886 - val_loss: 0.5345 - val_accuracy: 0.8912
Epoch 27/100
accuracy: 0.8888 - val_loss: 0.5344 - val_accuracy: 0.8922
Epoch 28/100
accuracy: 0.8881 - val_loss: 0.5349 - val_accuracy: 0.8901
Epoch 29/100
accuracy: 0.8869 - val_loss: 0.5356 - val_accuracy: 0.8889
Epoch 30/100
accuracy: 0.8881 - val_loss: 0.5348 - val_accuracy: 0.8900
Epoch 31/100
accuracy: 0.8881 - val_loss: 0.5353 - val_accuracy: 0.8928
Epoch 32/100
accuracy: 0.8893 - val_loss: 0.5347 - val_accuracy: 0.8905
Epoch 33/100
accuracy: 0.8908 - val_loss: 0.5352 - val_accuracy: 0.8900
Epoch 34/100
accuracy: 0.8892 - val_loss: 0.5342 - val_accuracy: 0.8913
Epoch 35/100
accuracy: 0.8889 - val_loss: 0.5341 - val_accuracy: 0.8910
Epoch 36/100
accuracy: 0.8879 - val_loss: 0.5347 - val_accuracy: 0.8919
```

```
Epoch 37/100
accuracy: 0.8885 - val_loss: 0.5352 - val_accuracy: 0.8939
Epoch 38/100
accuracy: 0.8899 - val_loss: 0.5360 - val_accuracy: 0.8902
Epoch 39/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5386 -
accuracy: 0.8892 - val_loss: 0.5349 - val_accuracy: 0.8922
Epoch 40/100
accuracy: 0.8894 - val_loss: 0.5343 - val_accuracy: 0.8916
Epoch 41/100
accuracy: 0.8888 - val_loss: 0.5342 - val_accuracy: 0.8919
Epoch 42/100
accuracy: 0.8879 - val_loss: 0.5355 - val_accuracy: 0.8924
Epoch 43/100
accuracy: 0.8896 - val_loss: 0.5345 - val_accuracy: 0.8909
Epoch 44/100
accuracy: 0.8873 - val_loss: 0.5349 - val_accuracy: 0.8911
Epoch 45/100
accuracy: 0.8875 - val_loss: 0.5350 - val_accuracy: 0.8901
Epoch 46/100
accuracy: 0.8894 - val_loss: 0.5344 - val_accuracy: 0.8917
Epoch 47/100
accuracy: 0.8881 - val_loss: 0.5351 - val_accuracy: 0.8931
Epoch 48/100
accuracy: 0.8884 - val_loss: 0.5345 - val_accuracy: 0.8898
Epoch 49/100
accuracy: 0.8881 - val_loss: 0.5351 - val_accuracy: 0.8894
Epoch 50/100
accuracy: 0.8878 - val_loss: 0.5344 - val_accuracy: 0.8905
Epoch 51/100
accuracy: 0.8878 - val_loss: 0.5349 - val_accuracy: 0.8887
Epoch 52/100
accuracy: 0.8884 - val_loss: 0.5350 - val_accuracy: 0.8907
```

```
Epoch 53/100
accuracy: 0.8894 - val_loss: 0.5349 - val_accuracy: 0.8918
Epoch 54/100
accuracy: 0.8892 - val_loss: 0.5348 - val_accuracy: 0.8933
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8888 - val_loss: 0.5356 - val_accuracy: 0.8906
Epoch 56/100
accuracy: 0.8896 - val_loss: 0.5347 - val_accuracy: 0.8917
Epoch 57/100
accuracy: 0.8882 - val_loss: 0.5349 - val_accuracy: 0.8892
Epoch 58/100
accuracy: 0.8879 - val_loss: 0.5347 - val_accuracy: 0.8911
Epoch 59/100
1349/1349 [============= - - 2s 1ms/step - loss: 0.5382 -
accuracy: 0.8885 - val_loss: 0.5344 - val_accuracy: 0.8894
Epoch 60/100
accuracy: 0.8890 - val_loss: 0.5348 - val_accuracy: 0.8907
Epoch 61/100
accuracy: 0.8883 - val_loss: 0.5340 - val_accuracy: 0.8909
Epoch 62/100
accuracy: 0.8873 - val_loss: 0.5342 - val_accuracy: 0.8913
Epoch 63/100
accuracy: 0.8886 - val_loss: 0.5356 - val_accuracy: 0.8896
Epoch 64/100
accuracy: 0.8871 - val_loss: 0.5353 - val_accuracy: 0.8888
Epoch 65/100
accuracy: 0.8893 - val_loss: 0.5343 - val_accuracy: 0.8909
Epoch 66/100
accuracy: 0.8882 - val_loss: 0.5340 - val_accuracy: 0.8918
Epoch 67/100
accuracy: 0.8894 - val_loss: 0.5339 - val_accuracy: 0.8905
Epoch 68/100
accuracy: 0.8873 - val_loss: 0.5345 - val_accuracy: 0.8904
```

```
Epoch 69/100
accuracy: 0.8881 - val_loss: 0.5337 - val_accuracy: 0.8910
Epoch 70/100
accuracy: 0.8880 - val_loss: 0.5340 - val_accuracy: 0.8908
Epoch 71/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5388 -
accuracy: 0.8890 - val_loss: 0.5340 - val_accuracy: 0.8920
Epoch 72/100
accuracy: 0.8887 - val_loss: 0.5343 - val_accuracy: 0.8899
Epoch 73/100
accuracy: 0.8879 - val_loss: 0.5338 - val_accuracy: 0.8919
Epoch 74/100
accuracy: 0.8894 - val_loss: 0.5337 - val_accuracy: 0.8924
Epoch 75/100
accuracy: 0.8891 - val_loss: 0.5341 - val_accuracy: 0.8917
Epoch 76/100
accuracy: 0.8897 - val_loss: 0.5344 - val_accuracy: 0.8924
Epoch 77/100
accuracy: 0.8890 - val_loss: 0.5368 - val_accuracy: 0.8873
Epoch 78/100
accuracy: 0.8895 - val_loss: 0.5347 - val_accuracy: 0.8915
Epoch 79/100
accuracy: 0.8892 - val_loss: 0.5345 - val_accuracy: 0.8920
Epoch 80/100
accuracy: 0.8888 - val_loss: 0.5352 - val_accuracy: 0.8894
Epoch 81/100
accuracy: 0.8891 - val_loss: 0.5348 - val_accuracy: 0.8895
Epoch 82/100
accuracy: 0.8886 - val_loss: 0.5346 - val_accuracy: 0.8919
Epoch 83/100
accuracy: 0.8902 - val_loss: 0.5349 - val_accuracy: 0.8922
Epoch 84/100
accuracy: 0.8886 - val_loss: 0.5359 - val_accuracy: 0.8913
```

```
Epoch 85/100
accuracy: 0.8890 - val_loss: 0.5348 - val_accuracy: 0.8931
Epoch 86/100
accuracy: 0.8894 - val_loss: 0.5346 - val_accuracy: 0.8917
Epoch 87/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5388 -
accuracy: 0.8890 - val_loss: 0.5347 - val_accuracy: 0.8909
Epoch 88/100
accuracy: 0.8886 - val_loss: 0.5349 - val_accuracy: 0.8902
Epoch 89/100
accuracy: 0.8888 - val_loss: 0.5350 - val_accuracy: 0.8895
Epoch 90/100
accuracy: 0.8895 - val_loss: 0.5344 - val_accuracy: 0.8918
Epoch 91/100
accuracy: 0.8892 - val_loss: 0.5360 - val_accuracy: 0.8893
Epoch 92/100
accuracy: 0.8886 - val_loss: 0.5347 - val_accuracy: 0.8905
Epoch 93/100
accuracy: 0.8889 - val_loss: 0.5349 - val_accuracy: 0.8924
Epoch 94/100
accuracy: 0.8885 - val_loss: 0.5349 - val_accuracy: 0.8906
Epoch 95/100
accuracy: 0.8895 - val_loss: 0.5344 - val_accuracy: 0.8927
Epoch 96/100
accuracy: 0.8893 - val_loss: 0.5349 - val_accuracy: 0.8905
Epoch 97/100
accuracy: 0.8878 - val_loss: 0.5351 - val_accuracy: 0.8892
Epoch 98/100
accuracy: 0.8876 - val_loss: 0.5354 - val_accuracy: 0.8899
Epoch 99/100
accuracy: 0.8868 - val_loss: 0.5350 - val_accuracy: 0.8870
Epoch 100/100
accuracy: 0.8871 - val_loss: 0.5341 - val_accuracy: 0.8910
```

```
Epoch 1/100
accuracy: 0.8877 - val_loss: 0.5339 - val_accuracy: 0.8905
Epoch 2/100
accuracy: 0.8883 - val_loss: 0.5339 - val_accuracy: 0.8896
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8883 - val_loss: 0.5347 - val_accuracy: 0.8922
Epoch 4/100
accuracy: 0.8888 - val_loss: 0.5345 - val_accuracy: 0.8907
Epoch 5/100
accuracy: 0.8883 - val_loss: 0.5347 - val_accuracy: 0.8901
Epoch 6/100
accuracy: 0.8877 - val_loss: 0.5342 - val_accuracy: 0.8902
Epoch 7/100
accuracy: 0.8890 - val_loss: 0.5344 - val_accuracy: 0.8904
Epoch 8/100
accuracy: 0.8890 - val_loss: 0.5339 - val_accuracy: 0.8908
Epoch 9/100
accuracy: 0.8879 - val_loss: 0.5343 - val_accuracy: 0.8900
Epoch 10/100
accuracy: 0.8893 - val_loss: 0.5342 - val_accuracy: 0.8909
Epoch 11/100
accuracy: 0.8884 - val_loss: 0.5337 - val_accuracy: 0.8913
Epoch 12/100
accuracy: 0.8881 - val_loss: 0.5345 - val_accuracy: 0.8907
Epoch 13/100
accuracy: 0.8894 - val_loss: 0.5342 - val_accuracy: 0.8894
Epoch 14/100
accuracy: 0.8887 - val_loss: 0.5362 - val_accuracy: 0.8871
Epoch 15/100
accuracy: 0.8887 - val_loss: 0.5342 - val_accuracy: 0.8898
Epoch 16/100
accuracy: 0.8878 - val_loss: 0.5348 - val_accuracy: 0.8876
```

```
Epoch 17/100
accuracy: 0.8881 - val_loss: 0.5339 - val_accuracy: 0.8923
Epoch 18/100
accuracy: 0.8885 - val_loss: 0.5337 - val_accuracy: 0.8915
Epoch 19/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5387 -
accuracy: 0.8879 - val_loss: 0.5337 - val_accuracy: 0.8904
Epoch 20/100
accuracy: 0.8891 - val_loss: 0.5356 - val_accuracy: 0.8901
Epoch 21/100
accuracy: 0.8893 - val_loss: 0.5344 - val_accuracy: 0.8928
Epoch 22/100
accuracy: 0.8886 - val_loss: 0.5355 - val_accuracy: 0.8891
Epoch 23/100
accuracy: 0.8897 - val_loss: 0.5348 - val_accuracy: 0.8914
Epoch 24/100
accuracy: 0.8899 - val_loss: 0.5353 - val_accuracy: 0.8917
Epoch 25/100
accuracy: 0.8889 - val_loss: 0.5344 - val_accuracy: 0.8913
Epoch 26/100
accuracy: 0.8893 - val_loss: 0.5345 - val_accuracy: 0.8922
Epoch 27/100
accuracy: 0.8903 - val_loss: 0.5349 - val_accuracy: 0.8896
Epoch 28/100
accuracy: 0.8876 - val_loss: 0.5339 - val_accuracy: 0.8912
Epoch 29/100
accuracy: 0.8896 - val_loss: 0.5345 - val_accuracy: 0.8903
Epoch 30/100
accuracy: 0.8891 - val_loss: 0.5348 - val_accuracy: 0.8940
Epoch 31/100
accuracy: 0.8900 - val_loss: 0.5342 - val_accuracy: 0.8914
Epoch 32/100
accuracy: 0.8887 - val_loss: 0.5348 - val_accuracy: 0.8913
```

```
Epoch 33/100
accuracy: 0.8896 - val_loss: 0.5348 - val_accuracy: 0.8913
Epoch 34/100
accuracy: 0.8892 - val_loss: 0.5355 - val_accuracy: 0.8884
Epoch 35/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5383 -
accuracy: 0.8888 - val_loss: 0.5344 - val_accuracy: 0.8917
Epoch 36/100
accuracy: 0.8893 - val_loss: 0.5349 - val_accuracy: 0.8918
Epoch 37/100
accuracy: 0.8895 - val_loss: 0.5345 - val_accuracy: 0.8917
Epoch 38/100
accuracy: 0.8903 - val_loss: 0.5340 - val_accuracy: 0.8929
Epoch 39/100
accuracy: 0.8893 - val_loss: 0.5344 - val_accuracy: 0.8919
Epoch 40/100
accuracy: 0.8890 - val_loss: 0.5349 - val_accuracy: 0.8907
Epoch 41/100
accuracy: 0.8896 - val_loss: 0.5354 - val_accuracy: 0.8889
Epoch 42/100
accuracy: 0.8877 - val_loss: 0.5343 - val_accuracy: 0.8916
Epoch 43/100
accuracy: 0.8883 - val_loss: 0.5358 - val_accuracy: 0.8930
Epoch 44/100
accuracy: 0.8897 - val_loss: 0.5343 - val_accuracy: 0.8918
Epoch 45/100
accuracy: 0.8896 - val_loss: 0.5343 - val_accuracy: 0.8930
Epoch 46/100
accuracy: 0.8901 - val_loss: 0.5341 - val_accuracy: 0.8923
Epoch 47/100
accuracy: 0.8895 - val_loss: 0.5346 - val_accuracy: 0.8918
Epoch 48/100
accuracy: 0.8900 - val_loss: 0.5345 - val_accuracy: 0.8916
```

```
Epoch 49/100
accuracy: 0.8887 - val_loss: 0.5344 - val_accuracy: 0.8920
Epoch 50/100
accuracy: 0.8888 - val_loss: 0.5350 - val_accuracy: 0.8906
Epoch 51/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5381 -
accuracy: 0.8891 - val_loss: 0.5345 - val_accuracy: 0.8900
Epoch 52/100
accuracy: 0.8887 - val_loss: 0.5345 - val_accuracy: 0.8919
Epoch 53/100
accuracy: 0.8884 - val_loss: 0.5343 - val_accuracy: 0.8921
Epoch 54/100
accuracy: 0.8895 - val_loss: 0.5351 - val_accuracy: 0.8898
Epoch 55/100
accuracy: 0.8898 - val_loss: 0.5370 - val_accuracy: 0.8868
Epoch 56/100
accuracy: 0.8899 - val_loss: 0.5364 - val_accuracy: 0.8894
Epoch 57/100
accuracy: 0.8897 - val_loss: 0.5342 - val_accuracy: 0.8931
Epoch 58/100
accuracy: 0.8892 - val_loss: 0.5345 - val_accuracy: 0.8925
Epoch 59/100
accuracy: 0.8893 - val_loss: 0.5343 - val_accuracy: 0.8911
Epoch 60/100
accuracy: 0.8884 - val_loss: 0.5343 - val_accuracy: 0.8924
Epoch 61/100
accuracy: 0.8879 - val_loss: 0.5340 - val_accuracy: 0.8902
Epoch 62/100
accuracy: 0.8884 - val_loss: 0.5346 - val_accuracy: 0.8890
Epoch 63/100
accuracy: 0.8876 - val_loss: 0.5344 - val_accuracy: 0.8907
Epoch 64/100
accuracy: 0.8877 - val_loss: 0.5340 - val_accuracy: 0.8909
```

```
Epoch 65/100
accuracy: 0.8885 - val_loss: 0.5337 - val_accuracy: 0.8905
Epoch 66/100
accuracy: 0.8881 - val_loss: 0.5342 - val_accuracy: 0.8916
Epoch 67/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5397 -
accuracy: 0.8886 - val_loss: 0.5347 - val_accuracy: 0.8923
Epoch 68/100
accuracy: 0.8897 - val_loss: 0.5349 - val_accuracy: 0.8918
Epoch 69/100
accuracy: 0.8886 - val_loss: 0.5342 - val_accuracy: 0.8914
Epoch 70/100
accuracy: 0.8889 - val_loss: 0.5342 - val_accuracy: 0.8920
Epoch 71/100
1349/1349 [============= - - 2s 1ms/step - loss: 0.5397 -
accuracy: 0.8886 - val_loss: 0.5347 - val_accuracy: 0.8924
Epoch 72/100
accuracy: 0.8897 - val_loss: 0.5349 - val_accuracy: 0.8911
Epoch 73/100
accuracy: 0.8885 - val_loss: 0.5341 - val_accuracy: 0.8896
Epoch 74/100
accuracy: 0.8864 - val_loss: 0.5352 - val_accuracy: 0.8918
Epoch 75/100
accuracy: 0.8879 - val_loss: 0.5344 - val_accuracy: 0.8904
Epoch 76/100
accuracy: 0.8882 - val_loss: 0.5357 - val_accuracy: 0.8882
Epoch 77/100
accuracy: 0.8874 - val_loss: 0.5345 - val_accuracy: 0.8915
Epoch 78/100
accuracy: 0.8894 - val_loss: 0.5341 - val_accuracy: 0.8912
Epoch 79/100
accuracy: 0.8880 - val_loss: 0.5347 - val_accuracy: 0.8915
Epoch 80/100
accuracy: 0.8887 - val_loss: 0.5361 - val_accuracy: 0.8880
```

```
Epoch 81/100
accuracy: 0.8898 - val_loss: 0.5343 - val_accuracy: 0.8909
Epoch 82/100
accuracy: 0.8881 - val_loss: 0.5364 - val_accuracy: 0.8886
Epoch 83/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5383 -
accuracy: 0.8888 - val_loss: 0.5342 - val_accuracy: 0.8903
Epoch 84/100
accuracy: 0.8891 - val_loss: 0.5340 - val_accuracy: 0.8898
Epoch 85/100
accuracy: 0.8880 - val_loss: 0.5343 - val_accuracy: 0.8893
Epoch 86/100
accuracy: 0.8885 - val_loss: 0.5340 - val_accuracy: 0.8892
Epoch 87/100
accuracy: 0.8889 - val_loss: 0.5338 - val_accuracy: 0.8913
Epoch 88/100
accuracy: 0.8887 - val_loss: 0.5343 - val_accuracy: 0.8912
Epoch 89/100
accuracy: 0.8880 - val_loss: 0.5344 - val_accuracy: 0.8909
Epoch 90/100
accuracy: 0.8878 - val_loss: 0.5343 - val_accuracy: 0.8906
Epoch 91/100
accuracy: 0.8891 - val_loss: 0.5350 - val_accuracy: 0.8918
Epoch 92/100
accuracy: 0.8888 - val_loss: 0.5345 - val_accuracy: 0.8900
Epoch 93/100
accuracy: 0.8884 - val_loss: 0.5339 - val_accuracy: 0.8898
Epoch 94/100
accuracy: 0.8891 - val_loss: 0.5339 - val_accuracy: 0.8916
Epoch 95/100
accuracy: 0.8893 - val_loss: 0.5346 - val_accuracy: 0.8903
Epoch 96/100
accuracy: 0.8880 - val_loss: 0.5337 - val_accuracy: 0.8908
```

```
Epoch 97/100
accuracy: 0.8872 - val_loss: 0.5345 - val_accuracy: 0.8916
Epoch 98/100
accuracy: 0.8883 - val_loss: 0.5349 - val_accuracy: 0.8905
Epoch 99/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5386 -
accuracy: 0.8886 - val_loss: 0.5348 - val_accuracy: 0.8891
Epoch 100/100
accuracy: 0.8882 - val_loss: 0.5341 - val_accuracy: 0.8913
Epoch 1/100
accuracy: 0.8882 - val_loss: 0.5346 - val_accuracy: 0.8886
Epoch 2/100
accuracy: 0.8883 - val_loss: 0.5342 - val_accuracy: 0.8898
Epoch 3/100
accuracy: 0.8891 - val_loss: 0.5342 - val_accuracy: 0.8903
Epoch 4/100
accuracy: 0.8880 - val_loss: 0.5346 - val_accuracy: 0.8889
Epoch 5/100
accuracy: 0.8879 - val_loss: 0.5348 - val_accuracy: 0.8892
Epoch 6/100
accuracy: 0.8887 - val_loss: 0.5338 - val_accuracy: 0.8914
Epoch 7/100
accuracy: 0.8893 - val_loss: 0.5346 - val_accuracy: 0.8908
Epoch 8/100
accuracy: 0.8888 - val_loss: 0.5342 - val_accuracy: 0.8900
Epoch 9/100
accuracy: 0.8896 - val_loss: 0.5345 - val_accuracy: 0.8930
Epoch 10/100
accuracy: 0.8896 - val_loss: 0.5344 - val_accuracy: 0.8931
Epoch 11/100
accuracy: 0.8893 - val_loss: 0.5342 - val_accuracy: 0.8910
Epoch 12/100
accuracy: 0.8892 - val_loss: 0.5341 - val_accuracy: 0.8900
```

```
Epoch 13/100
accuracy: 0.8895 - val_loss: 0.5344 - val_accuracy: 0.8915
Epoch 14/100
accuracy: 0.8889 - val_loss: 0.5342 - val_accuracy: 0.8914
Epoch 15/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5390 -
accuracy: 0.8903 - val_loss: 0.5347 - val_accuracy: 0.8921
Epoch 16/100
accuracy: 0.8900 - val_loss: 0.5342 - val_accuracy: 0.8919
Epoch 17/100
accuracy: 0.8902 - val_loss: 0.5345 - val_accuracy: 0.8916
Epoch 18/100
accuracy: 0.8895 - val_loss: 0.5342 - val_accuracy: 0.8914
Epoch 19/100
accuracy: 0.8894 - val_loss: 0.5350 - val_accuracy: 0.8895
Epoch 20/100
accuracy: 0.8890 - val_loss: 0.5347 - val_accuracy: 0.8906
Epoch 21/100
accuracy: 0.8898 - val_loss: 0.5339 - val_accuracy: 0.8925
Epoch 22/100
accuracy: 0.8890 - val_loss: 0.5341 - val_accuracy: 0.8923
Epoch 23/100
accuracy: 0.8895 - val_loss: 0.5343 - val_accuracy: 0.8921
Epoch 24/100
accuracy: 0.8888 - val_loss: 0.5340 - val_accuracy: 0.8935
Epoch 25/100
accuracy: 0.8900 - val_loss: 0.5343 - val_accuracy: 0.8915
Epoch 26/100
accuracy: 0.8899 - val_loss: 0.5344 - val_accuracy: 0.8929
Epoch 27/100
accuracy: 0.8896 - val_loss: 0.5352 - val_accuracy: 0.8903
Epoch 28/100
accuracy: 0.8909 - val_loss: 0.5345 - val_accuracy: 0.8914
```

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Epoch 29/100
accuracy: 0.8900 - val_loss: 0.5340 - val_accuracy: 0.8935
Epoch 30/100
accuracy: 0.8907 - val_loss: 0.5342 - val_accuracy: 0.8930
Epoch 31/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5392 -
accuracy: 0.8879 - val_loss: 0.5341 - val_accuracy: 0.8933
Epoch 32/100
accuracy: 0.8887 - val_loss: 0.5345 - val_accuracy: 0.8907
Epoch 33/100
accuracy: 0.8895 - val_loss: 0.5341 - val_accuracy: 0.8911
Epoch 34/100
accuracy: 0.8901 - val_loss: 0.5354 - val_accuracy: 0.8897
Epoch 35/100
accuracy: 0.8890 - val_loss: 0.5343 - val_accuracy: 0.8901
Epoch 36/100
accuracy: 0.8895 - val_loss: 0.5340 - val_accuracy: 0.8910
Epoch 37/100
accuracy: 0.8887 - val_loss: 0.5343 - val_accuracy: 0.8901
Epoch 38/100
accuracy: 0.8872 - val_loss: 0.5343 - val_accuracy: 0.8888
Epoch 39/100
accuracy: 0.8887 - val_loss: 0.5348 - val_accuracy: 0.8919
Epoch 40/100
accuracy: 0.8895 - val_loss: 0.5345 - val_accuracy: 0.8905
Epoch 41/100
accuracy: 0.8880 - val_loss: 0.5368 - val_accuracy: 0.8878
Epoch 42/100
accuracy: 0.8885 - val_loss: 0.5343 - val_accuracy: 0.8916
Epoch 43/100
accuracy: 0.8881 - val_loss: 0.5342 - val_accuracy: 0.8912
Epoch 44/100
accuracy: 0.8889 - val_loss: 0.5337 - val_accuracy: 0.8913
```

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Epoch 45/100
accuracy: 0.8891 - val_loss: 0.5363 - val_accuracy: 0.8886
Epoch 46/100
accuracy: 0.8894 - val_loss: 0.5342 - val_accuracy: 0.8917
Epoch 47/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5383 -
accuracy: 0.8890 - val_loss: 0.5338 - val_accuracy: 0.8922
Epoch 48/100
accuracy: 0.8887 - val_loss: 0.5346 - val_accuracy: 0.8933
Epoch 49/100
accuracy: 0.8902 - val_loss: 0.5343 - val_accuracy: 0.8920
Epoch 50/100
accuracy: 0.8893 - val_loss: 0.5347 - val_accuracy: 0.8906
Epoch 51/100
accuracy: 0.8891 - val_loss: 0.5342 - val_accuracy: 0.8926
Epoch 52/100
accuracy: 0.8899 - val_loss: 0.5351 - val_accuracy: 0.8915
Epoch 53/100
accuracy: 0.8898 - val_loss: 0.5332 - val_accuracy: 0.8940
Epoch 54/100
accuracy: 0.8904 - val_loss: 0.5343 - val_accuracy: 0.8928
Epoch 55/100
accuracy: 0.8885 - val_loss: 0.5344 - val_accuracy: 0.8901
Epoch 56/100
accuracy: 0.8882 - val_loss: 0.5346 - val_accuracy: 0.8913
Epoch 57/100
accuracy: 0.8895 - val_loss: 0.5339 - val_accuracy: 0.8928
Epoch 58/100
accuracy: 0.8910 - val_loss: 0.5349 - val_accuracy: 0.8916
Epoch 59/100
accuracy: 0.8898 - val_loss: 0.5363 - val_accuracy: 0.8882
Epoch 60/100
accuracy: 0.8903 - val_loss: 0.5355 - val_accuracy: 0.8913
```

```
Epoch 61/100
accuracy: 0.8899 - val_loss: 0.5346 - val_accuracy: 0.8932
Epoch 62/100
accuracy: 0.8898 - val_loss: 0.5354 - val_accuracy: 0.8905
Epoch 63/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5383 -
accuracy: 0.8905 - val_loss: 0.5344 - val_accuracy: 0.8918
Epoch 64/100
accuracy: 0.8899 - val_loss: 0.5337 - val_accuracy: 0.8932
Epoch 65/100
accuracy: 0.8893 - val_loss: 0.5342 - val_accuracy: 0.8914
Epoch 66/100
accuracy: 0.8903 - val_loss: 0.5340 - val_accuracy: 0.8926
Epoch 67/100
1349/1349 [============= - - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8895 - val_loss: 0.5345 - val_accuracy: 0.8925
Epoch 68/100
accuracy: 0.8895 - val_loss: 0.5349 - val_accuracy: 0.8904
Epoch 69/100
accuracy: 0.8896 - val_loss: 0.5340 - val_accuracy: 0.8923
Epoch 70/100
accuracy: 0.8903 - val_loss: 0.5347 - val_accuracy: 0.8916
Epoch 71/100
accuracy: 0.8899 - val_loss: 0.5341 - val_accuracy: 0.8921
Epoch 72/100
accuracy: 0.8893 - val_loss: 0.5340 - val_accuracy: 0.8922
Epoch 73/100
accuracy: 0.8890 - val_loss: 0.5344 - val_accuracy: 0.8914
Epoch 74/100
accuracy: 0.8882 - val_loss: 0.5340 - val_accuracy: 0.8918
Epoch 75/100
accuracy: 0.8899 - val_loss: 0.5340 - val_accuracy: 0.8925
Epoch 76/100
accuracy: 0.8896 - val_loss: 0.5347 - val_accuracy: 0.8933
```

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Epoch 77/100
accuracy: 0.8885 - val_loss: 0.5349 - val_accuracy: 0.8905
Epoch 78/100
accuracy: 0.8876 - val_loss: 0.5362 - val_accuracy: 0.8879
Epoch 79/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5390 -
accuracy: 0.8879 - val_loss: 0.5340 - val_accuracy: 0.8906
Epoch 80/100
accuracy: 0.8876 - val_loss: 0.5347 - val_accuracy: 0.8904
Epoch 81/100
accuracy: 0.8893 - val_loss: 0.5333 - val_accuracy: 0.8917
Epoch 82/100
accuracy: 0.8881 - val_loss: 0.5353 - val_accuracy: 0.8884
Epoch 83/100
1349/1349 [============= - - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8879 - val_loss: 0.5339 - val_accuracy: 0.8921
Epoch 84/100
accuracy: 0.8881 - val_loss: 0.5364 - val_accuracy: 0.8892
Epoch 85/100
accuracy: 0.8890 - val_loss: 0.5348 - val_accuracy: 0.8924
Epoch 86/100
accuracy: 0.8904 - val_loss: 0.5342 - val_accuracy: 0.8939
Epoch 87/100
accuracy: 0.8894 - val_loss: 0.5342 - val_accuracy: 0.8916
Epoch 88/100
accuracy: 0.8902 - val_loss: 0.5338 - val_accuracy: 0.8929
Epoch 89/100
accuracy: 0.8893 - val_loss: 0.5338 - val_accuracy: 0.8926
Epoch 90/100
accuracy: 0.8902 - val_loss: 0.5354 - val_accuracy: 0.8904
Epoch 91/100
accuracy: 0.8899 - val_loss: 0.5372 - val_accuracy: 0.8888
Epoch 92/100
accuracy: 0.8888 - val_loss: 0.5344 - val_accuracy: 0.8926
```

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Epoch 93/100
accuracy: 0.8894 - val_loss: 0.5343 - val_accuracy: 0.8927
Epoch 94/100
accuracy: 0.8903 - val_loss: 0.5345 - val_accuracy: 0.8922
Epoch 95/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5378 -
accuracy: 0.8895 - val_loss: 0.5342 - val_accuracy: 0.8917
Epoch 96/100
accuracy: 0.8894 - val_loss: 0.5371 - val_accuracy: 0.8883
Epoch 97/100
accuracy: 0.8882 - val_loss: 0.5344 - val_accuracy: 0.8917
Epoch 98/100
accuracy: 0.8883 - val_loss: 0.5338 - val_accuracy: 0.8926
Epoch 99/100
1349/1349 [============= - - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8874 - val_loss: 0.5343 - val_accuracy: 0.8886
Epoch 100/100
accuracy: 0.8885 - val_loss: 0.5352 - val_accuracy: 0.8904
Epoch 1/100
accuracy: 0.8874 - val_loss: 0.5338 - val_accuracy: 0.8915
Epoch 2/100
accuracy: 0.8885 - val_loss: 0.5344 - val_accuracy: 0.8921
Epoch 3/100
accuracy: 0.8892 - val_loss: 0.5337 - val_accuracy: 0.8923
Epoch 4/100
accuracy: 0.8883 - val_loss: 0.5343 - val_accuracy: 0.8929
Epoch 5/100
accuracy: 0.8878 - val_loss: 0.5340 - val_accuracy: 0.8912
Epoch 6/100
accuracy: 0.8891 - val_loss: 0.5345 - val_accuracy: 0.8914
Epoch 7/100
accuracy: 0.8880 - val_loss: 0.5348 - val_accuracy: 0.8892
Epoch 8/100
accuracy: 0.8882 - val_loss: 0.5338 - val_accuracy: 0.8913
```

```
Epoch 9/100
accuracy: 0.8896 - val_loss: 0.5355 - val_accuracy: 0.8886
Epoch 10/100
accuracy: 0.8876 - val_loss: 0.5343 - val_accuracy: 0.8886
Epoch 11/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5396 -
accuracy: 0.8874 - val_loss: 0.5348 - val_accuracy: 0.8911
Epoch 12/100
accuracy: 0.8885 - val_loss: 0.5338 - val_accuracy: 0.8900
Epoch 13/100
accuracy: 0.8874 - val_loss: 0.5346 - val_accuracy: 0.8900
Epoch 14/100
accuracy: 0.8871 - val_loss: 0.5351 - val_accuracy: 0.8885
Epoch 15/100
accuracy: 0.8881 - val_loss: 0.5339 - val_accuracy: 0.8902
Epoch 16/100
accuracy: 0.8898 - val_loss: 0.5344 - val_accuracy: 0.8919
Epoch 17/100
accuracy: 0.8879 - val_loss: 0.5367 - val_accuracy: 0.8875
Epoch 18/100
accuracy: 0.8893 - val_loss: 0.5353 - val_accuracy: 0.8888
Epoch 19/100
accuracy: 0.8886 - val_loss: 0.5334 - val_accuracy: 0.8918
Epoch 20/100
accuracy: 0.8891 - val_loss: 0.5342 - val_accuracy: 0.8913
Epoch 21/100
accuracy: 0.8884 - val_loss: 0.5342 - val_accuracy: 0.8909
Epoch 22/100
accuracy: 0.8879 - val_loss: 0.5351 - val_accuracy: 0.8912
Epoch 23/100
accuracy: 0.8888 - val_loss: 0.5352 - val_accuracy: 0.8871
Epoch 24/100
accuracy: 0.8864 - val_loss: 0.5344 - val_accuracy: 0.8888
```

```
Epoch 25/100
accuracy: 0.8872 - val_loss: 0.5340 - val_accuracy: 0.8912
Epoch 26/100
accuracy: 0.8883 - val_loss: 0.5352 - val_accuracy: 0.8882
Epoch 27/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5384 -
accuracy: 0.8882 - val_loss: 0.5340 - val_accuracy: 0.8916
Epoch 28/100
accuracy: 0.8900 - val_loss: 0.5339 - val_accuracy: 0.8896
Epoch 29/100
accuracy: 0.8880 - val_loss: 0.5343 - val_accuracy: 0.8906
Epoch 30/100
accuracy: 0.8881 - val_loss: 0.5343 - val_accuracy: 0.8884
Epoch 31/100
1349/1349 [============= - - 2s 1ms/step - loss: 0.5387 -
accuracy: 0.8882 - val_loss: 0.5346 - val_accuracy: 0.8909
Epoch 32/100
accuracy: 0.8894 - val_loss: 0.5338 - val_accuracy: 0.8912
Epoch 33/100
accuracy: 0.8889 - val_loss: 0.5344 - val_accuracy: 0.8907
Epoch 34/100
accuracy: 0.8885 - val_loss: 0.5338 - val_accuracy: 0.8908
Epoch 35/100
accuracy: 0.8882 - val_loss: 0.5339 - val_accuracy: 0.8912
Epoch 36/100
accuracy: 0.8894 - val_loss: 0.5347 - val_accuracy: 0.8911
Epoch 37/100
accuracy: 0.8890 - val_loss: 0.5355 - val_accuracy: 0.8911
Epoch 38/100
accuracy: 0.8895 - val_loss: 0.5340 - val_accuracy: 0.8910
Epoch 39/100
accuracy: 0.8886 - val_loss: 0.5351 - val_accuracy: 0.8889
Epoch 40/100
accuracy: 0.8879 - val_loss: 0.5349 - val_accuracy: 0.8887
```

```
Epoch 41/100
accuracy: 0.8891 - val_loss: 0.5350 - val_accuracy: 0.8902
Epoch 42/100
accuracy: 0.8894 - val_loss: 0.5340 - val_accuracy: 0.8913
Epoch 43/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5384 -
accuracy: 0.8884 - val_loss: 0.5336 - val_accuracy: 0.8914
Epoch 44/100
accuracy: 0.8884 - val_loss: 0.5337 - val_accuracy: 0.8917
Epoch 45/100
accuracy: 0.8899 - val_loss: 0.5338 - val_accuracy: 0.8915
Epoch 46/100
accuracy: 0.8888 - val_loss: 0.5342 - val_accuracy: 0.8919
Epoch 47/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5383 -
accuracy: 0.8892 - val_loss: 0.5348 - val_accuracy: 0.8905
Epoch 48/100
accuracy: 0.8888 - val_loss: 0.5338 - val_accuracy: 0.8910
Epoch 49/100
accuracy: 0.8892 - val_loss: 0.5335 - val_accuracy: 0.8919
Epoch 50/100
accuracy: 0.8895 - val_loss: 0.5339 - val_accuracy: 0.8921
Epoch 51/100
accuracy: 0.8893 - val_loss: 0.5343 - val_accuracy: 0.8921
Epoch 52/100
accuracy: 0.8896 - val_loss: 0.5344 - val_accuracy: 0.8903
Epoch 53/100
accuracy: 0.8890 - val_loss: 0.5345 - val_accuracy: 0.8895
Epoch 54/100
accuracy: 0.8876 - val_loss: 0.5345 - val_accuracy: 0.8885
Epoch 55/100
accuracy: 0.8887 - val_loss: 0.5341 - val_accuracy: 0.8914
Epoch 56/100
accuracy: 0.8881 - val_loss: 0.5335 - val_accuracy: 0.8903
```

```
Epoch 57/100
accuracy: 0.8870 - val_loss: 0.5334 - val_accuracy: 0.8911
Epoch 58/100
accuracy: 0.8882 - val_loss: 0.5343 - val_accuracy: 0.8906
Epoch 59/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5388 -
accuracy: 0.8887 - val_loss: 0.5336 - val_accuracy: 0.8919
Epoch 60/100
accuracy: 0.8883 - val_loss: 0.5360 - val_accuracy: 0.8876
Epoch 61/100
accuracy: 0.8894 - val_loss: 0.5335 - val_accuracy: 0.8905
Epoch 62/100
accuracy: 0.8887 - val_loss: 0.5338 - val_accuracy: 0.8915
Epoch 63/100
accuracy: 0.8883 - val_loss: 0.5352 - val_accuracy: 0.8891
Epoch 64/100
accuracy: 0.8873 - val_loss: 0.5334 - val_accuracy: 0.8916
Epoch 65/100
accuracy: 0.8878 - val_loss: 0.5337 - val_accuracy: 0.8924
Epoch 66/100
accuracy: 0.8891 - val_loss: 0.5335 - val_accuracy: 0.8915
Epoch 67/100
accuracy: 0.8889 - val_loss: 0.5336 - val_accuracy: 0.8906
Epoch 68/100
accuracy: 0.8893 - val_loss: 0.5352 - val_accuracy: 0.8906
Epoch 69/100
accuracy: 0.8892 - val_loss: 0.5341 - val_accuracy: 0.8922
Epoch 70/100
accuracy: 0.8884 - val_loss: 0.5340 - val_accuracy: 0.8923
Epoch 71/100
accuracy: 0.8890 - val_loss: 0.5339 - val_accuracy: 0.8912
Epoch 72/100
accuracy: 0.8903 - val_loss: 0.5341 - val_accuracy: 0.8901
```

```
Epoch 73/100
accuracy: 0.8892 - val_loss: 0.5340 - val_accuracy: 0.8915
Epoch 74/100
accuracy: 0.8893 - val_loss: 0.5350 - val_accuracy: 0.8892
Epoch 75/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8880 - val_loss: 0.5338 - val_accuracy: 0.8923
Epoch 76/100
accuracy: 0.8897 - val_loss: 0.5336 - val_accuracy: 0.8921
Epoch 77/100
accuracy: 0.8885 - val_loss: 0.5336 - val_accuracy: 0.8917
Epoch 78/100
accuracy: 0.8900 - val_loss: 0.5341 - val_accuracy: 0.8915
Epoch 79/100
accuracy: 0.8884 - val_loss: 0.5338 - val_accuracy: 0.8909
Epoch 80/100
accuracy: 0.8867 - val_loss: 0.5335 - val_accuracy: 0.8913
Epoch 81/100
accuracy: 0.8897 - val_loss: 0.5348 - val_accuracy: 0.8876
Epoch 82/100
accuracy: 0.8877 - val_loss: 0.5333 - val_accuracy: 0.8911
Epoch 83/100
accuracy: 0.8874 - val_loss: 0.5340 - val_accuracy: 0.8892
Epoch 84/100
accuracy: 0.8876 - val_loss: 0.5335 - val_accuracy: 0.8909
Epoch 85/100
accuracy: 0.8886 - val_loss: 0.5336 - val_accuracy: 0.8906
Epoch 86/100
accuracy: 0.8877 - val_loss: 0.5341 - val_accuracy: 0.8915
Epoch 87/100
accuracy: 0.8891 - val_loss: 0.5337 - val_accuracy: 0.8906
Epoch 88/100
accuracy: 0.8894 - val_loss: 0.5358 - val_accuracy: 0.8911
```

```
Epoch 89/100
accuracy: 0.8888 - val_loss: 0.5339 - val_accuracy: 0.8905
Epoch 90/100
accuracy: 0.8877 - val_loss: 0.5345 - val_accuracy: 0.8911
Epoch 91/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5376 -
accuracy: 0.8892 - val_loss: 0.5336 - val_accuracy: 0.8913
Epoch 92/100
accuracy: 0.8892 - val_loss: 0.5338 - val_accuracy: 0.8907
Epoch 93/100
accuracy: 0.8895 - val_loss: 0.5342 - val_accuracy: 0.8917
Epoch 94/100
accuracy: 0.8892 - val_loss: 0.5335 - val_accuracy: 0.8924
Epoch 95/100
accuracy: 0.8889 - val_loss: 0.5337 - val_accuracy: 0.8910
Epoch 96/100
accuracy: 0.8882 - val_loss: 0.5348 - val_accuracy: 0.8900
Epoch 97/100
accuracy: 0.8888 - val_loss: 0.5342 - val_accuracy: 0.8909
Epoch 98/100
accuracy: 0.8899 - val_loss: 0.5338 - val_accuracy: 0.8921
Epoch 99/100
accuracy: 0.8889 - val_loss: 0.5341 - val_accuracy: 0.8914
Epoch 100/100
accuracy: 0.8899 - val_loss: 0.5338 - val_accuracy: 0.8916
Epoch 1/100
accuracy: 0.8898 - val_loss: 0.5339 - val_accuracy: 0.8916
Epoch 2/100
accuracy: 0.8891 - val_loss: 0.5339 - val_accuracy: 0.8921
Epoch 3/100
accuracy: 0.8887 - val_loss: 0.5339 - val_accuracy: 0.8902
Epoch 4/100
accuracy: 0.8896 - val_loss: 0.5342 - val_accuracy: 0.8919
```

```
Epoch 5/100
accuracy: 0.8892 - val_loss: 0.5339 - val_accuracy: 0.8896
Epoch 6/100
accuracy: 0.8884 - val_loss: 0.5351 - val_accuracy: 0.8894
Epoch 7/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5380 -
accuracy: 0.8894 - val_loss: 0.5340 - val_accuracy: 0.8922
Epoch 8/100
accuracy: 0.8901 - val_loss: 0.5356 - val_accuracy: 0.8884
Epoch 9/100
accuracy: 0.8888 - val_loss: 0.5339 - val_accuracy: 0.8912
Epoch 10/100
accuracy: 0.8899 - val_loss: 0.5334 - val_accuracy: 0.8918
Epoch 11/100
accuracy: 0.8901 - val_loss: 0.5339 - val_accuracy: 0.8910
Epoch 12/100
accuracy: 0.8894 - val_loss: 0.5333 - val_accuracy: 0.8923
Epoch 13/100
accuracy: 0.8898 - val_loss: 0.5339 - val_accuracy: 0.8922
Epoch 14/100
accuracy: 0.8900 - val_loss: 0.5338 - val_accuracy: 0.8923
Epoch 15/100
accuracy: 0.8894 - val_loss: 0.5354 - val_accuracy: 0.8921
Epoch 16/100
accuracy: 0.8890 - val_loss: 0.5339 - val_accuracy: 0.8907
Epoch 17/100
accuracy: 0.8897 - val_loss: 0.5345 - val_accuracy: 0.8919
Epoch 18/100
accuracy: 0.8895 - val_loss: 0.5332 - val_accuracy: 0.8917
Epoch 19/100
accuracy: 0.8890 - val_loss: 0.5340 - val_accuracy: 0.8923
Epoch 20/100
accuracy: 0.8895 - val_loss: 0.5339 - val_accuracy: 0.8900
```

```
Epoch 21/100
accuracy: 0.8892 - val_loss: 0.5341 - val_accuracy: 0.8893
Epoch 22/100
accuracy: 0.8895 - val_loss: 0.5345 - val_accuracy: 0.8891
Epoch 23/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5371 -
accuracy: 0.8908 - val_loss: 0.5339 - val_accuracy: 0.8914
Epoch 24/100
accuracy: 0.8885 - val_loss: 0.5338 - val_accuracy: 0.8903
Epoch 25/100
accuracy: 0.8879 - val_loss: 0.5344 - val_accuracy: 0.8913
Epoch 26/100
accuracy: 0.8888 - val_loss: 0.5338 - val_accuracy: 0.8913
Epoch 27/100
accuracy: 0.8885 - val_loss: 0.5340 - val_accuracy: 0.8916
Epoch 28/100
accuracy: 0.8888 - val_loss: 0.5340 - val_accuracy: 0.8902
Epoch 29/100
accuracy: 0.8886 - val_loss: 0.5361 - val_accuracy: 0.8887
Epoch 30/100
accuracy: 0.8897 - val_loss: 0.5337 - val_accuracy: 0.8911
Epoch 31/100
accuracy: 0.8903 - val_loss: 0.5341 - val_accuracy: 0.8929
Epoch 32/100
accuracy: 0.8894 - val_loss: 0.5343 - val_accuracy: 0.8925
Epoch 33/100
accuracy: 0.8893 - val_loss: 0.5345 - val_accuracy: 0.8917
Epoch 34/100
accuracy: 0.8904 - val_loss: 0.5336 - val_accuracy: 0.8915
Epoch 35/100
accuracy: 0.8896 - val_loss: 0.5338 - val_accuracy: 0.8917
Epoch 36/100
accuracy: 0.8897 - val_loss: 0.5336 - val_accuracy: 0.8930
```

```
Epoch 37/100
accuracy: 0.8908 - val_loss: 0.5340 - val_accuracy: 0.8924
Epoch 38/100
accuracy: 0.8896 - val_loss: 0.5339 - val_accuracy: 0.8914
Epoch 39/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5380 -
accuracy: 0.8896 - val_loss: 0.5336 - val_accuracy: 0.8909
Epoch 40/100
accuracy: 0.8892 - val_loss: 0.5337 - val_accuracy: 0.8905
Epoch 41/100
accuracy: 0.8888 - val_loss: 0.5332 - val_accuracy: 0.8924
Epoch 42/100
accuracy: 0.8895 - val_loss: 0.5338 - val_accuracy: 0.8921
Epoch 43/100
accuracy: 0.8910 - val_loss: 0.5343 - val_accuracy: 0.8918
Epoch 44/100
accuracy: 0.8898 - val_loss: 0.5340 - val_accuracy: 0.8921
Epoch 45/100
accuracy: 0.8883 - val_loss: 0.5341 - val_accuracy: 0.8921
Epoch 46/100
accuracy: 0.8901 - val_loss: 0.5337 - val_accuracy: 0.8915
Epoch 47/100
accuracy: 0.8891 - val_loss: 0.5332 - val_accuracy: 0.8928
Epoch 48/100
accuracy: 0.8899 - val_loss: 0.5336 - val_accuracy: 0.8921
Epoch 49/100
accuracy: 0.8888 - val_loss: 0.5339 - val_accuracy: 0.8913
Epoch 50/100
accuracy: 0.8888 - val_loss: 0.5349 - val_accuracy: 0.8879
Epoch 51/100
accuracy: 0.8891 - val_loss: 0.5336 - val_accuracy: 0.8909
Epoch 52/100
accuracy: 0.8889 - val_loss: 0.5342 - val_accuracy: 0.8912
```

```
Epoch 53/100
accuracy: 0.8890 - val_loss: 0.5360 - val_accuracy: 0.8885
Epoch 54/100
accuracy: 0.8893 - val_loss: 0.5336 - val_accuracy: 0.8918
Epoch 55/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5375 -
accuracy: 0.8899 - val_loss: 0.5339 - val_accuracy: 0.8917
Epoch 56/100
accuracy: 0.8901 - val_loss: 0.5333 - val_accuracy: 0.8913
Epoch 57/100
accuracy: 0.8896 - val_loss: 0.5335 - val_accuracy: 0.8924
Epoch 58/100
accuracy: 0.8894 - val_loss: 0.5342 - val_accuracy: 0.8913
Epoch 59/100
accuracy: 0.8884 - val_loss: 0.5349 - val_accuracy: 0.8910
Epoch 60/100
accuracy: 0.8886 - val_loss: 0.5343 - val_accuracy: 0.8905
Epoch 61/100
accuracy: 0.8876 - val_loss: 0.5339 - val_accuracy: 0.8902
Epoch 62/100
accuracy: 0.8896 - val_loss: 0.5333 - val_accuracy: 0.8916
Epoch 63/100
accuracy: 0.8898 - val_loss: 0.5337 - val_accuracy: 0.8912
Epoch 64/100
accuracy: 0.8896 - val_loss: 0.5337 - val_accuracy: 0.8897
Epoch 65/100
accuracy: 0.8888 - val_loss: 0.5336 - val_accuracy: 0.8906
Epoch 66/100
accuracy: 0.8899 - val_loss: 0.5337 - val_accuracy: 0.8921
Epoch 67/100
accuracy: 0.8893 - val_loss: 0.5336 - val_accuracy: 0.8917
Epoch 68/100
accuracy: 0.8892 - val_loss: 0.5339 - val_accuracy: 0.8912
```

```
Epoch 69/100
accuracy: 0.8903 - val_loss: 0.5335 - val_accuracy: 0.8922
Epoch 70/100
accuracy: 0.8904 - val_loss: 0.5341 - val_accuracy: 0.8919
Epoch 71/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8886 - val_loss: 0.5342 - val_accuracy: 0.8911
Epoch 72/100
accuracy: 0.8883 - val_loss: 0.5334 - val_accuracy: 0.8919
Epoch 73/100
accuracy: 0.8891 - val_loss: 0.5345 - val_accuracy: 0.8895
Epoch 74/100
accuracy: 0.8893 - val_loss: 0.5349 - val_accuracy: 0.8897
Epoch 75/100
accuracy: 0.8893 - val_loss: 0.5345 - val_accuracy: 0.8920
Epoch 76/100
accuracy: 0.8894 - val_loss: 0.5344 - val_accuracy: 0.8902
Epoch 77/100
accuracy: 0.8883 - val_loss: 0.5344 - val_accuracy: 0.8905
Epoch 78/100
accuracy: 0.8885 - val_loss: 0.5335 - val_accuracy: 0.8911
Epoch 79/100
accuracy: 0.8890 - val_loss: 0.5331 - val_accuracy: 0.8927
Epoch 80/100
accuracy: 0.8886 - val_loss: 0.5340 - val_accuracy: 0.8921
Epoch 81/100
accuracy: 0.8899 - val_loss: 0.5342 - val_accuracy: 0.8923
Epoch 82/100
accuracy: 0.8897 - val_loss: 0.5340 - val_accuracy: 0.8927
Epoch 83/100
accuracy: 0.8896 - val_loss: 0.5343 - val_accuracy: 0.8905
Epoch 84/100
accuracy: 0.8896 - val_loss: 0.5338 - val_accuracy: 0.8921
```

```
Epoch 85/100
accuracy: 0.8888 - val_loss: 0.5356 - val_accuracy: 0.8882
Epoch 86/100
accuracy: 0.8902 - val_loss: 0.5336 - val_accuracy: 0.8917
Epoch 87/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5389 -
accuracy: 0.8888 - val_loss: 0.5340 - val_accuracy: 0.8919
Epoch 88/100
accuracy: 0.8890 - val_loss: 0.5336 - val_accuracy: 0.8924
Epoch 89/100
accuracy: 0.8899 - val_loss: 0.5336 - val_accuracy: 0.8922
Epoch 90/100
accuracy: 0.8898 - val_loss: 0.5358 - val_accuracy: 0.8900
Epoch 91/100
accuracy: 0.8882 - val_loss: 0.5335 - val_accuracy: 0.8899
Epoch 92/100
accuracy: 0.8879 - val_loss: 0.5333 - val_accuracy: 0.8923
Epoch 93/100
accuracy: 0.8890 - val_loss: 0.5335 - val_accuracy: 0.8920
Epoch 94/100
accuracy: 0.8895 - val_loss: 0.5342 - val_accuracy: 0.8931
Epoch 95/100
accuracy: 0.8903 - val_loss: 0.5338 - val_accuracy: 0.8914
Epoch 96/100
accuracy: 0.8894 - val_loss: 0.5338 - val_accuracy: 0.8916
Epoch 97/100
accuracy: 0.8889 - val_loss: 0.5334 - val_accuracy: 0.8936
Epoch 98/100
accuracy: 0.8899 - val_loss: 0.5337 - val_accuracy: 0.8920
Epoch 99/100
accuracy: 0.8899 - val_loss: 0.5345 - val_accuracy: 0.8925
Epoch 100/100
accuracy: 0.8893 - val_loss: 0.5340 - val_accuracy: 0.8914
```

```
Epoch 1/100
accuracy: 0.8885 - val_loss: 0.5335 - val_accuracy: 0.8905
Epoch 2/100
accuracy: 0.8882 - val_loss: 0.5344 - val_accuracy: 0.8922
Epoch 3/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5378 -
accuracy: 0.8902 - val_loss: 0.5349 - val_accuracy: 0.8917
Epoch 4/100
accuracy: 0.8886 - val_loss: 0.5339 - val_accuracy: 0.8915
Epoch 5/100
accuracy: 0.8887 - val_loss: 0.5338 - val_accuracy: 0.8923
Epoch 6/100
accuracy: 0.8899 - val_loss: 0.5334 - val_accuracy: 0.8933
Epoch 7/100
accuracy: 0.8897 - val_loss: 0.5340 - val_accuracy: 0.8925
Epoch 8/100
accuracy: 0.8899 - val_loss: 0.5340 - val_accuracy: 0.8912
Epoch 9/100
accuracy: 0.8895 - val_loss: 0.5349 - val_accuracy: 0.8877
Epoch 10/100
accuracy: 0.8903 - val_loss: 0.5365 - val_accuracy: 0.8926
Epoch 11/100
accuracy: 0.8907 - val_loss: 0.5341 - val_accuracy: 0.8930
Epoch 12/100
accuracy: 0.8902 - val_loss: 0.5335 - val_accuracy: 0.8918
Epoch 13/100
accuracy: 0.8895 - val_loss: 0.5333 - val_accuracy: 0.8926
Epoch 14/100
accuracy: 0.8894 - val_loss: 0.5339 - val_accuracy: 0.8912
Epoch 15/100
accuracy: 0.8896 - val_loss: 0.5341 - val_accuracy: 0.8931
Epoch 16/100
accuracy: 0.8908 - val_loss: 0.5335 - val_accuracy: 0.8924
```

```
Epoch 17/100
accuracy: 0.8889 - val_loss: 0.5335 - val_accuracy: 0.8915
Epoch 18/100
accuracy: 0.8897 - val_loss: 0.5339 - val_accuracy: 0.8915
Epoch 19/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5390 -
accuracy: 0.8888 - val_loss: 0.5344 - val_accuracy: 0.8925
Epoch 20/100
accuracy: 0.8892 - val_loss: 0.5346 - val_accuracy: 0.8901
Epoch 21/100
accuracy: 0.8895 - val_loss: 0.5335 - val_accuracy: 0.8915
Epoch 22/100
accuracy: 0.8896 - val_loss: 0.5340 - val_accuracy: 0.8924
Epoch 23/100
accuracy: 0.8903 - val_loss: 0.5347 - val_accuracy: 0.8903
Epoch 24/100
accuracy: 0.8891 - val_loss: 0.5340 - val_accuracy: 0.8921
Epoch 25/100
accuracy: 0.8902 - val_loss: 0.5345 - val_accuracy: 0.8912
Epoch 26/100
accuracy: 0.8900 - val_loss: 0.5346 - val_accuracy: 0.8922
Epoch 27/100
accuracy: 0.8905 - val_loss: 0.5333 - val_accuracy: 0.8925
Epoch 28/100
accuracy: 0.8899 - val_loss: 0.5335 - val_accuracy: 0.8921
Epoch 29/100
accuracy: 0.8906 - val_loss: 0.5339 - val_accuracy: 0.8918
Epoch 30/100
accuracy: 0.8899 - val_loss: 0.5340 - val_accuracy: 0.8911
Epoch 31/100
accuracy: 0.8895 - val_loss: 0.5336 - val_accuracy: 0.8927
Epoch 32/100
accuracy: 0.8906 - val_loss: 0.5339 - val_accuracy: 0.8915
```

```
Epoch 33/100
accuracy: 0.8906 - val_loss: 0.5341 - val_accuracy: 0.8904
Epoch 34/100
accuracy: 0.8893 - val_loss: 0.5338 - val_accuracy: 0.8925
Epoch 35/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5375 -
accuracy: 0.8903 - val_loss: 0.5336 - val_accuracy: 0.8919
Epoch 36/100
accuracy: 0.8890 - val_loss: 0.5337 - val_accuracy: 0.8909
Epoch 37/100
accuracy: 0.8896 - val_loss: 0.5336 - val_accuracy: 0.8923
Epoch 38/100
accuracy: 0.8893 - val_loss: 0.5333 - val_accuracy: 0.8929
Epoch 39/100
accuracy: 0.8906 - val_loss: 0.5342 - val_accuracy: 0.8923
Epoch 40/100
accuracy: 0.8914 - val_loss: 0.5335 - val_accuracy: 0.8919
Epoch 41/100
accuracy: 0.8899 - val_loss: 0.5337 - val_accuracy: 0.8928
Epoch 42/100
accuracy: 0.8903 - val_loss: 0.5337 - val_accuracy: 0.8907
Epoch 43/100
accuracy: 0.8895 - val_loss: 0.5333 - val_accuracy: 0.8929
Epoch 44/100
accuracy: 0.8898 - val_loss: 0.5357 - val_accuracy: 0.8927
Epoch 45/100
accuracy: 0.8909 - val_loss: 0.5339 - val_accuracy: 0.8922
Epoch 46/100
accuracy: 0.8902 - val_loss: 0.5343 - val_accuracy: 0.8922
Epoch 47/100
accuracy: 0.8902 - val_loss: 0.5345 - val_accuracy: 0.8928
Epoch 48/100
accuracy: 0.8903 - val_loss: 0.5335 - val_accuracy: 0.8914
```

```
Epoch 49/100
accuracy: 0.8905 - val_loss: 0.5343 - val_accuracy: 0.8933
Epoch 50/100
accuracy: 0.8892 - val_loss: 0.5340 - val_accuracy: 0.8909
Epoch 51/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5373 -
accuracy: 0.8896 - val_loss: 0.5345 - val_accuracy: 0.8914
Epoch 52/100
accuracy: 0.8890 - val_loss: 0.5335 - val_accuracy: 0.8916
Epoch 53/100
accuracy: 0.8908 - val_loss: 0.5333 - val_accuracy: 0.8922
Epoch 54/100
accuracy: 0.8892 - val_loss: 0.5344 - val_accuracy: 0.8908
Epoch 55/100
accuracy: 0.8891 - val_loss: 0.5350 - val_accuracy: 0.8886
Epoch 56/100
accuracy: 0.8895 - val_loss: 0.5340 - val_accuracy: 0.8904
Epoch 57/100
accuracy: 0.8885 - val_loss: 0.5341 - val_accuracy: 0.8890
Epoch 58/100
accuracy: 0.8884 - val_loss: 0.5336 - val_accuracy: 0.8905
Epoch 59/100
accuracy: 0.8893 - val_loss: 0.5346 - val_accuracy: 0.8899
Epoch 60/100
accuracy: 0.8895 - val_loss: 0.5334 - val_accuracy: 0.8911
Epoch 61/100
accuracy: 0.8904 - val_loss: 0.5336 - val_accuracy: 0.8922
Epoch 62/100
accuracy: 0.8899 - val_loss: 0.5344 - val_accuracy: 0.8903
Epoch 63/100
accuracy: 0.8888 - val_loss: 0.5333 - val_accuracy: 0.8919
Epoch 64/100
accuracy: 0.8902 - val_loss: 0.5335 - val_accuracy: 0.8926
```

```
Epoch 65/100
accuracy: 0.8892 - val_loss: 0.5340 - val_accuracy: 0.8919
Epoch 66/100
accuracy: 0.8899 - val_loss: 0.5335 - val_accuracy: 0.8929
Epoch 67/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8899 - val_loss: 0.5337 - val_accuracy: 0.8920
Epoch 68/100
accuracy: 0.8898 - val_loss: 0.5334 - val_accuracy: 0.8923
Epoch 69/100
accuracy: 0.8898 - val_loss: 0.5335 - val_accuracy: 0.8928
Epoch 70/100
accuracy: 0.8898 - val_loss: 0.5333 - val_accuracy: 0.8920
Epoch 71/100
accuracy: 0.8903 - val_loss: 0.5337 - val_accuracy: 0.8929
Epoch 72/100
accuracy: 0.8890 - val_loss: 0.5335 - val_accuracy: 0.8919
Epoch 73/100
accuracy: 0.8890 - val_loss: 0.5334 - val_accuracy: 0.8921
Epoch 74/100
accuracy: 0.8899 - val_loss: 0.5341 - val_accuracy: 0.8919
Epoch 75/100
accuracy: 0.8903 - val_loss: 0.5352 - val_accuracy: 0.8890
Epoch 76/100
accuracy: 0.8900 - val_loss: 0.5335 - val_accuracy: 0.8927
Epoch 77/100
accuracy: 0.8903 - val_loss: 0.5335 - val_accuracy: 0.8923
Epoch 78/100
accuracy: 0.8889 - val_loss: 0.5338 - val_accuracy: 0.8920
Epoch 79/100
accuracy: 0.8909 - val_loss: 0.5333 - val_accuracy: 0.8915
Epoch 80/100
accuracy: 0.8894 - val_loss: 0.5337 - val_accuracy: 0.8918
```

```
Epoch 81/100
accuracy: 0.8899 - val_loss: 0.5339 - val_accuracy: 0.8923
Epoch 82/100
accuracy: 0.8897 - val_loss: 0.5337 - val_accuracy: 0.8924
Epoch 83/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5381 -
accuracy: 0.8903 - val_loss: 0.5335 - val_accuracy: 0.8925
Epoch 84/100
accuracy: 0.8906 - val_loss: 0.5331 - val_accuracy: 0.8916
Epoch 85/100
accuracy: 0.8894 - val_loss: 0.5338 - val_accuracy: 0.8913
Epoch 86/100
accuracy: 0.8897 - val_loss: 0.5338 - val_accuracy: 0.8931
Epoch 87/100
accuracy: 0.8903 - val_loss: 0.5333 - val_accuracy: 0.8918
Epoch 88/100
accuracy: 0.8899 - val_loss: 0.5335 - val_accuracy: 0.8929
Epoch 89/100
accuracy: 0.8884 - val_loss: 0.5346 - val_accuracy: 0.8909
Epoch 90/100
accuracy: 0.8883 - val_loss: 0.5333 - val_accuracy: 0.8901
Epoch 91/100
accuracy: 0.8889 - val_loss: 0.5340 - val_accuracy: 0.8918
Epoch 92/100
accuracy: 0.8896 - val_loss: 0.5337 - val_accuracy: 0.8928
Epoch 93/100
accuracy: 0.8905 - val_loss: 0.5343 - val_accuracy: 0.8898
Epoch 94/100
accuracy: 0.8893 - val_loss: 0.5342 - val_accuracy: 0.8914
Epoch 95/100
accuracy: 0.8903 - val_loss: 0.5332 - val_accuracy: 0.8932
Epoch 96/100
accuracy: 0.8912 - val_loss: 0.5331 - val_accuracy: 0.8924
```

```
Epoch 97/100
accuracy: 0.8910 - val_loss: 0.5333 - val_accuracy: 0.8928
Epoch 98/100
accuracy: 0.8892 - val_loss: 0.5342 - val_accuracy: 0.8911
Epoch 99/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5373 -
accuracy: 0.8903 - val_loss: 0.5343 - val_accuracy: 0.8911
Epoch 100/100
accuracy: 0.8890 - val_loss: 0.5338 - val_accuracy: 0.8913
Epoch 1/100
accuracy: 0.8902 - val_loss: 0.5344 - val_accuracy: 0.8907
Epoch 2/100
accuracy: 0.8899 - val_loss: 0.5338 - val_accuracy: 0.8915
Epoch 3/100
accuracy: 0.8894 - val_loss: 0.5336 - val_accuracy: 0.8911
Epoch 4/100
accuracy: 0.8887 - val_loss: 0.5342 - val_accuracy: 0.8912
Epoch 5/100
accuracy: 0.8889 - val_loss: 0.5337 - val_accuracy: 0.8916
Epoch 6/100
accuracy: 0.8895 - val_loss: 0.5335 - val_accuracy: 0.8911
Epoch 7/100
accuracy: 0.8905 - val_loss: 0.5336 - val_accuracy: 0.8922
Epoch 8/100
accuracy: 0.8899 - val_loss: 0.5341 - val_accuracy: 0.8895
Epoch 9/100
accuracy: 0.8876 - val_loss: 0.5338 - val_accuracy: 0.8900
Epoch 10/100
accuracy: 0.8896 - val_loss: 0.5339 - val_accuracy: 0.8922
Epoch 11/100
accuracy: 0.8909 - val_loss: 0.5333 - val_accuracy: 0.8929
Epoch 12/100
accuracy: 0.8905 - val_loss: 0.5333 - val_accuracy: 0.8924
```

```
Epoch 13/100
accuracy: 0.8887 - val_loss: 0.5346 - val_accuracy: 0.8889
Epoch 14/100
accuracy: 0.8879 - val_loss: 0.5364 - val_accuracy: 0.8876
Epoch 15/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5370 -
accuracy: 0.8903 - val_loss: 0.5340 - val_accuracy: 0.8921
Epoch 16/100
accuracy: 0.8896 - val_loss: 0.5343 - val_accuracy: 0.8913
Epoch 17/100
accuracy: 0.8907 - val_loss: 0.5346 - val_accuracy: 0.8913
Epoch 18/100
accuracy: 0.8905 - val_loss: 0.5342 - val_accuracy: 0.8918
Epoch 19/100
accuracy: 0.8906 - val_loss: 0.5338 - val_accuracy: 0.8927
Epoch 20/100
accuracy: 0.8905 - val_loss: 0.5338 - val_accuracy: 0.8921
Epoch 21/100
accuracy: 0.8891 - val_loss: 0.5355 - val_accuracy: 0.8870
Epoch 22/100
accuracy: 0.8892 - val_loss: 0.5336 - val_accuracy: 0.8920
Epoch 23/100
accuracy: 0.8905 - val_loss: 0.5337 - val_accuracy: 0.8922
Epoch 24/100
accuracy: 0.8898 - val_loss: 0.5335 - val_accuracy: 0.8916
Epoch 25/100
accuracy: 0.8894 - val_loss: 0.5337 - val_accuracy: 0.8924
Epoch 26/100
accuracy: 0.8886 - val_loss: 0.5337 - val_accuracy: 0.8899
Epoch 27/100
accuracy: 0.8893 - val_loss: 0.5337 - val_accuracy: 0.8911
Epoch 28/100
accuracy: 0.8892 - val_loss: 0.5353 - val_accuracy: 0.8881
```

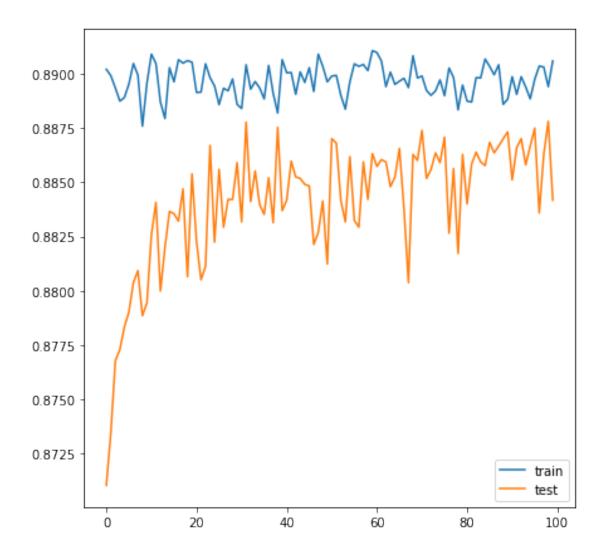
```
Epoch 29/100
accuracy: 0.8898 - val_loss: 0.5337 - val_accuracy: 0.8904
Epoch 30/100
accuracy: 0.8886 - val_loss: 0.5347 - val_accuracy: 0.8885
Epoch 31/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5375 -
accuracy: 0.8884 - val_loss: 0.5337 - val_accuracy: 0.8918
Epoch 32/100
accuracy: 0.8904 - val_loss: 0.5339 - val_accuracy: 0.8928
Epoch 33/100
accuracy: 0.8893 - val_loss: 0.5336 - val_accuracy: 0.8924
Epoch 34/100
accuracy: 0.8896 - val_loss: 0.5340 - val_accuracy: 0.8921
Epoch 35/100
accuracy: 0.8894 - val_loss: 0.5355 - val_accuracy: 0.8890
Epoch 36/100
accuracy: 0.8888 - val_loss: 0.5348 - val_accuracy: 0.8913
Epoch 37/100
accuracy: 0.8904 - val_loss: 0.5333 - val_accuracy: 0.8916
Epoch 38/100
accuracy: 0.8891 - val_loss: 0.5336 - val_accuracy: 0.8917
Epoch 39/100
accuracy: 0.8882 - val_loss: 0.5335 - val_accuracy: 0.8910
Epoch 40/100
accuracy: 0.8907 - val_loss: 0.5339 - val_accuracy: 0.8924
Epoch 41/100
accuracy: 0.8900 - val_loss: 0.5337 - val_accuracy: 0.8926
Epoch 42/100
accuracy: 0.8901 - val_loss: 0.5340 - val_accuracy: 0.8913
Epoch 43/100
accuracy: 0.8891 - val_loss: 0.5334 - val_accuracy: 0.8918
Epoch 44/100
accuracy: 0.8901 - val_loss: 0.5335 - val_accuracy: 0.8921
```

```
Epoch 45/100
accuracy: 0.8896 - val_loss: 0.5336 - val_accuracy: 0.8925
Epoch 46/100
accuracy: 0.8903 - val_loss: 0.5333 - val_accuracy: 0.8925
Epoch 47/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5388 -
accuracy: 0.8892 - val_loss: 0.5339 - val_accuracy: 0.8916
Epoch 48/100
accuracy: 0.8909 - val_loss: 0.5335 - val_accuracy: 0.8914
Epoch 49/100
accuracy: 0.8903 - val_loss: 0.5341 - val_accuracy: 0.8906
Epoch 50/100
accuracy: 0.8896 - val_loss: 0.5344 - val_accuracy: 0.8913
Epoch 51/100
1349/1349 [============= - - 2s 1ms/step - loss: 0.5380 -
accuracy: 0.8899 - val_loss: 0.5339 - val_accuracy: 0.8929
Epoch 52/100
accuracy: 0.8899 - val_loss: 0.5341 - val_accuracy: 0.8912
Epoch 53/100
accuracy: 0.8890 - val_loss: 0.5339 - val_accuracy: 0.8915
Epoch 54/100
accuracy: 0.8884 - val_loss: 0.5337 - val_accuracy: 0.8907
Epoch 55/100
accuracy: 0.8896 - val_loss: 0.5339 - val_accuracy: 0.8919
Epoch 56/100
accuracy: 0.8905 - val_loss: 0.5338 - val_accuracy: 0.8921
Epoch 57/100
accuracy: 0.8903 - val_loss: 0.5336 - val_accuracy: 0.8925
Epoch 58/100
accuracy: 0.8904 - val_loss: 0.5334 - val_accuracy: 0.8914
Epoch 59/100
accuracy: 0.8902 - val_loss: 0.5335 - val_accuracy: 0.8923
Epoch 60/100
accuracy: 0.8911 - val_loss: 0.5336 - val_accuracy: 0.8928
```

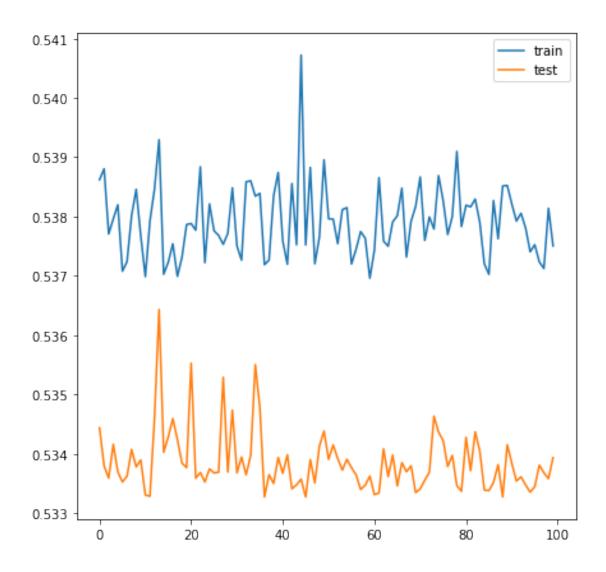
```
Epoch 61/100
accuracy: 0.8910 - val_loss: 0.5333 - val_accuracy: 0.8924
Epoch 62/100
accuracy: 0.8906 - val_loss: 0.5333 - val_accuracy: 0.8920
Epoch 63/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.5376 -
accuracy: 0.8894 - val_loss: 0.5341 - val_accuracy: 0.8897
Epoch 64/100
accuracy: 0.8901 - val_loss: 0.5336 - val_accuracy: 0.8918
Epoch 65/100
accuracy: 0.8895 - val_loss: 0.5340 - val_accuracy: 0.8918
Epoch 66/100
accuracy: 0.8897 - val_loss: 0.5335 - val_accuracy: 0.8924
Epoch 67/100
accuracy: 0.8898 - val_loss: 0.5338 - val_accuracy: 0.8925
Epoch 68/100
accuracy: 0.8894 - val_loss: 0.5337 - val_accuracy: 0.8918
Epoch 69/100
accuracy: 0.8908 - val_loss: 0.5338 - val_accuracy: 0.8931
Epoch 70/100
accuracy: 0.8898 - val_loss: 0.5333 - val_accuracy: 0.8921
Epoch 71/100
accuracy: 0.8899 - val_loss: 0.5334 - val_accuracy: 0.8918
Epoch 72/100
accuracy: 0.8892 - val_loss: 0.5336 - val_accuracy: 0.8912
Epoch 73/100
accuracy: 0.8890 - val_loss: 0.5337 - val_accuracy: 0.8931
Epoch 74/100
accuracy: 0.8892 - val_loss: 0.5346 - val_accuracy: 0.8899
Epoch 75/100
accuracy: 0.8897 - val_loss: 0.5344 - val_accuracy: 0.8901
Epoch 76/100
accuracy: 0.8890 - val_loss: 0.5342 - val_accuracy: 0.8914
```

```
Epoch 77/100
accuracy: 0.8903 - val_loss: 0.5338 - val_accuracy: 0.8920
Epoch 78/100
accuracy: 0.8898 - val_loss: 0.5340 - val_accuracy: 0.8911
Epoch 79/100
1349/1349 [============ ] - 2s 1ms/step - loss: 0.5391 -
accuracy: 0.8883 - val_loss: 0.5335 - val_accuracy: 0.8921
Epoch 80/100
accuracy: 0.8895 - val_loss: 0.5334 - val_accuracy: 0.8911
Epoch 81/100
accuracy: 0.8887 - val_loss: 0.5343 - val_accuracy: 0.8895
Epoch 82/100
accuracy: 0.8887 - val_loss: 0.5337 - val_accuracy: 0.8922
Epoch 83/100
accuracy: 0.8898 - val_loss: 0.5344 - val_accuracy: 0.8919
Epoch 84/100
accuracy: 0.8898 - val_loss: 0.5340 - val_accuracy: 0.8919
Epoch 85/100
accuracy: 0.8907 - val_loss: 0.5334 - val_accuracy: 0.8915
Epoch 86/100
accuracy: 0.8903 - val_loss: 0.5334 - val_accuracy: 0.8920
Epoch 87/100
accuracy: 0.8900 - val_loss: 0.5335 - val_accuracy: 0.8922
Epoch 88/100
accuracy: 0.8904 - val_loss: 0.5338 - val_accuracy: 0.8919
Epoch 89/100
accuracy: 0.8886 - val_loss: 0.5333 - val_accuracy: 0.8910
Epoch 90/100
accuracy: 0.8888 - val_loss: 0.5341 - val_accuracy: 0.8920
Epoch 91/100
accuracy: 0.8899 - val_loss: 0.5338 - val_accuracy: 0.8924
Epoch 92/100
accuracy: 0.8891 - val_loss: 0.5335 - val_accuracy: 0.8918
```

```
Epoch 93/100
   accuracy: 0.8899 - val_loss: 0.5336 - val_accuracy: 0.8920
   Epoch 94/100
   accuracy: 0.8894 - val_loss: 0.5335 - val_accuracy: 0.8922
   Epoch 95/100
   accuracy: 0.8888 - val_loss: 0.5333 - val_accuracy: 0.8925
   Epoch 96/100
   accuracy: 0.8898 - val_loss: 0.5334 - val_accuracy: 0.8915
   Epoch 97/100
   accuracy: 0.8904 - val_loss: 0.5338 - val_accuracy: 0.8916
   Epoch 98/100
   accuracy: 0.8903 - val_loss: 0.5337 - val_accuracy: 0.8925
   Epoch 99/100
   accuracy: 0.8894 - val_loss: 0.5336 - val_accuracy: 0.8918
   Epoch 100/100
   accuracy: 0.8906 - val_loss: 0.5339 - val_accuracy: 0.8927
[89]: len(val accuracy list)
[89]: 10
[90]: plt.figure(figsize=(7, 7))
   plt.plot(history.history['accuracy'], label='train')
   plt.plot(history.history['val_accuracy'], label='test')
   plt.legend()
   plt.savefig('accuracy_entropy.jpg', dpi=200)
```



```
[91]: plt.figure(figsize=(7, 7))
    plt.plot(history.history['loss'], label='train')
    plt.plot(history.history['val_loss'], label='test')
    plt.legend()
    plt.savefig('loss_entropy.jpg', dpi=200)
```



```
val_accuracy_list2 = []
while len(val_accuracy_list2) < 10:</pre>
  history2 = model2.fit(x_train,
              y_train,
              batch_size=50,
              epochs=100,
             validation_data=(x_test, y_test))
  val_accuracy_list2.append(history2.history['val_accuracy'])
Epoch 1/100
accuracy: 0.8344 - val_loss: 0.0935 - val_accuracy: 0.8719
Epoch 2/100
accuracy: 0.8712 - val_loss: 0.0882 - val_accuracy: 0.8749
Epoch 3/100
accuracy: 0.8749 - val_loss: 0.0870 - val_accuracy: 0.8761
Epoch 4/100
accuracy: 0.8781 - val_loss: 0.0848 - val_accuracy: 0.8838
Epoch 5/100
accuracy: 0.8806 - val_loss: 0.0842 - val_accuracy: 0.8874
Epoch 6/100
accuracy: 0.8838 - val_loss: 0.0830 - val_accuracy: 0.8870
Epoch 7/100
accuracy: 0.8854 - val_loss: 0.0814 - val_accuracy: 0.8884
Epoch 8/100
accuracy: 0.8873 - val_loss: 0.0800 - val_accuracy: 0.8904
Epoch 9/100
accuracy: 0.8874 - val_loss: 0.0801 - val_accuracy: 0.8895
Epoch 10/100
accuracy: 0.8895 - val_loss: 0.0797 - val_accuracy: 0.8891
Epoch 11/100
1349/1349 [============ ] - 1s 1ms/step - loss: 0.0825 -
accuracy: 0.8882 - val_loss: 0.0794 - val_accuracy: 0.8897
Epoch 12/100
accuracy: 0.8894 - val_loss: 0.0790 - val_accuracy: 0.8903
Epoch 13/100
```

```
accuracy: 0.8895 - val_loss: 0.0790 - val_accuracy: 0.8898
Epoch 14/100
accuracy: 0.8895 - val_loss: 0.0784 - val_accuracy: 0.8903
Epoch 15/100
accuracy: 0.8889 - val_loss: 0.0796 - val_accuracy: 0.8904
Epoch 16/100
accuracy: 0.8894 - val_loss: 0.0786 - val_accuracy: 0.8899
Epoch 17/100
accuracy: 0.8902 - val_loss: 0.0785 - val_accuracy: 0.8906
Epoch 18/100
accuracy: 0.8902 - val_loss: 0.0780 - val_accuracy: 0.8907
Epoch 19/100
accuracy: 0.8900 - val_loss: 0.0791 - val_accuracy: 0.8891
Epoch 20/100
accuracy: 0.8901 - val_loss: 0.0774 - val_accuracy: 0.8901
Epoch 21/100
accuracy: 0.8892 - val_loss: 0.0769 - val_accuracy: 0.8906
Epoch 22/100
accuracy: 0.8908 - val_loss: 0.0769 - val_accuracy: 0.8900
Epoch 23/100
accuracy: 0.8905 - val_loss: 0.0773 - val_accuracy: 0.8902
Epoch 24/100
accuracy: 0.8905 - val_loss: 0.0773 - val_accuracy: 0.8894
Epoch 25/100
accuracy: 0.8902 - val loss: 0.0765 - val accuracy: 0.8916
Epoch 26/100
accuracy: 0.8900 - val_loss: 0.0770 - val_accuracy: 0.8901
Epoch 27/100
accuracy: 0.8895 - val_loss: 0.0765 - val_accuracy: 0.8906
Epoch 28/100
accuracy: 0.8910 - val_loss: 0.0766 - val_accuracy: 0.8912
Epoch 29/100
```

```
accuracy: 0.8905 - val_loss: 0.0772 - val_accuracy: 0.8893
Epoch 30/100
accuracy: 0.8912 - val_loss: 0.0761 - val_accuracy: 0.8912
Epoch 31/100
accuracy: 0.8910 - val_loss: 0.0765 - val_accuracy: 0.8921
Epoch 32/100
accuracy: 0.8915 - val_loss: 0.0767 - val_accuracy: 0.8910
Epoch 33/100
accuracy: 0.8920 - val_loss: 0.0764 - val_accuracy: 0.8902
Epoch 34/100
accuracy: 0.8908 - val_loss: 0.0758 - val_accuracy: 0.8915
Epoch 35/100
accuracy: 0.8913 - val_loss: 0.0761 - val_accuracy: 0.8918
Epoch 36/100
accuracy: 0.8918 - val_loss: 0.0765 - val_accuracy: 0.8918
Epoch 37/100
accuracy: 0.8919 - val_loss: 0.0756 - val_accuracy: 0.8913
Epoch 38/100
accuracy: 0.8918 - val_loss: 0.0755 - val_accuracy: 0.8915
Epoch 39/100
accuracy: 0.8918 - val_loss: 0.0754 - val_accuracy: 0.8912
Epoch 40/100
accuracy: 0.8907 - val_loss: 0.0754 - val_accuracy: 0.8914
Epoch 41/100
accuracy: 0.8910 - val loss: 0.0756 - val accuracy: 0.8907
Epoch 42/100
accuracy: 0.8911 - val_loss: 0.0759 - val_accuracy: 0.8909
Epoch 43/100
accuracy: 0.8918 - val_loss: 0.0756 - val_accuracy: 0.8911
Epoch 44/100
accuracy: 0.8915 - val_loss: 0.0757 - val_accuracy: 0.8905
Epoch 45/100
```

```
accuracy: 0.8914 - val_loss: 0.0751 - val_accuracy: 0.8923
Epoch 46/100
accuracy: 0.8922 - val_loss: 0.0754 - val_accuracy: 0.8917
Epoch 47/100
accuracy: 0.8927 - val_loss: 0.0782 - val_accuracy: 0.8917
Epoch 48/100
accuracy: 0.8911 - val_loss: 0.0748 - val_accuracy: 0.8916
Epoch 49/100
accuracy: 0.8921 - val_loss: 0.0751 - val_accuracy: 0.8913
Epoch 50/100
accuracy: 0.8906 - val_loss: 0.0752 - val_accuracy: 0.8910
Epoch 51/100
accuracy: 0.8914 - val_loss: 0.0756 - val_accuracy: 0.8912
Epoch 52/100
accuracy: 0.8921 - val_loss: 0.0752 - val_accuracy: 0.8914
Epoch 53/100
accuracy: 0.8907 - val_loss: 0.0759 - val_accuracy: 0.8909
Epoch 54/100
accuracy: 0.8912 - val_loss: 0.0748 - val_accuracy: 0.8912
accuracy: 0.8923 - val_loss: 0.0748 - val_accuracy: 0.8920
Epoch 56/100
accuracy: 0.8910 - val_loss: 0.0752 - val_accuracy: 0.8921
Epoch 57/100
accuracy: 0.8926 - val loss: 0.0752 - val accuracy: 0.8909
Epoch 58/100
accuracy: 0.8918 - val_loss: 0.0747 - val_accuracy: 0.8926
Epoch 59/100
accuracy: 0.8925 - val_loss: 0.0747 - val_accuracy: 0.8920
Epoch 60/100
accuracy: 0.8923 - val_loss: 0.0750 - val_accuracy: 0.8927
Epoch 61/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0790 -
```

```
accuracy: 0.8920 - val_loss: 0.0749 - val_accuracy: 0.8931
Epoch 62/100
accuracy: 0.8911 - val_loss: 0.0747 - val_accuracy: 0.8927
Epoch 63/100
accuracy: 0.8924 - val_loss: 0.0750 - val_accuracy: 0.8921
Epoch 64/100
accuracy: 0.8917 - val_loss: 0.0761 - val_accuracy: 0.8900
Epoch 65/100
accuracy: 0.8921 - val_loss: 0.0756 - val_accuracy: 0.8898
Epoch 66/100
accuracy: 0.8927 - val_loss: 0.0747 - val_accuracy: 0.8927
Epoch 67/100
accuracy: 0.8920 - val_loss: 0.0749 - val_accuracy: 0.8916
Epoch 68/100
accuracy: 0.8912 - val_loss: 0.0763 - val_accuracy: 0.8901
Epoch 69/100
accuracy: 0.8923 - val_loss: 0.0748 - val_accuracy: 0.8918
Epoch 70/100
accuracy: 0.8929 - val_loss: 0.0747 - val_accuracy: 0.8920
Epoch 71/100
accuracy: 0.8902 - val_loss: 0.0752 - val_accuracy: 0.8946
Epoch 72/100
accuracy: 0.8921 - val_loss: 0.0761 - val_accuracy: 0.8899
Epoch 73/100
accuracy: 0.8915 - val loss: 0.0748 - val accuracy: 0.8923
Epoch 74/100
accuracy: 0.8909 - val_loss: 0.0770 - val_accuracy: 0.8941
Epoch 75/100
accuracy: 0.8907 - val_loss: 0.0763 - val_accuracy: 0.8929
Epoch 76/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0799 -
accuracy: 0.8901 - val_loss: 0.0778 - val_accuracy: 0.8912
Epoch 77/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0790 -
```

```
accuracy: 0.8916 - val_loss: 0.0763 - val_accuracy: 0.8923
Epoch 78/100
accuracy: 0.8915 - val_loss: 0.0751 - val_accuracy: 0.8926
Epoch 79/100
accuracy: 0.8909 - val_loss: 0.0787 - val_accuracy: 0.8893
Epoch 80/100
accuracy: 0.8917 - val_loss: 0.0759 - val_accuracy: 0.8932
Epoch 81/100
accuracy: 0.8907 - val_loss: 0.0746 - val_accuracy: 0.8947
Epoch 82/100
accuracy: 0.8918 - val_loss: 0.0752 - val_accuracy: 0.8971
Epoch 83/100
accuracy: 0.8907 - val_loss: 0.0753 - val_accuracy: 0.8935
Epoch 84/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0786 -
accuracy: 0.8918 - val_loss: 0.0746 - val_accuracy: 0.8941
Epoch 85/100
accuracy: 0.8913 - val_loss: 0.0748 - val_accuracy: 0.8959
Epoch 86/100
accuracy: 0.8926 - val_loss: 0.0748 - val_accuracy: 0.8966
Epoch 87/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0785 -
accuracy: 0.8918 - val_loss: 0.0744 - val_accuracy: 0.8969
Epoch 88/100
accuracy: 0.8924 - val_loss: 0.0747 - val_accuracy: 0.8964
Epoch 89/100
accuracy: 0.8924 - val loss: 0.0746 - val accuracy: 0.8956
Epoch 90/100
accuracy: 0.8920 - val_loss: 0.0741 - val_accuracy: 0.8965
Epoch 91/100
accuracy: 0.8920 - val_loss: 0.0737 - val_accuracy: 0.8980
Epoch 92/100
accuracy: 0.8919 - val_loss: 0.0743 - val_accuracy: 0.8976
Epoch 93/100
```

```
accuracy: 0.8919 - val_loss: 0.0745 - val_accuracy: 0.8963
Epoch 94/100
accuracy: 0.8931 - val_loss: 0.0737 - val_accuracy: 0.8974
Epoch 95/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0779 -
accuracy: 0.8929 - val_loss: 0.0744 - val_accuracy: 0.8975
Epoch 96/100
accuracy: 0.8920 - val_loss: 0.0734 - val_accuracy: 0.8974
Epoch 97/100
accuracy: 0.8924 - val_loss: 0.0737 - val_accuracy: 0.8973
Epoch 98/100
accuracy: 0.8929 - val_loss: 0.0736 - val_accuracy: 0.8982
Epoch 99/100
accuracy: 0.8934 - val_loss: 0.0746 - val_accuracy: 0.8939
Epoch 100/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0779 -
accuracy: 0.8919 - val_loss: 0.0732 - val_accuracy: 0.8994
Epoch 1/100
accuracy: 0.8924 - val_loss: 0.0748 - val_accuracy: 0.8964
Epoch 2/100
accuracy: 0.8930 - val_loss: 0.0739 - val_accuracy: 0.8966
accuracy: 0.8935 - val_loss: 0.0748 - val_accuracy: 0.8967
Epoch 4/100
accuracy: 0.8918 - val_loss: 0.0733 - val_accuracy: 0.8974
Epoch 5/100
accuracy: 0.8924 - val loss: 0.0743 - val accuracy: 0.8955
Epoch 6/100
accuracy: 0.8930 - val_loss: 0.0740 - val_accuracy: 0.8967
Epoch 7/100
accuracy: 0.8916 - val_loss: 0.0740 - val_accuracy: 0.8967
Epoch 8/100
accuracy: 0.8932 - val_loss: 0.0729 - val_accuracy: 0.8987
Epoch 9/100
```

```
accuracy: 0.8929 - val_loss: 0.0734 - val_accuracy: 0.8976
Epoch 10/100
accuracy: 0.8925 - val_loss: 0.0733 - val_accuracy: 0.8982
Epoch 11/100
accuracy: 0.8941 - val_loss: 0.0739 - val_accuracy: 0.8967
Epoch 12/100
accuracy: 0.8930 - val_loss: 0.0735 - val_accuracy: 0.8970
Epoch 13/100
accuracy: 0.8924 - val_loss: 0.0730 - val_accuracy: 0.8996
Epoch 14/100
accuracy: 0.8923 - val_loss: 0.0739 - val_accuracy: 0.8948
Epoch 15/100
accuracy: 0.8928 - val_loss: 0.0751 - val_accuracy: 0.8950
Epoch 16/100
1349/1349 [============= - - 1s 1ms/step - loss: 0.0781 -
accuracy: 0.8916 - val_loss: 0.0736 - val_accuracy: 0.8968
Epoch 17/100
accuracy: 0.8935 - val_loss: 0.0749 - val_accuracy: 0.8949
Epoch 18/100
accuracy: 0.8928 - val_loss: 0.0741 - val_accuracy: 0.8962
Epoch 19/100
accuracy: 0.8932 - val_loss: 0.0732 - val_accuracy: 0.8973
Epoch 20/100
accuracy: 0.8937 - val_loss: 0.0732 - val_accuracy: 0.8970
Epoch 21/100
accuracy: 0.8933 - val loss: 0.0734 - val accuracy: 0.8981
Epoch 22/100
accuracy: 0.8921 - val_loss: 0.0733 - val_accuracy: 0.8972
Epoch 23/100
accuracy: 0.8916 - val_loss: 0.0731 - val_accuracy: 0.8983
Epoch 24/100
accuracy: 0.8919 - val_loss: 0.0739 - val_accuracy: 0.8971
Epoch 25/100
```

```
accuracy: 0.8936 - val_loss: 0.0731 - val_accuracy: 0.8978
Epoch 26/100
accuracy: 0.8936 - val_loss: 0.0736 - val_accuracy: 0.8965
Epoch 27/100
accuracy: 0.8931 - val_loss: 0.0737 - val_accuracy: 0.8982
Epoch 28/100
accuracy: 0.8920 - val_loss: 0.0740 - val_accuracy: 0.8987
Epoch 29/100
accuracy: 0.8933 - val_loss: 0.0737 - val_accuracy: 0.8955
Epoch 30/100
accuracy: 0.8931 - val_loss: 0.0736 - val_accuracy: 0.9002
Epoch 31/100
accuracy: 0.8936 - val_loss: 0.0745 - val_accuracy: 0.8983
Epoch 32/100
accuracy: 0.8929 - val_loss: 0.0730 - val_accuracy: 0.8989
Epoch 33/100
accuracy: 0.8934 - val_loss: 0.0736 - val_accuracy: 0.8964
Epoch 34/100
accuracy: 0.8935 - val_loss: 0.0747 - val_accuracy: 0.8980
accuracy: 0.8914 - val_loss: 0.0738 - val_accuracy: 0.8960
Epoch 36/100
accuracy: 0.8929 - val_loss: 0.0736 - val_accuracy: 0.8994
Epoch 37/100
accuracy: 0.8921 - val loss: 0.0736 - val accuracy: 0.8960
Epoch 38/100
accuracy: 0.8927 - val_loss: 0.0728 - val_accuracy: 0.8976
Epoch 39/100
accuracy: 0.8922 - val_loss: 0.0733 - val_accuracy: 0.8985
Epoch 40/100
accuracy: 0.8926 - val_loss: 0.0737 - val_accuracy: 0.8957
Epoch 41/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0778 -
```

```
accuracy: 0.8930 - val_loss: 0.0733 - val_accuracy: 0.8960
Epoch 42/100
accuracy: 0.8923 - val_loss: 0.0735 - val_accuracy: 0.8969
Epoch 43/100
accuracy: 0.8922 - val_loss: 0.0738 - val_accuracy: 0.8979
Epoch 44/100
accuracy: 0.8919 - val_loss: 0.0730 - val_accuracy: 0.8994
Epoch 45/100
accuracy: 0.8936 - val_loss: 0.0734 - val_accuracy: 0.8960
Epoch 46/100
accuracy: 0.8933 - val_loss: 0.0740 - val_accuracy: 0.8954
Epoch 47/100
accuracy: 0.8930 - val_loss: 0.0731 - val_accuracy: 0.8970
Epoch 48/100
accuracy: 0.8928 - val_loss: 0.0734 - val_accuracy: 0.8975
Epoch 49/100
accuracy: 0.8921 - val_loss: 0.0726 - val_accuracy: 0.9003
Epoch 50/100
accuracy: 0.8918 - val_loss: 0.0742 - val_accuracy: 0.8980
Epoch 51/100
accuracy: 0.8920 - val_loss: 0.0743 - val_accuracy: 0.8969
Epoch 52/100
accuracy: 0.8929 - val_loss: 0.0729 - val_accuracy: 0.8971
Epoch 53/100
accuracy: 0.8936 - val loss: 0.0739 - val accuracy: 0.8967
Epoch 54/100
accuracy: 0.8927 - val_loss: 0.0739 - val_accuracy: 0.8958
Epoch 55/100
accuracy: 0.8920 - val_loss: 0.0736 - val_accuracy: 0.8972
Epoch 56/100
accuracy: 0.8927 - val_loss: 0.0742 - val_accuracy: 0.8962
Epoch 57/100
```

```
accuracy: 0.8923 - val_loss: 0.0753 - val_accuracy: 0.8925
Epoch 58/100
accuracy: 0.8930 - val_loss: 0.0735 - val_accuracy: 0.8957
Epoch 59/100
accuracy: 0.8932 - val_loss: 0.0736 - val_accuracy: 0.8963
Epoch 60/100
accuracy: 0.8931 - val_loss: 0.0738 - val_accuracy: 0.8963
Epoch 61/100
accuracy: 0.8931 - val_loss: 0.0739 - val_accuracy: 0.8967
Epoch 62/100
accuracy: 0.8921 - val_loss: 0.0733 - val_accuracy: 0.8993
Epoch 63/100
accuracy: 0.8942 - val_loss: 0.0731 - val_accuracy: 0.8982
Epoch 64/100
accuracy: 0.8943 - val_loss: 0.0731 - val_accuracy: 0.8972
Epoch 65/100
accuracy: 0.8930 - val_loss: 0.0741 - val_accuracy: 0.8976
Epoch 66/100
accuracy: 0.8937 - val_loss: 0.0734 - val_accuracy: 0.8976
Epoch 67/100
accuracy: 0.8930 - val_loss: 0.0733 - val_accuracy: 0.8957
Epoch 68/100
accuracy: 0.8930 - val_loss: 0.0733 - val_accuracy: 0.8974
Epoch 69/100
accuracy: 0.8932 - val loss: 0.0732 - val accuracy: 0.8965
Epoch 70/100
accuracy: 0.8924 - val_loss: 0.0732 - val_accuracy: 0.8969
Epoch 71/100
accuracy: 0.8940 - val_loss: 0.0733 - val_accuracy: 0.8976
Epoch 72/100
accuracy: 0.8933 - val_loss: 0.0736 - val_accuracy: 0.8978
Epoch 73/100
```

```
accuracy: 0.8921 - val_loss: 0.0738 - val_accuracy: 0.8952
Epoch 74/100
accuracy: 0.8937 - val_loss: 0.0737 - val_accuracy: 0.8963
Epoch 75/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0774 -
accuracy: 0.8934 - val_loss: 0.0730 - val_accuracy: 0.8969
Epoch 76/100
accuracy: 0.8947 - val_loss: 0.0739 - val_accuracy: 0.8976
Epoch 77/100
accuracy: 0.8925 - val_loss: 0.0736 - val_accuracy: 0.8966
Epoch 78/100
accuracy: 0.8932 - val_loss: 0.0732 - val_accuracy: 0.8982
Epoch 79/100
accuracy: 0.8940 - val_loss: 0.0732 - val_accuracy: 0.8978
Epoch 80/100
accuracy: 0.8941 - val_loss: 0.0737 - val_accuracy: 0.8968
Epoch 81/100
accuracy: 0.8935 - val_loss: 0.0730 - val_accuracy: 0.8981
Epoch 82/100
accuracy: 0.8927 - val_loss: 0.0734 - val_accuracy: 0.8969
Epoch 83/100
accuracy: 0.8931 - val_loss: 0.0727 - val_accuracy: 0.8981
Epoch 84/100
accuracy: 0.8935 - val_loss: 0.0740 - val_accuracy: 0.8979
Epoch 85/100
accuracy: 0.8928 - val loss: 0.0734 - val accuracy: 0.8979
Epoch 86/100
accuracy: 0.8929 - val_loss: 0.0732 - val_accuracy: 0.8974
Epoch 87/100
accuracy: 0.8925 - val_loss: 0.0731 - val_accuracy: 0.8973
Epoch 88/100
accuracy: 0.8942 - val_loss: 0.0736 - val_accuracy: 0.8960
Epoch 89/100
```

```
accuracy: 0.8916 - val_loss: 0.0736 - val_accuracy: 0.8966
Epoch 90/100
accuracy: 0.8947 - val_loss: 0.0731 - val_accuracy: 0.8971
Epoch 91/100
accuracy: 0.8944 - val_loss: 0.0733 - val_accuracy: 0.8952
Epoch 92/100
accuracy: 0.8936 - val_loss: 0.0744 - val_accuracy: 0.8947
Epoch 93/100
accuracy: 0.8928 - val_loss: 0.0736 - val_accuracy: 0.8980
Epoch 94/100
accuracy: 0.8941 - val_loss: 0.0736 - val_accuracy: 0.8956
Epoch 95/100
accuracy: 0.8941 - val_loss: 0.0738 - val_accuracy: 0.8961
Epoch 96/100
accuracy: 0.8933 - val_loss: 0.0745 - val_accuracy: 0.8954
Epoch 97/100
accuracy: 0.8938 - val_loss: 0.0738 - val_accuracy: 0.8956
Epoch 98/100
accuracy: 0.8945 - val_loss: 0.0725 - val_accuracy: 0.8987
Epoch 99/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0790 -
accuracy: 0.8929 - val_loss: 0.0735 - val_accuracy: 0.8983
Epoch 100/100
accuracy: 0.8933 - val_loss: 0.0731 - val_accuracy: 0.8981
Epoch 1/100
accuracy: 0.8941 - val loss: 0.0733 - val accuracy: 0.8972
Epoch 2/100
accuracy: 0.8943 - val_loss: 0.0744 - val_accuracy: 0.8940
Epoch 3/100
accuracy: 0.8931 - val_loss: 0.0740 - val_accuracy: 0.8970
Epoch 4/100
accuracy: 0.8939 - val_loss: 0.0733 - val_accuracy: 0.8984
Epoch 5/100
```

```
accuracy: 0.8952 - val_loss: 0.0738 - val_accuracy: 0.8972
Epoch 6/100
accuracy: 0.8939 - val_loss: 0.0737 - val_accuracy: 0.8972
Epoch 7/100
accuracy: 0.8934 - val_loss: 0.0734 - val_accuracy: 0.8961
Epoch 8/100
accuracy: 0.8933 - val_loss: 0.0732 - val_accuracy: 0.8981
Epoch 9/100
accuracy: 0.8943 - val_loss: 0.0730 - val_accuracy: 0.8986
Epoch 10/100
1349/1349 [============= ] - 1s 1ms/step - loss: 0.0775 -
accuracy: 0.8926 - val_loss: 0.0738 - val_accuracy: 0.8970
Epoch 11/100
accuracy: 0.8946 - val_loss: 0.0731 - val_accuracy: 0.8983
Epoch 12/100
accuracy: 0.8944 - val_loss: 0.0731 - val_accuracy: 0.8995
Epoch 13/100
accuracy: 0.8941 - val_loss: 0.0727 - val_accuracy: 0.8996
Epoch 14/100
accuracy: 0.8931 - val_loss: 0.0728 - val_accuracy: 0.8979
accuracy: 0.8931 - val_loss: 0.0730 - val_accuracy: 0.8978
Epoch 16/100
accuracy: 0.8928 - val_loss: 0.0737 - val_accuracy: 0.8969
Epoch 17/100
accuracy: 0.8934 - val loss: 0.0734 - val accuracy: 0.8986
Epoch 18/100
accuracy: 0.8935 - val_loss: 0.0735 - val_accuracy: 0.8961
Epoch 19/100
accuracy: 0.8939 - val_loss: 0.0732 - val_accuracy: 0.8979
Epoch 20/100
accuracy: 0.8927 - val_loss: 0.0731 - val_accuracy: 0.8978
Epoch 21/100
```

```
accuracy: 0.8941 - val_loss: 0.0731 - val_accuracy: 0.8970
Epoch 22/100
accuracy: 0.8924 - val_loss: 0.0746 - val_accuracy: 0.8954
Epoch 23/100
accuracy: 0.8942 - val_loss: 0.0733 - val_accuracy: 0.8966
Epoch 24/100
accuracy: 0.8940 - val_loss: 0.0736 - val_accuracy: 0.8981
Epoch 25/100
accuracy: 0.8930 - val_loss: 0.0738 - val_accuracy: 0.8975
Epoch 26/100
accuracy: 0.8929 - val_loss: 0.0728 - val_accuracy: 0.8994
Epoch 27/100
accuracy: 0.8937 - val_loss: 0.0731 - val_accuracy: 0.8990
Epoch 28/100
1349/1349 [============== ] - 1s 1ms/step - loss: 0.0775 -
accuracy: 0.8940 - val_loss: 0.0739 - val_accuracy: 0.8980
Epoch 29/100
accuracy: 0.8938 - val_loss: 0.0735 - val_accuracy: 0.8968
Epoch 30/100
accuracy: 0.8939 - val_loss: 0.0729 - val_accuracy: 0.8993
Epoch 31/100
accuracy: 0.8934 - val_loss: 0.0733 - val_accuracy: 0.8969
Epoch 32/100
accuracy: 0.8937 - val_loss: 0.0737 - val_accuracy: 0.8975
Epoch 33/100
accuracy: 0.8936 - val loss: 0.0740 - val accuracy: 0.8942
Epoch 34/100
accuracy: 0.8943 - val_loss: 0.0732 - val_accuracy: 0.8985
Epoch 35/100
accuracy: 0.8934 - val_loss: 0.0736 - val_accuracy: 0.8981
Epoch 36/100
accuracy: 0.8934 - val_loss: 0.0730 - val_accuracy: 0.8978
Epoch 37/100
```

```
accuracy: 0.8934 - val_loss: 0.0729 - val_accuracy: 0.8987
Epoch 38/100
accuracy: 0.8941 - val_loss: 0.0732 - val_accuracy: 0.8982
Epoch 39/100
accuracy: 0.8940 - val_loss: 0.0734 - val_accuracy: 0.8955
Epoch 40/100
accuracy: 0.8937 - val_loss: 0.0731 - val_accuracy: 0.8975
Epoch 41/100
accuracy: 0.8931 - val_loss: 0.0730 - val_accuracy: 0.8981
Epoch 42/100
accuracy: 0.8940 - val_loss: 0.0739 - val_accuracy: 0.8966
Epoch 43/100
accuracy: 0.8933 - val_loss: 0.0734 - val_accuracy: 0.8968
Epoch 44/100
accuracy: 0.8930 - val_loss: 0.0733 - val_accuracy: 0.8971
Epoch 45/100
accuracy: 0.8927 - val_loss: 0.0732 - val_accuracy: 0.8973
Epoch 46/100
accuracy: 0.8941 - val_loss: 0.0739 - val_accuracy: 0.8982
Epoch 47/100
accuracy: 0.8943 - val_loss: 0.0726 - val_accuracy: 0.8970
Epoch 48/100
accuracy: 0.8935 - val_loss: 0.0730 - val_accuracy: 0.8971
Epoch 49/100
accuracy: 0.8937 - val_loss: 0.0732 - val_accuracy: 0.8967
Epoch 50/100
accuracy: 0.8939 - val_loss: 0.0725 - val_accuracy: 0.8980
Epoch 51/100
accuracy: 0.8940 - val_loss: 0.0737 - val_accuracy: 0.8962
Epoch 52/100
accuracy: 0.8937 - val_loss: 0.0735 - val_accuracy: 0.8971
Epoch 53/100
```

```
accuracy: 0.8941 - val_loss: 0.0741 - val_accuracy: 0.8978
Epoch 54/100
accuracy: 0.8944 - val_loss: 0.0734 - val_accuracy: 0.8955
Epoch 55/100
accuracy: 0.8933 - val_loss: 0.0730 - val_accuracy: 0.8977
Epoch 56/100
accuracy: 0.8940 - val_loss: 0.0732 - val_accuracy: 0.8966
Epoch 57/100
accuracy: 0.8936 - val_loss: 0.0727 - val_accuracy: 0.8976
Epoch 58/100
accuracy: 0.8953 - val_loss: 0.0734 - val_accuracy: 0.8948
Epoch 59/100
accuracy: 0.8926 - val_loss: 0.0726 - val_accuracy: 0.8988
Epoch 60/100
accuracy: 0.8931 - val_loss: 0.0736 - val_accuracy: 0.8978
Epoch 61/100
accuracy: 0.8926 - val_loss: 0.0727 - val_accuracy: 0.8974
Epoch 62/100
accuracy: 0.8941 - val_loss: 0.0731 - val_accuracy: 0.8978
Epoch 63/100
accuracy: 0.8935 - val_loss: 0.0729 - val_accuracy: 0.8991
Epoch 64/100
accuracy: 0.8944 - val_loss: 0.0731 - val_accuracy: 0.8959
Epoch 65/100
accuracy: 0.8936 - val loss: 0.0733 - val accuracy: 0.8955
Epoch 66/100
accuracy: 0.8941 - val_loss: 0.0732 - val_accuracy: 0.8975
Epoch 67/100
accuracy: 0.8930 - val_loss: 0.0734 - val_accuracy: 0.8991
Epoch 68/100
accuracy: 0.8939 - val_loss: 0.0726 - val_accuracy: 0.8981
Epoch 69/100
```

```
accuracy: 0.8940 - val_loss: 0.0728 - val_accuracy: 0.8986
Epoch 70/100
accuracy: 0.8924 - val_loss: 0.0728 - val_accuracy: 0.8977
Epoch 71/100
accuracy: 0.8929 - val_loss: 0.0729 - val_accuracy: 0.8986
Epoch 72/100
accuracy: 0.8930 - val_loss: 0.0732 - val_accuracy: 0.8957
Epoch 73/100
accuracy: 0.8936 - val_loss: 0.0731 - val_accuracy: 0.8974
Epoch 74/100
accuracy: 0.8947 - val_loss: 0.0727 - val_accuracy: 0.8993
Epoch 75/100
accuracy: 0.8932 - val_loss: 0.0735 - val_accuracy: 0.8988
Epoch 76/100
accuracy: 0.8944 - val_loss: 0.0732 - val_accuracy: 0.8970
Epoch 77/100
accuracy: 0.8935 - val_loss: 0.0731 - val_accuracy: 0.8958
Epoch 78/100
accuracy: 0.8932 - val_loss: 0.0725 - val_accuracy: 0.8984
Epoch 79/100
1349/1349 [============= ] - 1s 1ms/step - loss: 0.0765 -
accuracy: 0.8949 - val_loss: 0.0730 - val_accuracy: 0.8999
Epoch 80/100
accuracy: 0.8941 - val_loss: 0.0730 - val_accuracy: 0.8975
Epoch 81/100
accuracy: 0.8927 - val_loss: 0.0735 - val_accuracy: 0.8966
Epoch 82/100
accuracy: 0.8954 - val_loss: 0.0732 - val_accuracy: 0.8976
Epoch 83/100
accuracy: 0.8930 - val_loss: 0.0728 - val_accuracy: 0.8986
Epoch 84/100
accuracy: 0.8933 - val_loss: 0.0728 - val_accuracy: 0.8986
Epoch 85/100
```

```
accuracy: 0.8935 - val_loss: 0.0730 - val_accuracy: 0.8968
Epoch 86/100
accuracy: 0.8942 - val_loss: 0.0747 - val_accuracy: 0.8959
Epoch 87/100
accuracy: 0.8927 - val_loss: 0.0725 - val_accuracy: 0.8985
Epoch 88/100
accuracy: 0.8939 - val_loss: 0.0728 - val_accuracy: 0.8969
Epoch 89/100
accuracy: 0.8949 - val_loss: 0.0731 - val_accuracy: 0.8974
Epoch 90/100
accuracy: 0.8938 - val_loss: 0.0732 - val_accuracy: 0.8972
Epoch 91/100
accuracy: 0.8939 - val_loss: 0.0735 - val_accuracy: 0.8972
Epoch 92/100
accuracy: 0.8931 - val_loss: 0.0733 - val_accuracy: 0.8967
Epoch 93/100
accuracy: 0.8930 - val_loss: 0.0727 - val_accuracy: 0.8993
Epoch 94/100
accuracy: 0.8937 - val_loss: 0.0731 - val_accuracy: 0.8976
accuracy: 0.8941 - val_loss: 0.0732 - val_accuracy: 0.8970
Epoch 96/100
accuracy: 0.8932 - val_loss: 0.0731 - val_accuracy: 0.8965
Epoch 97/100
accuracy: 0.8928 - val loss: 0.0734 - val accuracy: 0.8976
Epoch 98/100
accuracy: 0.8927 - val_loss: 0.0731 - val_accuracy: 0.8991
Epoch 99/100
accuracy: 0.8936 - val_loss: 0.0730 - val_accuracy: 0.9004
Epoch 100/100
1349/1349 [============= ] - 1s 1ms/step - loss: 0.0768 -
accuracy: 0.8939 - val_loss: 0.0729 - val_accuracy: 0.8963
Epoch 1/100
```

```
accuracy: 0.8937 - val_loss: 0.0734 - val_accuracy: 0.8948
Epoch 2/100
accuracy: 0.8944 - val_loss: 0.0742 - val_accuracy: 0.8971
Epoch 3/100
accuracy: 0.8942 - val_loss: 0.0741 - val_accuracy: 0.8965
Epoch 4/100
accuracy: 0.8941 - val_loss: 0.0736 - val_accuracy: 0.8977
Epoch 5/100
accuracy: 0.8943 - val_loss: 0.0748 - val_accuracy: 0.8952
Epoch 6/100
accuracy: 0.8944 - val_loss: 0.0735 - val_accuracy: 0.8962
Epoch 7/100
accuracy: 0.8937 - val_loss: 0.0735 - val_accuracy: 0.8968
Epoch 8/100
accuracy: 0.8934 - val_loss: 0.0733 - val_accuracy: 0.8971
Epoch 9/100
accuracy: 0.8937 - val_loss: 0.0736 - val_accuracy: 0.8982
Epoch 10/100
accuracy: 0.8939 - val_loss: 0.0731 - val_accuracy: 0.8978
accuracy: 0.8937 - val_loss: 0.0751 - val_accuracy: 0.8954
Epoch 12/100
accuracy: 0.8922 - val_loss: 0.0729 - val_accuracy: 0.8984
Epoch 13/100
accuracy: 0.8939 - val loss: 0.0733 - val accuracy: 0.8969
Epoch 14/100
accuracy: 0.8934 - val_loss: 0.0735 - val_accuracy: 0.8962
Epoch 15/100
accuracy: 0.8941 - val_loss: 0.0734 - val_accuracy: 0.8974
Epoch 16/100
accuracy: 0.8931 - val_loss: 0.0734 - val_accuracy: 0.8966
Epoch 17/100
```

```
accuracy: 0.8933 - val_loss: 0.0744 - val_accuracy: 0.8962
Epoch 18/100
accuracy: 0.8941 - val_loss: 0.0735 - val_accuracy: 0.8982
Epoch 19/100
accuracy: 0.8944 - val_loss: 0.0729 - val_accuracy: 0.8993
Epoch 20/100
accuracy: 0.8937 - val_loss: 0.0733 - val_accuracy: 0.8957
Epoch 21/100
accuracy: 0.8936 - val_loss: 0.0731 - val_accuracy: 0.8978
Epoch 22/100
accuracy: 0.8947 - val_loss: 0.0727 - val_accuracy: 0.8982
Epoch 23/100
accuracy: 0.8936 - val_loss: 0.0732 - val_accuracy: 0.8976
Epoch 24/100
accuracy: 0.8940 - val_loss: 0.0733 - val_accuracy: 0.8981
Epoch 25/100
accuracy: 0.8940 - val_loss: 0.0734 - val_accuracy: 0.8974
Epoch 26/100
accuracy: 0.8935 - val_loss: 0.0733 - val_accuracy: 0.8973
Epoch 27/100
accuracy: 0.8937 - val_loss: 0.0726 - val_accuracy: 0.8980
Epoch 28/100
accuracy: 0.8936 - val_loss: 0.0727 - val_accuracy: 0.8974
Epoch 29/100
accuracy: 0.8945 - val loss: 0.0731 - val accuracy: 0.8984
Epoch 30/100
accuracy: 0.8940 - val_loss: 0.0733 - val_accuracy: 0.8960
Epoch 31/100
accuracy: 0.8937 - val_loss: 0.0737 - val_accuracy: 0.8966
Epoch 32/100
accuracy: 0.8936 - val_loss: 0.0731 - val_accuracy: 0.8972
Epoch 33/100
```

```
accuracy: 0.8947 - val_loss: 0.0731 - val_accuracy: 0.8957
Epoch 34/100
accuracy: 0.8941 - val_loss: 0.0731 - val_accuracy: 0.8977
Epoch 35/100
accuracy: 0.8937 - val_loss: 0.0730 - val_accuracy: 0.8977
Epoch 36/100
accuracy: 0.8929 - val_loss: 0.0732 - val_accuracy: 0.8966
Epoch 37/100
accuracy: 0.8936 - val_loss: 0.0725 - val_accuracy: 0.8985
Epoch 38/100
accuracy: 0.8935 - val_loss: 0.0736 - val_accuracy: 0.8955
Epoch 39/100
accuracy: 0.8937 - val_loss: 0.0725 - val_accuracy: 0.8977
Epoch 40/100
accuracy: 0.8944 - val_loss: 0.0730 - val_accuracy: 0.8988
Epoch 41/100
accuracy: 0.8939 - val_loss: 0.0733 - val_accuracy: 0.8977
Epoch 42/100
accuracy: 0.8940 - val_loss: 0.0730 - val_accuracy: 0.8972
Epoch 43/100
accuracy: 0.8933 - val_loss: 0.0727 - val_accuracy: 0.8980
Epoch 44/100
accuracy: 0.8939 - val_loss: 0.0735 - val_accuracy: 0.8977
Epoch 45/100
accuracy: 0.8936 - val loss: 0.0739 - val accuracy: 0.8955
Epoch 46/100
accuracy: 0.8933 - val_loss: 0.0725 - val_accuracy: 0.8971
Epoch 47/100
accuracy: 0.8932 - val_loss: 0.0735 - val_accuracy: 0.8966
Epoch 48/100
1349/1349 [============= ] - 1s 1ms/step - loss: 0.0768 -
accuracy: 0.8938 - val_loss: 0.0745 - val_accuracy: 0.8945
Epoch 49/100
```

```
accuracy: 0.8937 - val_loss: 0.0743 - val_accuracy: 0.8936
Epoch 50/100
accuracy: 0.8927 - val_loss: 0.0733 - val_accuracy: 0.8974
Epoch 51/100
accuracy: 0.8938 - val_loss: 0.0721 - val_accuracy: 0.8987
Epoch 52/100
accuracy: 0.8939 - val_loss: 0.0732 - val_accuracy: 0.8962
Epoch 53/100
accuracy: 0.8938 - val_loss: 0.0730 - val_accuracy: 0.8965
Epoch 54/100
accuracy: 0.8940 - val_loss: 0.0729 - val_accuracy: 0.8981
Epoch 55/100
accuracy: 0.8942 - val_loss: 0.0743 - val_accuracy: 0.8995
Epoch 56/100
accuracy: 0.8944 - val_loss: 0.0726 - val_accuracy: 0.8976
Epoch 57/100
accuracy: 0.8943 - val_loss: 0.0730 - val_accuracy: 0.8974
Epoch 58/100
accuracy: 0.8933 - val_loss: 0.0724 - val_accuracy: 0.8997
Epoch 59/100
accuracy: 0.8949 - val_loss: 0.0726 - val_accuracy: 0.8982
Epoch 60/100
accuracy: 0.8936 - val_loss: 0.0724 - val_accuracy: 0.8983
Epoch 61/100
accuracy: 0.8930 - val loss: 0.0733 - val accuracy: 0.8979
Epoch 62/100
accuracy: 0.8939 - val_loss: 0.0733 - val_accuracy: 0.8988
Epoch 63/100
accuracy: 0.8940 - val_loss: 0.0723 - val_accuracy: 0.8969
Epoch 64/100
accuracy: 0.8938 - val_loss: 0.0735 - val_accuracy: 0.8961
Epoch 65/100
```

```
accuracy: 0.8923 - val_loss: 0.0723 - val_accuracy: 0.8998
Epoch 66/100
accuracy: 0.8944 - val_loss: 0.0727 - val_accuracy: 0.8995
Epoch 67/100
accuracy: 0.8942 - val_loss: 0.0732 - val_accuracy: 0.8955
Epoch 68/100
accuracy: 0.8941 - val_loss: 0.0730 - val_accuracy: 0.8988
Epoch 69/100
accuracy: 0.8938 - val_loss: 0.0724 - val_accuracy: 0.9002
Epoch 70/100
accuracy: 0.8932 - val_loss: 0.0730 - val_accuracy: 0.8954
Epoch 71/100
accuracy: 0.8940 - val_loss: 0.0732 - val_accuracy: 0.8974
Epoch 72/100
accuracy: 0.8931 - val_loss: 0.0727 - val_accuracy: 0.8995
Epoch 73/100
accuracy: 0.8942 - val_loss: 0.0725 - val_accuracy: 0.8984
Epoch 74/100
accuracy: 0.8937 - val_loss: 0.0727 - val_accuracy: 0.8977
1349/1349 [============= ] - 1s 1ms/step - loss: 0.0765 -
accuracy: 0.8935 - val_loss: 0.0726 - val_accuracy: 0.8978
Epoch 76/100
accuracy: 0.8942 - val_loss: 0.0732 - val_accuracy: 0.8956
Epoch 77/100
accuracy: 0.8935 - val loss: 0.0730 - val accuracy: 0.8974
Epoch 78/100
accuracy: 0.8928 - val_loss: 0.0727 - val_accuracy: 0.8984
Epoch 79/100
accuracy: 0.8945 - val_loss: 0.0729 - val_accuracy: 0.8984
Epoch 80/100
1349/1349 [============= ] - 1s 1ms/step - loss: 0.0766 -
accuracy: 0.8952 - val_loss: 0.0728 - val_accuracy: 0.8996
Epoch 81/100
```

```
accuracy: 0.8942 - val_loss: 0.0736 - val_accuracy: 0.8966
Epoch 82/100
accuracy: 0.8953 - val_loss: 0.0730 - val_accuracy: 0.8977
Epoch 83/100
accuracy: 0.8938 - val_loss: 0.0739 - val_accuracy: 0.8968
Epoch 84/100
accuracy: 0.8941 - val_loss: 0.0730 - val_accuracy: 0.8975
Epoch 85/100
accuracy: 0.8939 - val_loss: 0.0728 - val_accuracy: 0.8995
Epoch 86/100
accuracy: 0.8934 - val_loss: 0.0730 - val_accuracy: 0.8964
Epoch 87/100
accuracy: 0.8951 - val_loss: 0.0730 - val_accuracy: 0.8977
Epoch 88/100
accuracy: 0.8944 - val_loss: 0.0725 - val_accuracy: 0.8977
Epoch 89/100
accuracy: 0.8939 - val_loss: 0.0733 - val_accuracy: 0.8998
Epoch 90/100
accuracy: 0.8942 - val_loss: 0.0728 - val_accuracy: 0.8990
Epoch 91/100
accuracy: 0.8932 - val_loss: 0.0727 - val_accuracy: 0.8997
Epoch 92/100
accuracy: 0.8954 - val_loss: 0.0732 - val_accuracy: 0.8968
Epoch 93/100
accuracy: 0.8936 - val loss: 0.0728 - val accuracy: 0.8976
Epoch 94/100
accuracy: 0.8941 - val_loss: 0.0734 - val_accuracy: 0.8996
Epoch 95/100
accuracy: 0.8942 - val_loss: 0.0726 - val_accuracy: 0.9000
Epoch 96/100
accuracy: 0.8945 - val_loss: 0.0726 - val_accuracy: 0.9005
Epoch 97/100
```

```
accuracy: 0.8939 - val_loss: 0.0731 - val_accuracy: 0.8994
Epoch 98/100
accuracy: 0.8951 - val_loss: 0.0732 - val_accuracy: 0.8983
Epoch 99/100
accuracy: 0.8954 - val_loss: 0.0723 - val_accuracy: 0.8985
Epoch 100/100
1349/1349 [============= ] - 2s 1ms/step - loss: 0.0767 -
accuracy: 0.8939 - val_loss: 0.0731 - val_accuracy: 0.8967
Epoch 1/100
accuracy: 0.8940 - val_loss: 0.0727 - val_accuracy: 0.8995
Epoch 2/100
accuracy: 0.8948 - val_loss: 0.0725 - val_accuracy: 0.8975
Epoch 3/100
accuracy: 0.8943 - val_loss: 0.0737 - val_accuracy: 0.8954
Epoch 4/100
accuracy: 0.8935 - val_loss: 0.0731 - val_accuracy: 0.8969
Epoch 5/100
accuracy: 0.8942 - val_loss: 0.0729 - val_accuracy: 0.9000
Epoch 6/100
accuracy: 0.8946 - val_loss: 0.0744 - val_accuracy: 0.8946
Epoch 7/100
accuracy: 0.8942 - val_loss: 0.0728 - val_accuracy: 0.8989
Epoch 8/100
accuracy: 0.8952 - val_loss: 0.0727 - val_accuracy: 0.8983
Epoch 9/100
accuracy: 0.8932 - val loss: 0.0738 - val accuracy: 0.8969
Epoch 10/100
accuracy: 0.8954 - val_loss: 0.0733 - val_accuracy: 0.8983
Epoch 11/100
accuracy: 0.8947 - val_loss: 0.0732 - val_accuracy: 0.8986
Epoch 12/100
accuracy: 0.8946 - val_loss: 0.0736 - val_accuracy: 0.8959
Epoch 13/100
```

```
accuracy: 0.8946 - val_loss: 0.0735 - val_accuracy: 0.8968
Epoch 14/100
accuracy: 0.8937 - val_loss: 0.0736 - val_accuracy: 0.8969
Epoch 15/100
accuracy: 0.8939 - val_loss: 0.0732 - val_accuracy: 0.8968
Epoch 16/100
accuracy: 0.8943 - val_loss: 0.0737 - val_accuracy: 0.8977
Epoch 17/100
accuracy: 0.8952 - val_loss: 0.0730 - val_accuracy: 0.8979
Epoch 18/100
accuracy: 0.8951 - val_loss: 0.0731 - val_accuracy: 0.8968
Epoch 19/100
accuracy: 0.8932 - val_loss: 0.0735 - val_accuracy: 0.8964
Epoch 20/100
accuracy: 0.8933 - val_loss: 0.0731 - val_accuracy: 0.8983
Epoch 21/100
accuracy: 0.8933 - val_loss: 0.0740 - val_accuracy: 0.8952
Epoch 22/100
accuracy: 0.8923 - val_loss: 0.0735 - val_accuracy: 0.8971
Epoch 23/100
accuracy: 0.8936 - val_loss: 0.0726 - val_accuracy: 0.8998
Epoch 24/100
accuracy: 0.8935 - val_loss: 0.0732 - val_accuracy: 0.8985
Epoch 25/100
accuracy: 0.8947 - val_loss: 0.0727 - val_accuracy: 0.8971
Epoch 26/100
accuracy: 0.8940 - val_loss: 0.0720 - val_accuracy: 0.9019
Epoch 27/100
accuracy: 0.8932 - val_loss: 0.0731 - val_accuracy: 0.8992
Epoch 28/100
1349/1349 [============= ] - 1s 1ms/step - loss: 0.0768 -
accuracy: 0.8936 - val_loss: 0.0733 - val_accuracy: 0.8981
Epoch 29/100
```

```
accuracy: 0.8936 - val_loss: 0.0726 - val_accuracy: 0.8985
Epoch 30/100
accuracy: 0.8941 - val_loss: 0.0724 - val_accuracy: 0.8994
Epoch 31/100
accuracy: 0.8941 - val_loss: 0.0727 - val_accuracy: 0.9004
Epoch 32/100
accuracy: 0.8951 - val_loss: 0.0732 - val_accuracy: 0.8986
Epoch 33/100
accuracy: 0.8942 - val_loss: 0.0730 - val_accuracy: 0.8985
Epoch 34/100
accuracy: 0.8940 - val_loss: 0.0729 - val_accuracy: 0.8994
Epoch 35/100
accuracy: 0.8933 - val_loss: 0.0729 - val_accuracy: 0.8982
Epoch 36/100
accuracy: 0.8940 - val_loss: 0.0729 - val_accuracy: 0.8979
Epoch 37/100
accuracy: 0.8940 - val_loss: 0.0735 - val_accuracy: 0.8977
Epoch 38/100
accuracy: 0.8941 - val_loss: 0.0726 - val_accuracy: 0.8972
Epoch 39/100
accuracy: 0.8935 - val_loss: 0.0728 - val_accuracy: 0.8980
Epoch 40/100
accuracy: 0.8931 - val_loss: 0.0740 - val_accuracy: 0.8951
Epoch 41/100
accuracy: 0.8939 - val loss: 0.0726 - val accuracy: 0.8975
Epoch 42/100
accuracy: 0.8936 - val_loss: 0.0731 - val_accuracy: 0.8967
Epoch 43/100
accuracy: 0.8926 - val_loss: 0.0731 - val_accuracy: 0.8987
Epoch 44/100
accuracy: 0.8939 - val_loss: 0.0727 - val_accuracy: 0.8998
Epoch 45/100
```

```
accuracy: 0.8945 - val_loss: 0.0730 - val_accuracy: 0.8989
Epoch 46/100
accuracy: 0.8938 - val_loss: 0.0732 - val_accuracy: 0.8981
Epoch 47/100
accuracy: 0.8939 - val_loss: 0.0729 - val_accuracy: 0.8972
Epoch 48/100
accuracy: 0.8944 - val_loss: 0.0739 - val_accuracy: 0.8950
Epoch 49/100
accuracy: 0.8940 - val_loss: 0.0731 - val_accuracy: 0.8988
Epoch 50/100
accuracy: 0.8952 - val_loss: 0.0725 - val_accuracy: 0.8999
Epoch 51/100
accuracy: 0.8940 - val_loss: 0.0725 - val_accuracy: 0.8986
Epoch 52/100
accuracy: 0.8942 - val_loss: 0.0733 - val_accuracy: 0.8978
Epoch 53/100
accuracy: 0.8939 - val_loss: 0.0728 - val_accuracy: 0.8977
Epoch 54/100
accuracy: 0.8938 - val_loss: 0.0724 - val_accuracy: 0.9000
accuracy: 0.8936 - val_loss: 0.0731 - val_accuracy: 0.8972
Epoch 56/100
accuracy: 0.8930 - val_loss: 0.0731 - val_accuracy: 0.8985
Epoch 57/100
accuracy: 0.8940 - val loss: 0.0734 - val accuracy: 0.8947
Epoch 58/100
accuracy: 0.8935 - val_loss: 0.0735 - val_accuracy: 0.8979
Epoch 59/100
accuracy: 0.8944 - val_loss: 0.0730 - val_accuracy: 0.8972
Epoch 60/100
1349/1349 [============= ] - 1s 1ms/step - loss: 0.0766 -
accuracy: 0.8949 - val_loss: 0.0729 - val_accuracy: 0.8996
Epoch 61/100
```

```
accuracy: 0.8942 - val_loss: 0.0733 - val_accuracy: 0.8988
Epoch 62/100
accuracy: 0.8948 - val_loss: 0.0725 - val_accuracy: 0.9001
Epoch 63/100
accuracy: 0.8951 - val_loss: 0.0726 - val_accuracy: 0.8992
Epoch 64/100
accuracy: 0.8947 - val_loss: 0.0727 - val_accuracy: 0.8980
Epoch 65/100
accuracy: 0.8930 - val_loss: 0.0728 - val_accuracy: 0.8980
Epoch 66/100
accuracy: 0.8931 - val_loss: 0.0728 - val_accuracy: 0.8999
Epoch 67/100
accuracy: 0.8955 - val_loss: 0.0728 - val_accuracy: 0.8977
Epoch 68/100
accuracy: 0.8950 - val_loss: 0.0736 - val_accuracy: 0.8981
Epoch 69/100
accuracy: 0.8948 - val_loss: 0.0733 - val_accuracy: 0.8987
Epoch 70/100
accuracy: 0.8945 - val_loss: 0.0729 - val_accuracy: 0.8997
Epoch 71/100
accuracy: 0.8939 - val_loss: 0.0727 - val_accuracy: 0.8982
Epoch 72/100
accuracy: 0.8939 - val_loss: 0.0734 - val_accuracy: 0.8994
Epoch 73/100
accuracy: 0.8955 - val loss: 0.0731 - val accuracy: 0.8986
Epoch 74/100
accuracy: 0.8945 - val_loss: 0.0727 - val_accuracy: 0.9009
Epoch 75/100
accuracy: 0.8946 - val_loss: 0.0728 - val_accuracy: 0.8972
Epoch 76/100
accuracy: 0.8947 - val_loss: 0.0730 - val_accuracy: 0.8996
Epoch 77/100
```

```
accuracy: 0.8945 - val_loss: 0.0732 - val_accuracy: 0.8989
Epoch 78/100
accuracy: 0.8942 - val_loss: 0.0726 - val_accuracy: 0.8991
Epoch 79/100
accuracy: 0.8940 - val_loss: 0.0725 - val_accuracy: 0.8998
Epoch 80/100
accuracy: 0.8943 - val_loss: 0.0725 - val_accuracy: 0.9000
Epoch 81/100
accuracy: 0.8936 - val_loss: 0.0732 - val_accuracy: 0.8980
Epoch 82/100
accuracy: 0.8937 - val_loss: 0.0729 - val_accuracy: 0.8986
Epoch 83/100
accuracy: 0.8949 - val_loss: 0.0726 - val_accuracy: 0.8967
Epoch 84/100
accuracy: 0.8941 - val_loss: 0.0728 - val_accuracy: 0.8994
Epoch 85/100
accuracy: 0.8941 - val_loss: 0.0732 - val_accuracy: 0.8978
Epoch 86/100
accuracy: 0.8942 - val_loss: 0.0734 - val_accuracy: 0.8986
Epoch 87/100
accuracy: 0.8942 - val_loss: 0.0726 - val_accuracy: 0.8994
Epoch 88/100
accuracy: 0.8939 - val_loss: 0.0726 - val_accuracy: 0.8992
Epoch 89/100
accuracy: 0.8942 - val loss: 0.0725 - val accuracy: 0.8995
Epoch 90/100
accuracy: 0.8927 - val_loss: 0.0730 - val_accuracy: 0.8980
Epoch 91/100
accuracy: 0.8942 - val_loss: 0.0729 - val_accuracy: 0.8985
Epoch 92/100
accuracy: 0.8954 - val_loss: 0.0734 - val_accuracy: 0.8980
Epoch 93/100
```

```
accuracy: 0.8942 - val_loss: 0.0726 - val_accuracy: 0.8988
Epoch 94/100
accuracy: 0.8945 - val_loss: 0.0733 - val_accuracy: 0.8974
Epoch 95/100
accuracy: 0.8947 - val_loss: 0.0724 - val_accuracy: 0.9000
Epoch 96/100
accuracy: 0.8944 - val_loss: 0.0728 - val_accuracy: 0.8984
Epoch 97/100
accuracy: 0.8946 - val_loss: 0.0732 - val_accuracy: 0.8981
Epoch 98/100
accuracy: 0.8945 - val_loss: 0.0734 - val_accuracy: 0.8982
Epoch 99/100
accuracy: 0.8945 - val_loss: 0.0735 - val_accuracy: 0.8995
Epoch 100/100
accuracy: 0.8945 - val_loss: 0.0724 - val_accuracy: 0.8998
Epoch 1/100
accuracy: 0.8940 - val_loss: 0.0739 - val_accuracy: 0.8984
Epoch 2/100
accuracy: 0.8943 - val_loss: 0.0729 - val_accuracy: 0.8990
accuracy: 0.8943 - val_loss: 0.0726 - val_accuracy: 0.8995
Epoch 4/100
accuracy: 0.8941 - val_loss: 0.0730 - val_accuracy: 0.8978
Epoch 5/100
accuracy: 0.8943 - val loss: 0.0723 - val accuracy: 0.9007
Epoch 6/100
accuracy: 0.8945 - val_loss: 0.0729 - val_accuracy: 0.8986
Epoch 7/100
accuracy: 0.8939 - val_loss: 0.0726 - val_accuracy: 0.8996
Epoch 8/100
accuracy: 0.8944 - val_loss: 0.0728 - val_accuracy: 0.8978
Epoch 9/100
```

```
accuracy: 0.8945 - val_loss: 0.0727 - val_accuracy: 0.8992
Epoch 10/100
accuracy: 0.8953 - val_loss: 0.0738 - val_accuracy: 0.8958
Epoch 11/100
accuracy: 0.8948 - val_loss: 0.0726 - val_accuracy: 0.8995
Epoch 12/100
accuracy: 0.8948 - val_loss: 0.0727 - val_accuracy: 0.8988
Epoch 13/100
accuracy: 0.8951 - val_loss: 0.0733 - val_accuracy: 0.8995
Epoch 14/100
accuracy: 0.8929 - val_loss: 0.0726 - val_accuracy: 0.8987
Epoch 15/100
accuracy: 0.8950 - val_loss: 0.0726 - val_accuracy: 0.8986
Epoch 16/100
accuracy: 0.8943 - val_loss: 0.0730 - val_accuracy: 0.8986
Epoch 17/100
accuracy: 0.8952 - val_loss: 0.0728 - val_accuracy: 0.8973
Epoch 18/100
accuracy: 0.8943 - val_loss: 0.0724 - val_accuracy: 0.8985
Epoch 19/100
accuracy: 0.8941 - val_loss: 0.0725 - val_accuracy: 0.8987
Epoch 20/100
accuracy: 0.8938 - val_loss: 0.0732 - val_accuracy: 0.8967
Epoch 21/100
accuracy: 0.8931 - val_loss: 0.0726 - val_accuracy: 0.8982
Epoch 22/100
accuracy: 0.8959 - val_loss: 0.0731 - val_accuracy: 0.8979
Epoch 23/100
accuracy: 0.8940 - val_loss: 0.0730 - val_accuracy: 0.8991
Epoch 24/100
accuracy: 0.8947 - val_loss: 0.0748 - val_accuracy: 0.8936
Epoch 25/100
```

```
accuracy: 0.8945 - val_loss: 0.0733 - val_accuracy: 0.8985
Epoch 26/100
accuracy: 0.8948 - val_loss: 0.0731 - val_accuracy: 0.8968
Epoch 27/100
accuracy: 0.8946 - val_loss: 0.0728 - val_accuracy: 0.8982
Epoch 28/100
accuracy: 0.8945 - val_loss: 0.0726 - val_accuracy: 0.8992
Epoch 29/100
accuracy: 0.8947 - val_loss: 0.0723 - val_accuracy: 0.9000
Epoch 30/100
accuracy: 0.8946 - val_loss: 0.0731 - val_accuracy: 0.8985
Epoch 31/100
accuracy: 0.8945 - val_loss: 0.0732 - val_accuracy: 0.8976
Epoch 32/100
accuracy: 0.8957 - val_loss: 0.0734 - val_accuracy: 0.8975
Epoch 33/100
accuracy: 0.8953 - val_loss: 0.0730 - val_accuracy: 0.8995
Epoch 34/100
accuracy: 0.8943 - val_loss: 0.0724 - val_accuracy: 0.8988
accuracy: 0.8953 - val_loss: 0.0726 - val_accuracy: 0.8989
Epoch 36/100
accuracy: 0.8942 - val_loss: 0.0728 - val_accuracy: 0.8962
Epoch 37/100
accuracy: 0.8939 - val loss: 0.0725 - val accuracy: 0.9007
Epoch 38/100
accuracy: 0.8954 - val_loss: 0.0727 - val_accuracy: 0.8983
Epoch 39/100
accuracy: 0.8952 - val_loss: 0.0722 - val_accuracy: 0.8988
Epoch 40/100
accuracy: 0.8949 - val_loss: 0.0723 - val_accuracy: 0.8987
Epoch 41/100
```

```
accuracy: 0.8949 - val_loss: 0.0732 - val_accuracy: 0.8995
Epoch 42/100
accuracy: 0.8953 - val_loss: 0.0724 - val_accuracy: 0.9001
Epoch 43/100
accuracy: 0.8954 - val_loss: 0.0728 - val_accuracy: 0.8994
Epoch 44/100
accuracy: 0.8950 - val_loss: 0.0727 - val_accuracy: 0.8967
Epoch 45/100
accuracy: 0.8937 - val_loss: 0.0723 - val_accuracy: 0.8996
Epoch 46/100
accuracy: 0.8941 - val_loss: 0.0735 - val_accuracy: 0.8983
Epoch 47/100
accuracy: 0.8946 - val_loss: 0.0723 - val_accuracy: 0.9005
Epoch 48/100
accuracy: 0.8942 - val_loss: 0.0735 - val_accuracy: 0.8969
Epoch 49/100
accuracy: 0.8948 - val_loss: 0.0730 - val_accuracy: 0.8994
Epoch 50/100
accuracy: 0.8941 - val_loss: 0.0739 - val_accuracy: 0.8992
Epoch 51/100
accuracy: 0.8951 - val_loss: 0.0733 - val_accuracy: 0.8967
Epoch 52/100
accuracy: 0.8953 - val_loss: 0.0729 - val_accuracy: 0.8990
Epoch 53/100
accuracy: 0.8951 - val loss: 0.0730 - val accuracy: 0.8990
Epoch 54/100
accuracy: 0.8943 - val_loss: 0.0728 - val_accuracy: 0.8994
Epoch 55/100
accuracy: 0.8945 - val_loss: 0.0732 - val_accuracy: 0.8982
Epoch 56/100
accuracy: 0.8941 - val_loss: 0.0728 - val_accuracy: 0.8986
Epoch 57/100
```

```
accuracy: 0.8938 - val_loss: 0.0731 - val_accuracy: 0.8968
Epoch 58/100
accuracy: 0.8955 - val_loss: 0.0732 - val_accuracy: 0.8960
Epoch 59/100
accuracy: 0.8939 - val_loss: 0.0733 - val_accuracy: 0.8969
Epoch 60/100
accuracy: 0.8948 - val_loss: 0.0729 - val_accuracy: 0.8967
Epoch 61/100
accuracy: 0.8934 - val_loss: 0.0725 - val_accuracy: 0.8991
Epoch 62/100
accuracy: 0.8936 - val_loss: 0.0725 - val_accuracy: 0.8993
Epoch 63/100
accuracy: 0.8932 - val_loss: 0.0727 - val_accuracy: 0.8993
Epoch 64/100
accuracy: 0.8941 - val_loss: 0.0723 - val_accuracy: 0.9006
Epoch 65/100
accuracy: 0.8953 - val_loss: 0.0726 - val_accuracy: 0.8989
Epoch 66/100
accuracy: 0.8946 - val_loss: 0.0726 - val_accuracy: 0.8997
Epoch 67/100
accuracy: 0.8940 - val_loss: 0.0733 - val_accuracy: 0.8992
Epoch 68/100
accuracy: 0.8951 - val_loss: 0.0749 - val_accuracy: 0.8938
Epoch 69/100
accuracy: 0.8944 - val loss: 0.0732 - val accuracy: 0.8970
Epoch 70/100
accuracy: 0.8945 - val_loss: 0.0726 - val_accuracy: 0.8994
Epoch 71/100
accuracy: 0.8941 - val_loss: 0.0732 - val_accuracy: 0.9001
Epoch 72/100
accuracy: 0.8953 - val_loss: 0.0728 - val_accuracy: 0.8987
Epoch 73/100
```

```
accuracy: 0.8952 - val_loss: 0.0725 - val_accuracy: 0.8971
Epoch 74/100
accuracy: 0.8943 - val_loss: 0.0729 - val_accuracy: 0.8979
Epoch 75/100
accuracy: 0.8948 - val_loss: 0.0730 - val_accuracy: 0.8990
Epoch 76/100
accuracy: 0.8945 - val_loss: 0.0728 - val_accuracy: 0.8985
Epoch 77/100
accuracy: 0.8936 - val_loss: 0.0723 - val_accuracy: 0.9000
Epoch 78/100
accuracy: 0.8936 - val_loss: 0.0732 - val_accuracy: 0.8973
Epoch 79/100
accuracy: 0.8955 - val_loss: 0.0724 - val_accuracy: 0.8986
Epoch 80/100
accuracy: 0.8949 - val_loss: 0.0727 - val_accuracy: 0.8991
Epoch 81/100
accuracy: 0.8947 - val_loss: 0.0725 - val_accuracy: 0.8991
Epoch 82/100
accuracy: 0.8945 - val_loss: 0.0727 - val_accuracy: 0.8990
Epoch 83/100
accuracy: 0.8948 - val_loss: 0.0728 - val_accuracy: 0.8996
Epoch 84/100
accuracy: 0.8959 - val_loss: 0.0732 - val_accuracy: 0.8971
Epoch 85/100
accuracy: 0.8946 - val loss: 0.0726 - val accuracy: 0.8991
Epoch 86/100
accuracy: 0.8955 - val_loss: 0.0728 - val_accuracy: 0.8982
Epoch 87/100
accuracy: 0.8941 - val_loss: 0.0733 - val_accuracy: 0.8975
Epoch 88/100
accuracy: 0.8949 - val_loss: 0.0732 - val_accuracy: 0.8989
Epoch 89/100
```

```
accuracy: 0.8945 - val_loss: 0.0727 - val_accuracy: 0.9004
Epoch 90/100
accuracy: 0.8956 - val_loss: 0.0727 - val_accuracy: 0.8979
Epoch 91/100
accuracy: 0.8943 - val_loss: 0.0738 - val_accuracy: 0.8989
Epoch 92/100
accuracy: 0.8940 - val_loss: 0.0741 - val_accuracy: 0.8996
Epoch 93/100
accuracy: 0.8945 - val_loss: 0.0735 - val_accuracy: 0.8964
Epoch 94/100
accuracy: 0.8938 - val_loss: 0.0731 - val_accuracy: 0.8995
Epoch 95/100
accuracy: 0.8952 - val_loss: 0.0731 - val_accuracy: 0.8973
Epoch 96/100
accuracy: 0.8943 - val_loss: 0.0728 - val_accuracy: 0.8994
Epoch 97/100
accuracy: 0.8947 - val_loss: 0.0724 - val_accuracy: 0.8982
Epoch 98/100
accuracy: 0.8958 - val_loss: 0.0728 - val_accuracy: 0.8988
Epoch 99/100
accuracy: 0.8950 - val_loss: 0.0725 - val_accuracy: 0.8990
Epoch 100/100
accuracy: 0.8948 - val_loss: 0.0730 - val_accuracy: 0.8996
Epoch 1/100
accuracy: 0.8936 - val loss: 0.0736 - val accuracy: 0.8961
Epoch 2/100
accuracy: 0.8947 - val_loss: 0.0732 - val_accuracy: 0.8987
Epoch 3/100
accuracy: 0.8946 - val_loss: 0.0725 - val_accuracy: 0.9006
Epoch 4/100
accuracy: 0.8946 - val_loss: 0.0733 - val_accuracy: 0.8994
Epoch 5/100
```

```
accuracy: 0.8945 - val_loss: 0.0726 - val_accuracy: 0.8991
Epoch 6/100
accuracy: 0.8944 - val_loss: 0.0727 - val_accuracy: 0.8986
Epoch 7/100
accuracy: 0.8953 - val_loss: 0.0728 - val_accuracy: 0.8984
Epoch 8/100
accuracy: 0.8952 - val_loss: 0.0724 - val_accuracy: 0.8994
Epoch 9/100
accuracy: 0.8962 - val_loss: 0.0735 - val_accuracy: 0.8983
Epoch 10/100
accuracy: 0.8952 - val_loss: 0.0727 - val_accuracy: 0.8989
Epoch 11/100
accuracy: 0.8938 - val_loss: 0.0733 - val_accuracy: 0.8965
Epoch 12/100
accuracy: 0.8963 - val_loss: 0.0728 - val_accuracy: 0.8984
Epoch 13/100
accuracy: 0.8949 - val_loss: 0.0729 - val_accuracy: 0.8982
Epoch 14/100
accuracy: 0.8946 - val_loss: 0.0729 - val_accuracy: 0.8985
accuracy: 0.8939 - val_loss: 0.0739 - val_accuracy: 0.8950
Epoch 16/100
accuracy: 0.8948 - val_loss: 0.0726 - val_accuracy: 0.8991
Epoch 17/100
accuracy: 0.8949 - val loss: 0.0732 - val accuracy: 0.8987
Epoch 18/100
accuracy: 0.8949 - val_loss: 0.0729 - val_accuracy: 0.8983
Epoch 19/100
accuracy: 0.8943 - val_loss: 0.0732 - val_accuracy: 0.8983
Epoch 20/100
accuracy: 0.8953 - val_loss: 0.0725 - val_accuracy: 0.8992
Epoch 21/100
```

```
accuracy: 0.8947 - val_loss: 0.0731 - val_accuracy: 0.8977
Epoch 22/100
accuracy: 0.8935 - val_loss: 0.0729 - val_accuracy: 0.9000
Epoch 23/100
accuracy: 0.8961 - val_loss: 0.0730 - val_accuracy: 0.8979
Epoch 24/100
accuracy: 0.8954 - val_loss: 0.0728 - val_accuracy: 0.8990
Epoch 25/100
accuracy: 0.8939 - val_loss: 0.0728 - val_accuracy: 0.8982
Epoch 26/100
accuracy: 0.8945 - val_loss: 0.0732 - val_accuracy: 0.8988
Epoch 27/100
accuracy: 0.8948 - val_loss: 0.0726 - val_accuracy: 0.8995
Epoch 28/100
accuracy: 0.8941 - val_loss: 0.0729 - val_accuracy: 0.8982
Epoch 29/100
accuracy: 0.8943 - val_loss: 0.0728 - val_accuracy: 0.8979
Epoch 30/100
accuracy: 0.8949 - val_loss: 0.0727 - val_accuracy: 0.8985
Epoch 31/100
accuracy: 0.8945 - val_loss: 0.0731 - val_accuracy: 0.8987
Epoch 32/100
accuracy: 0.8949 - val_loss: 0.0731 - val_accuracy: 0.8979
Epoch 33/100
accuracy: 0.8960 - val loss: 0.0730 - val accuracy: 0.8991
Epoch 34/100
accuracy: 0.8962 - val_loss: 0.0727 - val_accuracy: 0.9002
Epoch 35/100
accuracy: 0.8944 - val_loss: 0.0732 - val_accuracy: 0.8991
Epoch 36/100
accuracy: 0.8961 - val_loss: 0.0726 - val_accuracy: 0.8995
Epoch 37/100
```

```
accuracy: 0.8949 - val_loss: 0.0728 - val_accuracy: 0.8992
Epoch 38/100
accuracy: 0.8960 - val_loss: 0.0725 - val_accuracy: 0.8987
Epoch 39/100
accuracy: 0.8953 - val_loss: 0.0733 - val_accuracy: 0.8970
Epoch 40/100
accuracy: 0.8941 - val_loss: 0.0729 - val_accuracy: 0.9010
Epoch 41/100
accuracy: 0.8948 - val_loss: 0.0730 - val_accuracy: 0.9011
Epoch 42/100
accuracy: 0.8956 - val_loss: 0.0737 - val_accuracy: 0.8950
Epoch 43/100
accuracy: 0.8943 - val_loss: 0.0727 - val_accuracy: 0.9004
Epoch 44/100
accuracy: 0.8951 - val_loss: 0.0720 - val_accuracy: 0.9007
Epoch 45/100
accuracy: 0.8948 - val_loss: 0.0727 - val_accuracy: 0.8980
Epoch 46/100
accuracy: 0.8951 - val_loss: 0.0728 - val_accuracy: 0.8991
Epoch 47/100
accuracy: 0.8949 - val_loss: 0.0727 - val_accuracy: 0.8992
Epoch 48/100
accuracy: 0.8944 - val_loss: 0.0728 - val_accuracy: 0.8987
Epoch 49/100
accuracy: 0.8951 - val loss: 0.0730 - val accuracy: 0.8988
Epoch 50/100
accuracy: 0.8943 - val_loss: 0.0725 - val_accuracy: 0.8989
Epoch 51/100
accuracy: 0.8953 - val_loss: 0.0739 - val_accuracy: 0.8977
Epoch 52/100
accuracy: 0.8949 - val_loss: 0.0735 - val_accuracy: 0.8987
Epoch 53/100
```

```
accuracy: 0.8953 - val_loss: 0.0732 - val_accuracy: 0.8988
Epoch 54/100
accuracy: 0.8946 - val_loss: 0.0730 - val_accuracy: 0.8989
Epoch 55/100
accuracy: 0.8958 - val_loss: 0.0724 - val_accuracy: 0.9003
Epoch 56/100
accuracy: 0.8953 - val_loss: 0.0720 - val_accuracy: 0.8995
Epoch 57/100
accuracy: 0.8948 - val_loss: 0.0726 - val_accuracy: 0.8992
Epoch 58/100
accuracy: 0.8956 - val_loss: 0.0727 - val_accuracy: 0.8987
Epoch 59/100
accuracy: 0.8945 - val_loss: 0.0726 - val_accuracy: 0.8993
Epoch 60/100
accuracy: 0.8948 - val_loss: 0.0723 - val_accuracy: 0.9001
Epoch 61/100
accuracy: 0.8961 - val_loss: 0.0730 - val_accuracy: 0.8978
Epoch 62/100
accuracy: 0.8955 - val_loss: 0.0730 - val_accuracy: 0.8986
Epoch 63/100
accuracy: 0.8938 - val_loss: 0.0727 - val_accuracy: 0.8993
Epoch 64/100
accuracy: 0.8946 - val_loss: 0.0726 - val_accuracy: 0.8983
Epoch 65/100
accuracy: 0.8947 - val_loss: 0.0729 - val_accuracy: 0.8984
Epoch 66/100
accuracy: 0.8949 - val_loss: 0.0731 - val_accuracy: 0.8986
Epoch 67/100
accuracy: 0.8949 - val_loss: 0.0728 - val_accuracy: 0.8992
Epoch 68/100
accuracy: 0.8963 - val_loss: 0.0731 - val_accuracy: 0.8975
Epoch 69/100
```

```
accuracy: 0.8951 - val_loss: 0.0725 - val_accuracy: 0.8990
Epoch 70/100
accuracy: 0.8948 - val_loss: 0.0725 - val_accuracy: 0.8998
Epoch 71/100
accuracy: 0.8961 - val_loss: 0.0729 - val_accuracy: 0.9009
Epoch 72/100
accuracy: 0.8951 - val_loss: 0.0733 - val_accuracy: 0.8975
Epoch 73/100
accuracy: 0.8956 - val_loss: 0.0729 - val_accuracy: 0.8985
Epoch 74/100
accuracy: 0.8949 - val_loss: 0.0731 - val_accuracy: 0.8991
Epoch 75/100
accuracy: 0.8951 - val_loss: 0.0730 - val_accuracy: 0.8986
Epoch 76/100
accuracy: 0.8954 - val_loss: 0.0724 - val_accuracy: 0.8993
Epoch 77/100
accuracy: 0.8959 - val_loss: 0.0731 - val_accuracy: 0.8984
Epoch 78/100
accuracy: 0.8957 - val_loss: 0.0727 - val_accuracy: 0.8973
Epoch 79/100
accuracy: 0.8953 - val_loss: 0.0727 - val_accuracy: 0.8984
Epoch 80/100
accuracy: 0.8949 - val_loss: 0.0730 - val_accuracy: 0.8987
Epoch 81/100
accuracy: 0.8949 - val loss: 0.0731 - val accuracy: 0.8987
Epoch 82/100
accuracy: 0.8945 - val_loss: 0.0730 - val_accuracy: 0.8974
Epoch 83/100
accuracy: 0.8952 - val_loss: 0.0726 - val_accuracy: 0.8996
Epoch 84/100
accuracy: 0.8957 - val_loss: 0.0732 - val_accuracy: 0.8984
Epoch 85/100
```

```
accuracy: 0.8951 - val_loss: 0.0724 - val_accuracy: 0.9003
Epoch 86/100
accuracy: 0.8951 - val_loss: 0.0723 - val_accuracy: 0.9003
Epoch 87/100
accuracy: 0.8958 - val_loss: 0.0735 - val_accuracy: 0.8964
Epoch 88/100
accuracy: 0.8939 - val_loss: 0.0731 - val_accuracy: 0.9000
Epoch 89/100
accuracy: 0.8952 - val_loss: 0.0725 - val_accuracy: 0.9002
Epoch 90/100
accuracy: 0.8954 - val_loss: 0.0733 - val_accuracy: 0.8965
Epoch 91/100
accuracy: 0.8959 - val_loss: 0.0725 - val_accuracy: 0.9004
Epoch 92/100
accuracy: 0.8956 - val_loss: 0.0730 - val_accuracy: 0.8975
Epoch 93/100
accuracy: 0.8946 - val_loss: 0.0726 - val_accuracy: 0.9000
Epoch 94/100
accuracy: 0.8961 - val_loss: 0.0724 - val_accuracy: 0.8992
accuracy: 0.8946 - val_loss: 0.0742 - val_accuracy: 0.8964
Epoch 96/100
accuracy: 0.8957 - val_loss: 0.0726 - val_accuracy: 0.8991
Epoch 97/100
accuracy: 0.8957 - val_loss: 0.0730 - val_accuracy: 0.8984
Epoch 98/100
accuracy: 0.8951 - val_loss: 0.0730 - val_accuracy: 0.8976
Epoch 99/100
accuracy: 0.8950 - val_loss: 0.0728 - val_accuracy: 0.8992
Epoch 100/100
accuracy: 0.8941 - val_loss: 0.0731 - val_accuracy: 0.8998
Epoch 1/100
```

```
accuracy: 0.8951 - val_loss: 0.0738 - val_accuracy: 0.8972
Epoch 2/100
accuracy: 0.8931 - val_loss: 0.0725 - val_accuracy: 0.9002
Epoch 3/100
accuracy: 0.8948 - val_loss: 0.0728 - val_accuracy: 0.8980
Epoch 4/100
accuracy: 0.8957 - val_loss: 0.0726 - val_accuracy: 0.8994
Epoch 5/100
accuracy: 0.8944 - val_loss: 0.0728 - val_accuracy: 0.9003
Epoch 6/100
accuracy: 0.8958 - val_loss: 0.0729 - val_accuracy: 0.8989
Epoch 7/100
accuracy: 0.8956 - val_loss: 0.0728 - val_accuracy: 0.8985
Epoch 8/100
accuracy: 0.8956 - val_loss: 0.0736 - val_accuracy: 0.8975
Epoch 9/100
accuracy: 0.8954 - val_loss: 0.0739 - val_accuracy: 0.8986
Epoch 10/100
accuracy: 0.8959 - val_loss: 0.0729 - val_accuracy: 0.8974
accuracy: 0.8949 - val_loss: 0.0730 - val_accuracy: 0.8984
Epoch 12/100
accuracy: 0.8936 - val_loss: 0.0734 - val_accuracy: 0.8997
Epoch 13/100
accuracy: 0.8938 - val_loss: 0.0737 - val_accuracy: 0.8962
Epoch 14/100
accuracy: 0.8954 - val_loss: 0.0727 - val_accuracy: 0.8982
Epoch 15/100
accuracy: 0.8954 - val_loss: 0.0725 - val_accuracy: 0.8989
Epoch 16/100
accuracy: 0.8951 - val_loss: 0.0727 - val_accuracy: 0.9000
Epoch 17/100
```

```
accuracy: 0.8965 - val_loss: 0.0731 - val_accuracy: 0.8986
Epoch 18/100
accuracy: 0.8956 - val_loss: 0.0723 - val_accuracy: 0.8994
Epoch 19/100
accuracy: 0.8953 - val_loss: 0.0731 - val_accuracy: 0.8962
Epoch 20/100
accuracy: 0.8950 - val_loss: 0.0725 - val_accuracy: 0.9001
Epoch 21/100
accuracy: 0.8943 - val_loss: 0.0730 - val_accuracy: 0.8985
Epoch 22/100
accuracy: 0.8964 - val_loss: 0.0723 - val_accuracy: 0.9002
Epoch 23/100
accuracy: 0.8950 - val_loss: 0.0725 - val_accuracy: 0.8992
Epoch 24/100
accuracy: 0.8939 - val_loss: 0.0724 - val_accuracy: 0.8998
Epoch 25/100
accuracy: 0.8956 - val_loss: 0.0725 - val_accuracy: 0.8996
Epoch 26/100
accuracy: 0.8947 - val_loss: 0.0724 - val_accuracy: 0.9003
Epoch 27/100
accuracy: 0.8945 - val_loss: 0.0732 - val_accuracy: 0.8997
Epoch 28/100
accuracy: 0.8965 - val_loss: 0.0726 - val_accuracy: 0.8994
Epoch 29/100
accuracy: 0.8956 - val loss: 0.0733 - val accuracy: 0.8980
Epoch 30/100
accuracy: 0.8961 - val_loss: 0.0728 - val_accuracy: 0.9008
Epoch 31/100
accuracy: 0.8953 - val_loss: 0.0727 - val_accuracy: 0.8992
Epoch 32/100
accuracy: 0.8949 - val_loss: 0.0729 - val_accuracy: 0.8994
Epoch 33/100
```

```
accuracy: 0.8948 - val_loss: 0.0724 - val_accuracy: 0.9001
Epoch 34/100
accuracy: 0.8951 - val_loss: 0.0729 - val_accuracy: 0.9005
Epoch 35/100
accuracy: 0.8951 - val_loss: 0.0728 - val_accuracy: 0.9001
Epoch 36/100
accuracy: 0.8961 - val_loss: 0.0726 - val_accuracy: 0.8994
Epoch 37/100
accuracy: 0.8963 - val_loss: 0.0726 - val_accuracy: 0.8999
Epoch 38/100
accuracy: 0.8941 - val_loss: 0.0739 - val_accuracy: 0.8948
Epoch 39/100
accuracy: 0.8957 - val_loss: 0.0725 - val_accuracy: 0.8982
Epoch 40/100
accuracy: 0.8948 - val_loss: 0.0731 - val_accuracy: 0.8982
Epoch 41/100
accuracy: 0.8944 - val_loss: 0.0726 - val_accuracy: 0.8992
Epoch 42/100
accuracy: 0.8947 - val_loss: 0.0724 - val_accuracy: 0.9001
Epoch 43/100
accuracy: 0.8958 - val_loss: 0.0725 - val_accuracy: 0.9009
Epoch 44/100
accuracy: 0.8963 - val_loss: 0.0731 - val_accuracy: 0.8991
Epoch 45/100
accuracy: 0.8945 - val loss: 0.0725 - val accuracy: 0.9006
Epoch 46/100
accuracy: 0.8963 - val_loss: 0.0729 - val_accuracy: 0.8984
Epoch 47/100
accuracy: 0.8948 - val_loss: 0.0727 - val_accuracy: 0.8991
Epoch 48/100
accuracy: 0.8963 - val_loss: 0.0725 - val_accuracy: 0.8984
Epoch 49/100
```

```
accuracy: 0.8948 - val_loss: 0.0730 - val_accuracy: 0.8983
Epoch 50/100
accuracy: 0.8949 - val_loss: 0.0734 - val_accuracy: 0.8978
Epoch 51/100
accuracy: 0.8946 - val_loss: 0.0728 - val_accuracy: 0.8995
Epoch 52/100
accuracy: 0.8945 - val_loss: 0.0733 - val_accuracy: 0.8975
Epoch 53/100
accuracy: 0.8960 - val_loss: 0.0729 - val_accuracy: 0.8988
Epoch 54/100
accuracy: 0.8950 - val_loss: 0.0729 - val_accuracy: 0.9002
Epoch 55/100
accuracy: 0.8959 - val_loss: 0.0729 - val_accuracy: 0.8988
Epoch 56/100
accuracy: 0.8953 - val_loss: 0.0726 - val_accuracy: 0.8994
Epoch 57/100
accuracy: 0.8964 - val_loss: 0.0732 - val_accuracy: 0.8987
Epoch 58/100
accuracy: 0.8954 - val_loss: 0.0731 - val_accuracy: 0.8992
Epoch 59/100
accuracy: 0.8954 - val_loss: 0.0734 - val_accuracy: 0.8989
Epoch 60/100
accuracy: 0.8946 - val_loss: 0.0742 - val_accuracy: 0.8961
Epoch 61/100
accuracy: 0.8945 - val_loss: 0.0727 - val_accuracy: 0.9002
Epoch 62/100
accuracy: 0.8958 - val_loss: 0.0742 - val_accuracy: 0.8972
Epoch 63/100
accuracy: 0.8968 - val_loss: 0.0728 - val_accuracy: 0.8996
Epoch 64/100
accuracy: 0.8960 - val_loss: 0.0733 - val_accuracy: 0.8992
Epoch 65/100
```

```
accuracy: 0.8945 - val_loss: 0.0738 - val_accuracy: 0.8968
Epoch 66/100
accuracy: 0.8951 - val_loss: 0.0728 - val_accuracy: 0.9004
Epoch 67/100
accuracy: 0.8954 - val_loss: 0.0737 - val_accuracy: 0.8984
Epoch 68/100
accuracy: 0.8946 - val_loss: 0.0725 - val_accuracy: 0.9009
Epoch 69/100
accuracy: 0.8946 - val_loss: 0.0738 - val_accuracy: 0.8972
Epoch 70/100
accuracy: 0.8946 - val_loss: 0.0725 - val_accuracy: 0.9001
Epoch 71/100
accuracy: 0.8948 - val_loss: 0.0737 - val_accuracy: 0.8947
Epoch 72/100
accuracy: 0.8945 - val_loss: 0.0730 - val_accuracy: 0.9012
Epoch 73/100
accuracy: 0.8965 - val_loss: 0.0727 - val_accuracy: 0.9003
Epoch 74/100
accuracy: 0.8948 - val_loss: 0.0735 - val_accuracy: 0.8985
accuracy: 0.8949 - val_loss: 0.0728 - val_accuracy: 0.8980
Epoch 76/100
accuracy: 0.8944 - val_loss: 0.0724 - val_accuracy: 0.9003
Epoch 77/100
accuracy: 0.8959 - val loss: 0.0724 - val accuracy: 0.9004
Epoch 78/100
accuracy: 0.8941 - val_loss: 0.0722 - val_accuracy: 0.8994
Epoch 79/100
accuracy: 0.8951 - val_loss: 0.0727 - val_accuracy: 0.8977
Epoch 80/100
accuracy: 0.8949 - val_loss: 0.0728 - val_accuracy: 0.8992
Epoch 81/100
```

```
accuracy: 0.8947 - val_loss: 0.0722 - val_accuracy: 0.8982
Epoch 82/100
accuracy: 0.8955 - val_loss: 0.0720 - val_accuracy: 0.9010
Epoch 83/100
accuracy: 0.8958 - val_loss: 0.0726 - val_accuracy: 0.9001
Epoch 84/100
accuracy: 0.8951 - val_loss: 0.0729 - val_accuracy: 0.8975
Epoch 85/100
accuracy: 0.8949 - val_loss: 0.0730 - val_accuracy: 0.9003
Epoch 86/100
accuracy: 0.8951 - val_loss: 0.0724 - val_accuracy: 0.8992
Epoch 87/100
accuracy: 0.8952 - val_loss: 0.0723 - val_accuracy: 0.8985
Epoch 88/100
accuracy: 0.8964 - val_loss: 0.0727 - val_accuracy: 0.8992
Epoch 89/100
accuracy: 0.8959 - val_loss: 0.0725 - val_accuracy: 0.8996
Epoch 90/100
accuracy: 0.8944 - val_loss: 0.0723 - val_accuracy: 0.8983
Epoch 91/100
accuracy: 0.8960 - val_loss: 0.0726 - val_accuracy: 0.8987
Epoch 92/100
accuracy: 0.8954 - val_loss: 0.0731 - val_accuracy: 0.8983
Epoch 93/100
accuracy: 0.8952 - val loss: 0.0723 - val accuracy: 0.8991
Epoch 94/100
accuracy: 0.8958 - val_loss: 0.0721 - val_accuracy: 0.9001
Epoch 95/100
accuracy: 0.8945 - val_loss: 0.0727 - val_accuracy: 0.8984
Epoch 96/100
accuracy: 0.8948 - val_loss: 0.0728 - val_accuracy: 0.8993
Epoch 97/100
```

```
accuracy: 0.8962 - val_loss: 0.0725 - val_accuracy: 0.8989
Epoch 98/100
accuracy: 0.8943 - val_loss: 0.0745 - val_accuracy: 0.8936
Epoch 99/100
accuracy: 0.8963 - val_loss: 0.0724 - val_accuracy: 0.8996
Epoch 100/100
accuracy: 0.8951 - val_loss: 0.0723 - val_accuracy: 0.9000
Epoch 1/100
accuracy: 0.8949 - val_loss: 0.0727 - val_accuracy: 0.8997
Epoch 2/100
accuracy: 0.8959 - val_loss: 0.0721 - val_accuracy: 0.9011
Epoch 3/100
accuracy: 0.8966 - val_loss: 0.0728 - val_accuracy: 0.9004
Epoch 4/100
accuracy: 0.8954 - val_loss: 0.0726 - val_accuracy: 0.8979
Epoch 5/100
accuracy: 0.8960 - val_loss: 0.0723 - val_accuracy: 0.9005
Epoch 6/100
accuracy: 0.8951 - val_loss: 0.0719 - val_accuracy: 0.9006
Epoch 7/100
accuracy: 0.8960 - val_loss: 0.0725 - val_accuracy: 0.9005
Epoch 8/100
accuracy: 0.8961 - val_loss: 0.0726 - val_accuracy: 0.9005
Epoch 9/100
accuracy: 0.8951 - val loss: 0.0720 - val accuracy: 0.9011
Epoch 10/100
accuracy: 0.8953 - val_loss: 0.0724 - val_accuracy: 0.8993
Epoch 11/100
accuracy: 0.8963 - val_loss: 0.0725 - val_accuracy: 0.9003
Epoch 12/100
accuracy: 0.8955 - val_loss: 0.0734 - val_accuracy: 0.9002
Epoch 13/100
```

```
accuracy: 0.8964 - val_loss: 0.0730 - val_accuracy: 0.8997
Epoch 14/100
accuracy: 0.8949 - val_loss: 0.0730 - val_accuracy: 0.8996
Epoch 15/100
accuracy: 0.8954 - val_loss: 0.0734 - val_accuracy: 0.8998
Epoch 16/100
accuracy: 0.8946 - val_loss: 0.0744 - val_accuracy: 0.8971
Epoch 17/100
accuracy: 0.8959 - val_loss: 0.0724 - val_accuracy: 0.8999
Epoch 18/100
accuracy: 0.8951 - val_loss: 0.0730 - val_accuracy: 0.8980
Epoch 19/100
accuracy: 0.8954 - val_loss: 0.0731 - val_accuracy: 0.8989
Epoch 20/100
accuracy: 0.8957 - val_loss: 0.0724 - val_accuracy: 0.8994
Epoch 21/100
accuracy: 0.8953 - val_loss: 0.0738 - val_accuracy: 0.8975
Epoch 22/100
accuracy: 0.8958 - val_loss: 0.0728 - val_accuracy: 0.8980
Epoch 23/100
accuracy: 0.8956 - val_loss: 0.0731 - val_accuracy: 0.9001
Epoch 24/100
accuracy: 0.8953 - val_loss: 0.0728 - val_accuracy: 0.8990
Epoch 25/100
accuracy: 0.8943 - val loss: 0.0734 - val accuracy: 0.8973
Epoch 26/100
accuracy: 0.8954 - val_loss: 0.0730 - val_accuracy: 0.9004
Epoch 27/100
accuracy: 0.8958 - val_loss: 0.0730 - val_accuracy: 0.8997
Epoch 28/100
accuracy: 0.8951 - val_loss: 0.0725 - val_accuracy: 0.8986
Epoch 29/100
```

```
accuracy: 0.8952 - val_loss: 0.0725 - val_accuracy: 0.9004
Epoch 30/100
accuracy: 0.8965 - val_loss: 0.0735 - val_accuracy: 0.8974
Epoch 31/100
accuracy: 0.8959 - val_loss: 0.0731 - val_accuracy: 0.8980
Epoch 32/100
accuracy: 0.8955 - val_loss: 0.0728 - val_accuracy: 0.8993
Epoch 33/100
accuracy: 0.8948 - val_loss: 0.0735 - val_accuracy: 0.8964
Epoch 34/100
accuracy: 0.8964 - val_loss: 0.0727 - val_accuracy: 0.9009
Epoch 35/100
accuracy: 0.8950 - val_loss: 0.0732 - val_accuracy: 0.8980
Epoch 36/100
accuracy: 0.8956 - val_loss: 0.0728 - val_accuracy: 0.8996
Epoch 37/100
accuracy: 0.8959 - val_loss: 0.0734 - val_accuracy: 0.8963
Epoch 38/100
accuracy: 0.8949 - val_loss: 0.0723 - val_accuracy: 0.8996
Epoch 39/100
accuracy: 0.8969 - val_loss: 0.0726 - val_accuracy: 0.9006
Epoch 40/100
accuracy: 0.8965 - val_loss: 0.0727 - val_accuracy: 0.8988
Epoch 41/100
accuracy: 0.8966 - val loss: 0.0733 - val accuracy: 0.8980
Epoch 42/100
accuracy: 0.8947 - val_loss: 0.0726 - val_accuracy: 0.9001
Epoch 43/100
accuracy: 0.8963 - val_loss: 0.0737 - val_accuracy: 0.8965
Epoch 44/100
accuracy: 0.8961 - val_loss: 0.0729 - val_accuracy: 0.8976
Epoch 45/100
```

```
accuracy: 0.8943 - val_loss: 0.0745 - val_accuracy: 0.8957
Epoch 46/100
accuracy: 0.8966 - val_loss: 0.0722 - val_accuracy: 0.9001
Epoch 47/100
accuracy: 0.8960 - val_loss: 0.0729 - val_accuracy: 0.9001
Epoch 48/100
accuracy: 0.8965 - val_loss: 0.0726 - val_accuracy: 0.8998
Epoch 49/100
accuracy: 0.8963 - val_loss: 0.0721 - val_accuracy: 0.9004
Epoch 50/100
accuracy: 0.8962 - val_loss: 0.0724 - val_accuracy: 0.8987
Epoch 51/100
accuracy: 0.8970 - val_loss: 0.0724 - val_accuracy: 0.8992
Epoch 52/100
accuracy: 0.8953 - val_loss: 0.0731 - val_accuracy: 0.9006
Epoch 53/100
accuracy: 0.8958 - val_loss: 0.0733 - val_accuracy: 0.8981
Epoch 54/100
accuracy: 0.8967 - val_loss: 0.0729 - val_accuracy: 0.9003
accuracy: 0.8965 - val_loss: 0.0727 - val_accuracy: 0.8997
Epoch 56/100
accuracy: 0.8956 - val_loss: 0.0729 - val_accuracy: 0.8996
Epoch 57/100
accuracy: 0.8961 - val loss: 0.0728 - val accuracy: 0.9008
Epoch 58/100
accuracy: 0.8947 - val_loss: 0.0725 - val_accuracy: 0.8997
Epoch 59/100
accuracy: 0.8949 - val_loss: 0.0722 - val_accuracy: 0.9000
Epoch 60/100
accuracy: 0.8955 - val_loss: 0.0726 - val_accuracy: 0.9007
Epoch 61/100
```

```
accuracy: 0.8952 - val_loss: 0.0731 - val_accuracy: 0.9016
Epoch 62/100
accuracy: 0.8958 - val_loss: 0.0736 - val_accuracy: 0.8996
Epoch 63/100
accuracy: 0.8955 - val_loss: 0.0729 - val_accuracy: 0.9003
Epoch 64/100
accuracy: 0.8952 - val_loss: 0.0734 - val_accuracy: 0.9002
Epoch 65/100
accuracy: 0.8962 - val_loss: 0.0724 - val_accuracy: 0.9004
Epoch 66/100
accuracy: 0.8956 - val_loss: 0.0733 - val_accuracy: 0.8988
Epoch 67/100
accuracy: 0.8960 - val_loss: 0.0733 - val_accuracy: 0.9000
Epoch 68/100
accuracy: 0.8964 - val_loss: 0.0727 - val_accuracy: 0.8987
Epoch 69/100
accuracy: 0.8959 - val_loss: 0.0727 - val_accuracy: 0.8996
Epoch 70/100
accuracy: 0.8953 - val_loss: 0.0728 - val_accuracy: 0.9005
Epoch 71/100
accuracy: 0.8961 - val_loss: 0.0731 - val_accuracy: 0.9003
Epoch 72/100
accuracy: 0.8950 - val_loss: 0.0723 - val_accuracy: 0.9001
Epoch 73/100
accuracy: 0.8950 - val loss: 0.0734 - val accuracy: 0.8977
Epoch 74/100
accuracy: 0.8967 - val_loss: 0.0725 - val_accuracy: 0.9008
Epoch 75/100
accuracy: 0.8957 - val_loss: 0.0729 - val_accuracy: 0.8984
Epoch 76/100
accuracy: 0.8960 - val_loss: 0.0732 - val_accuracy: 0.8991
Epoch 77/100
```

```
accuracy: 0.8943 - val_loss: 0.0724 - val_accuracy: 0.9006
Epoch 78/100
accuracy: 0.8955 - val_loss: 0.0729 - val_accuracy: 0.9002
Epoch 79/100
accuracy: 0.8962 - val_loss: 0.0731 - val_accuracy: 0.8981
Epoch 80/100
accuracy: 0.8959 - val_loss: 0.0724 - val_accuracy: 0.9007
Epoch 81/100
accuracy: 0.8959 - val_loss: 0.0730 - val_accuracy: 0.8988
Epoch 82/100
accuracy: 0.8965 - val_loss: 0.0726 - val_accuracy: 0.8982
Epoch 83/100
accuracy: 0.8961 - val_loss: 0.0728 - val_accuracy: 0.8990
Epoch 84/100
accuracy: 0.8958 - val_loss: 0.0728 - val_accuracy: 0.8988
Epoch 85/100
accuracy: 0.8965 - val_loss: 0.0726 - val_accuracy: 0.8992
Epoch 86/100
accuracy: 0.8959 - val_loss: 0.0725 - val_accuracy: 0.9000
Epoch 87/100
accuracy: 0.8957 - val_loss: 0.0735 - val_accuracy: 0.9000
Epoch 88/100
accuracy: 0.8951 - val_loss: 0.0731 - val_accuracy: 0.8980
Epoch 89/100
accuracy: 0.8952 - val_loss: 0.0727 - val_accuracy: 0.9002
Epoch 90/100
accuracy: 0.8955 - val_loss: 0.0738 - val_accuracy: 0.8986
Epoch 91/100
accuracy: 0.8952 - val_loss: 0.0728 - val_accuracy: 0.8994
Epoch 92/100
accuracy: 0.8953 - val_loss: 0.0729 - val_accuracy: 0.8999
Epoch 93/100
```

```
accuracy: 0.8954 - val_loss: 0.0737 - val_accuracy: 0.8997
Epoch 94/100
accuracy: 0.8962 - val_loss: 0.0726 - val_accuracy: 0.9003
Epoch 95/100
accuracy: 0.8959 - val_loss: 0.0727 - val_accuracy: 0.9001
Epoch 96/100
accuracy: 0.8956 - val_loss: 0.0731 - val_accuracy: 0.8999
Epoch 97/100
accuracy: 0.8963 - val_loss: 0.0733 - val_accuracy: 0.8977
Epoch 98/100
accuracy: 0.8948 - val_loss: 0.0722 - val_accuracy: 0.9008
Epoch 99/100
accuracy: 0.8959 - val_loss: 0.0727 - val_accuracy: 0.8995
Epoch 100/100
accuracy: 0.8966 - val_loss: 0.0726 - val_accuracy: 0.9004
Epoch 1/100
accuracy: 0.8969 - val_loss: 0.0722 - val_accuracy: 0.9000
Epoch 2/100
accuracy: 0.8966 - val_loss: 0.0730 - val_accuracy: 0.8988
accuracy: 0.8958 - val_loss: 0.0731 - val_accuracy: 0.8986
Epoch 4/100
accuracy: 0.8965 - val_loss: 0.0731 - val_accuracy: 0.8967
Epoch 5/100
accuracy: 0.8964 - val_loss: 0.0727 - val_accuracy: 0.9009
Epoch 6/100
accuracy: 0.8954 - val_loss: 0.0726 - val_accuracy: 0.8982
Epoch 7/100
accuracy: 0.8966 - val_loss: 0.0727 - val_accuracy: 0.8991
Epoch 8/100
accuracy: 0.8955 - val_loss: 0.0730 - val_accuracy: 0.8985
Epoch 9/100
```

```
accuracy: 0.8954 - val_loss: 0.0724 - val_accuracy: 0.9004
Epoch 10/100
accuracy: 0.8956 - val_loss: 0.0725 - val_accuracy: 0.9002
Epoch 11/100
accuracy: 0.8966 - val_loss: 0.0728 - val_accuracy: 0.8999
Epoch 12/100
accuracy: 0.8969 - val_loss: 0.0723 - val_accuracy: 0.9003
Epoch 13/100
accuracy: 0.8959 - val_loss: 0.0727 - val_accuracy: 0.8992
Epoch 14/100
accuracy: 0.8957 - val_loss: 0.0725 - val_accuracy: 0.9009
Epoch 15/100
accuracy: 0.8963 - val_loss: 0.0729 - val_accuracy: 0.8991
Epoch 16/100
accuracy: 0.8963 - val_loss: 0.0725 - val_accuracy: 0.9004
Epoch 17/100
accuracy: 0.8960 - val_loss: 0.0719 - val_accuracy: 0.9010
Epoch 18/100
accuracy: 0.8953 - val_loss: 0.0729 - val_accuracy: 0.8981
Epoch 19/100
accuracy: 0.8964 - val_loss: 0.0730 - val_accuracy: 0.8983
Epoch 20/100
accuracy: 0.8959 - val_loss: 0.0727 - val_accuracy: 0.8992
Epoch 21/100
accuracy: 0.8963 - val loss: 0.0736 - val accuracy: 0.8975
Epoch 22/100
accuracy: 0.8946 - val_loss: 0.0732 - val_accuracy: 0.8996
Epoch 23/100
accuracy: 0.8955 - val_loss: 0.0727 - val_accuracy: 0.8994
Epoch 24/100
accuracy: 0.8960 - val_loss: 0.0728 - val_accuracy: 0.8993
Epoch 25/100
```

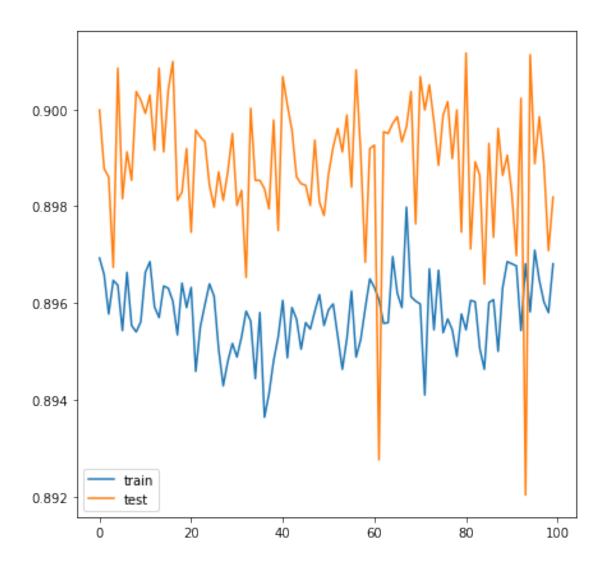
```
accuracy: 0.8964 - val_loss: 0.0731 - val_accuracy: 0.8984
Epoch 26/100
accuracy: 0.8961 - val_loss: 0.0733 - val_accuracy: 0.8980
Epoch 27/100
accuracy: 0.8950 - val_loss: 0.0729 - val_accuracy: 0.8987
Epoch 28/100
accuracy: 0.8943 - val_loss: 0.0732 - val_accuracy: 0.8981
Epoch 29/100
accuracy: 0.8948 - val_loss: 0.0725 - val_accuracy: 0.8987
Epoch 30/100
accuracy: 0.8952 - val_loss: 0.0728 - val_accuracy: 0.8995
Epoch 31/100
accuracy: 0.8949 - val_loss: 0.0725 - val_accuracy: 0.8980
Epoch 32/100
accuracy: 0.8953 - val_loss: 0.0737 - val_accuracy: 0.8983
Epoch 33/100
accuracy: 0.8958 - val_loss: 0.0736 - val_accuracy: 0.8965
Epoch 34/100
accuracy: 0.8956 - val_loss: 0.0723 - val_accuracy: 0.9000
accuracy: 0.8944 - val_loss: 0.0725 - val_accuracy: 0.8985
Epoch 36/100
accuracy: 0.8958 - val_loss: 0.0730 - val_accuracy: 0.8985
Epoch 37/100
accuracy: 0.8936 - val loss: 0.0723 - val accuracy: 0.8984
Epoch 38/100
accuracy: 0.8941 - val_loss: 0.0727 - val_accuracy: 0.8979
Epoch 39/100
accuracy: 0.8948 - val_loss: 0.0727 - val_accuracy: 0.8998
Epoch 40/100
accuracy: 0.8953 - val_loss: 0.0732 - val_accuracy: 0.8975
Epoch 41/100
```

```
accuracy: 0.8961 - val_loss: 0.0721 - val_accuracy: 0.9007
Epoch 42/100
accuracy: 0.8949 - val_loss: 0.0726 - val_accuracy: 0.9001
Epoch 43/100
accuracy: 0.8959 - val_loss: 0.0723 - val_accuracy: 0.8996
Epoch 44/100
accuracy: 0.8957 - val_loss: 0.0728 - val_accuracy: 0.8986
Epoch 45/100
accuracy: 0.8950 - val_loss: 0.0729 - val_accuracy: 0.8985
Epoch 46/100
accuracy: 0.8956 - val_loss: 0.0727 - val_accuracy: 0.8984
Epoch 47/100
accuracy: 0.8955 - val_loss: 0.0733 - val_accuracy: 0.8980
Epoch 48/100
accuracy: 0.8958 - val_loss: 0.0726 - val_accuracy: 0.8994
Epoch 49/100
accuracy: 0.8962 - val_loss: 0.0731 - val_accuracy: 0.8981
Epoch 50/100
accuracy: 0.8955 - val_loss: 0.0734 - val_accuracy: 0.8978
accuracy: 0.8959 - val_loss: 0.0729 - val_accuracy: 0.8987
Epoch 52/100
accuracy: 0.8960 - val_loss: 0.0728 - val_accuracy: 0.8992
Epoch 53/100
accuracy: 0.8953 - val loss: 0.0723 - val accuracy: 0.8996
Epoch 54/100
accuracy: 0.8946 - val_loss: 0.0734 - val_accuracy: 0.8991
Epoch 55/100
accuracy: 0.8953 - val_loss: 0.0726 - val_accuracy: 0.8999
Epoch 56/100
accuracy: 0.8962 - val_loss: 0.0739 - val_accuracy: 0.8984
Epoch 57/100
```

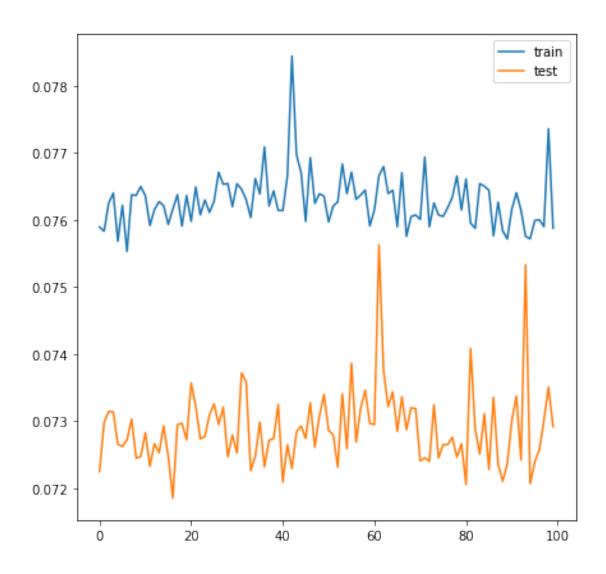
```
accuracy: 0.8949 - val_loss: 0.0727 - val_accuracy: 0.9008
Epoch 58/100
accuracy: 0.8952 - val_loss: 0.0732 - val_accuracy: 0.8991
Epoch 59/100
accuracy: 0.8959 - val_loss: 0.0735 - val_accuracy: 0.8968
Epoch 60/100
accuracy: 0.8965 - val_loss: 0.0730 - val_accuracy: 0.8992
Epoch 61/100
accuracy: 0.8963 - val_loss: 0.0729 - val_accuracy: 0.8993
Epoch 62/100
accuracy: 0.8961 - val_loss: 0.0756 - val_accuracy: 0.8928
Epoch 63/100
accuracy: 0.8956 - val_loss: 0.0738 - val_accuracy: 0.8995
Epoch 64/100
accuracy: 0.8956 - val_loss: 0.0732 - val_accuracy: 0.8995
Epoch 65/100
accuracy: 0.8970 - val_loss: 0.0734 - val_accuracy: 0.8997
Epoch 66/100
accuracy: 0.8962 - val_loss: 0.0728 - val_accuracy: 0.8998
Epoch 67/100
accuracy: 0.8959 - val_loss: 0.0734 - val_accuracy: 0.8993
Epoch 68/100
accuracy: 0.8980 - val_loss: 0.0729 - val_accuracy: 0.8996
Epoch 69/100
accuracy: 0.8961 - val loss: 0.0732 - val accuracy: 0.9004
Epoch 70/100
accuracy: 0.8960 - val_loss: 0.0732 - val_accuracy: 0.8976
Epoch 71/100
accuracy: 0.8960 - val_loss: 0.0724 - val_accuracy: 0.9007
Epoch 72/100
accuracy: 0.8941 - val_loss: 0.0724 - val_accuracy: 0.9000
Epoch 73/100
```

```
accuracy: 0.8967 - val_loss: 0.0724 - val_accuracy: 0.9005
Epoch 74/100
accuracy: 0.8954 - val_loss: 0.0732 - val_accuracy: 0.8997
Epoch 75/100
accuracy: 0.8967 - val_loss: 0.0724 - val_accuracy: 0.8988
Epoch 76/100
accuracy: 0.8954 - val_loss: 0.0726 - val_accuracy: 0.8999
Epoch 77/100
accuracy: 0.8957 - val_loss: 0.0726 - val_accuracy: 0.9002
Epoch 78/100
accuracy: 0.8954 - val_loss: 0.0728 - val_accuracy: 0.8990
Epoch 79/100
accuracy: 0.8949 - val_loss: 0.0725 - val_accuracy: 0.9000
Epoch 80/100
accuracy: 0.8958 - val_loss: 0.0727 - val_accuracy: 0.8975
Epoch 81/100
accuracy: 0.8954 - val_loss: 0.0721 - val_accuracy: 0.9012
Epoch 82/100
accuracy: 0.8961 - val_loss: 0.0741 - val_accuracy: 0.8971
Epoch 83/100
accuracy: 0.8960 - val_loss: 0.0729 - val_accuracy: 0.8989
Epoch 84/100
accuracy: 0.8951 - val_loss: 0.0725 - val_accuracy: 0.8986
Epoch 85/100
accuracy: 0.8946 - val loss: 0.0731 - val accuracy: 0.8964
Epoch 86/100
accuracy: 0.8960 - val_loss: 0.0723 - val_accuracy: 0.8993
Epoch 87/100
accuracy: 0.8961 - val_loss: 0.0734 - val_accuracy: 0.8974
Epoch 88/100
accuracy: 0.8950 - val_loss: 0.0724 - val_accuracy: 0.8996
Epoch 89/100
```

```
accuracy: 0.8963 - val_loss: 0.0721 - val_accuracy: 0.8986
  Epoch 90/100
  accuracy: 0.8969 - val_loss: 0.0724 - val_accuracy: 0.8991
  Epoch 91/100
  accuracy: 0.8968 - val_loss: 0.0730 - val_accuracy: 0.8983
  Epoch 92/100
  accuracy: 0.8968 - val_loss: 0.0734 - val_accuracy: 0.8970
  Epoch 93/100
  accuracy: 0.8954 - val_loss: 0.0724 - val_accuracy: 0.9002
  Epoch 94/100
  accuracy: 0.8968 - val_loss: 0.0753 - val_accuracy: 0.8920
  Epoch 95/100
  accuracy: 0.8958 - val_loss: 0.0721 - val_accuracy: 0.9011
  Epoch 96/100
  accuracy: 0.8971 - val_loss: 0.0724 - val_accuracy: 0.8989
  Epoch 97/100
  accuracy: 0.8965 - val_loss: 0.0726 - val_accuracy: 0.8998
  Epoch 98/100
  accuracy: 0.8960 - val_loss: 0.0730 - val_accuracy: 0.8989
  Epoch 99/100
  accuracy: 0.8958 - val_loss: 0.0735 - val_accuracy: 0.8971
  Epoch 100/100
  accuracy: 0.8968 - val_loss: 0.0729 - val_accuracy: 0.8982
[93]: plt.figure(figsize=(7, 7))
   plt.plot(history2.history['accuracy'], label='train')
   plt.plot(history2.history['val_accuracy'], label='test')
   plt.legend()
   plt.savefig('accuracy_mean_s_e.jpg', dpi=200)
```



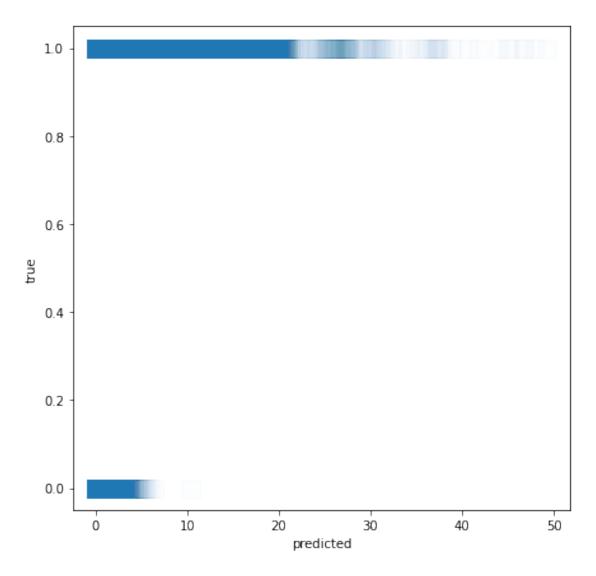
```
[94]: plt.figure(figsize=(7, 7))
   plt.plot(history2.history['loss'], label='train')
   plt.plot(history2.history['val_loss'], label='test')
   plt.legend()
   plt.savefig('loss_mean_s_e.jpg', dpi=200)
```



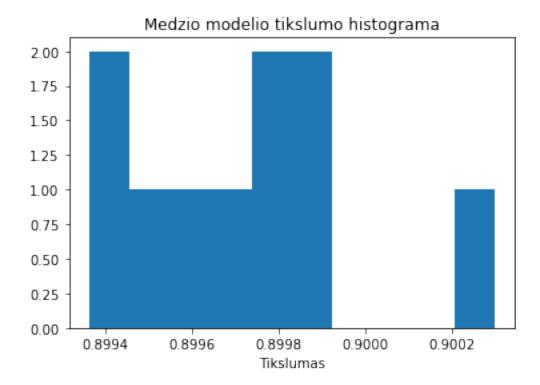
```
[95]: y_predict = model(x_train)
      y_predict
[95]: <tf.Tensor: shape=(67447, 1), dtype=float32, numpy=
      array([[0.
                       ],
             [1.1922374],
             [0.
                       ],
             [4.780798],
             [0.
                       ],
                       ]], dtype=float32)>
             [0.
[96]: plt.figure(figsize=(7, 7))
      plt.scatter(y_predict, y_train, s=200, marker='s', alpha=0.01)
      plt.xlabel('predicted')
```

```
plt.ylabel('true')
```

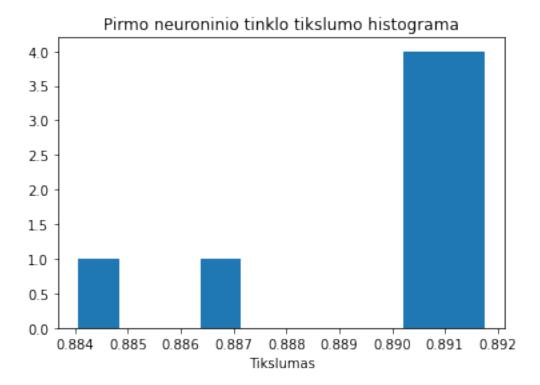
[96]: Text(0, 0.5, 'true')



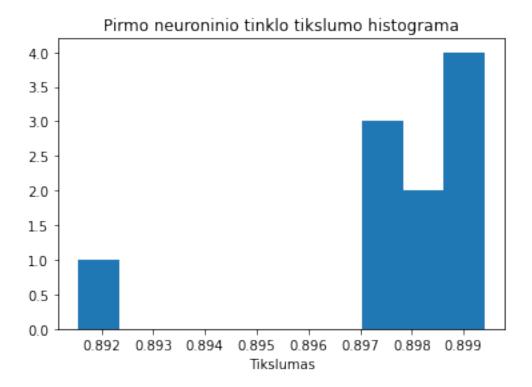
```
0.8903604763746261,
        0.8908493053913117,
        0.890832349061966,
        0.891294886469841,
        0.8907174974679947,
        0.8913187605142593,
        0.8917460054159164,
        0.8915069508552551]
[115]: val_accuracy_list2
[115]: [0.8915581548213959,
        0.8970635843276977,
        0.897506052851677,
        0.8975582921504974,
        0.8982560712099076,
        0.8984875082969666,
        0.8987352114915848,
        0.8989365547895432,
        0.8994035828113556,
        0.8988984972238541]
[119]: | score_list = []
       while len(score_list) != 10:
           model = DecisionTreeClassifier()
           model.fit(x_train, y_train)
           predictions = model.predict(x_test)
           score = accuracy_score(y_test, predictions)
           score_list.append(score)
[137]: plt.hist(score_list)
       plt.xlabel('Tikslumas')
       plt.title('Medzio modelio tikslumo histograma')
       plt.savefig('tree_hist.png')
```



```
[138]: plt.hist(val_accuracy_list)
   plt.xlabel('Tikslumas')
   plt.title('Pirmo neuroninio tinklo tikslumo histograma')
   plt.savefig('model_hist.png')
```



```
[135]: plt.hist(val_accuracy_list2)
   plt.xlabel('Tikslumas')
   plt.title('Antro neuroninio tinklo tikslumo histograma')
   plt.savefig('model2_hist.png', dpi=200)
```



```
[125]: score_mean = sum(score_list) / len(score_list)
       score_mean = score_mean.round(3)
[126]: model1_mean = sum(history.history['val_accuracy']) / len(history.
        ⇔history['val_accuracy'])
       model1_mean = round(model1_mean, 3)
[127]: model2_mean = sum(history2.history['val_accuracy']) / len(history2.
        ⇔history['val_accuracy'])
       model2_mean = round(model2_mean, 3)
[128]: print('Medzio klasifikavimo rezultatas yra', score mean)
       print('1 neuroninio tinklo klasifikavimo rezultatas yra', model1_mean)
       print('2 neuroninio tinklo klasifikavimo rezultatas yra', model2 mean)
      Medzio klasifikavimo rezultatas yra 0.9
      1 neuroninio tinklo klasifikavimo rezultatas yra 0.892
      2 neuroninio tinklo klasifikavimo rezultatas yra 0.899
[103]: model = DecisionTreeClassifier()
       model.fit(x data, y data)
       predictions = model.predict(x_data)
       y_data = y_data.astype(str)
```

```
Traceback (most recent call last)
Input In [103], in <cell line: 7>()
                 3 predictions = model.predict(x_data)
                 5 y_data = y_data.astype(str)
 ---> 7 tree.export_graphviz(model,
                8
                                                                                  out_file='price_predicttion_tree_new.dot',
                9. .
                                                                  feature_names=['Year', 'Distance_km', 'Power', 'Engine_s ze'],
                                                                                  class names=sorted(y data.unique()),
              10
              11
                                                                                 label='all',
              12
                                                                                 rounded=True,
              13
                                                                                 filled=True
              14
File ~\anaconda3\lib\site-packages\sklearn\tree\_export.py:889, in_
    →export_graphviz(decision_tree, out_file, max_depth, feature_names, u oclass_names, label, filled, leaves_parallel, impurity, node_ids, proportion, terms of the control o
    orotate, rounded, special_characters, precision, fontname)
           870
                                 out_file = StringIO()
           872 exporter = _DOTTreeExporter(
           873
                                 out file=out file,
           874
                                 max_depth=max_depth,
         (...)
           887
                                 fontname=fontname,
 --> 889 exporter.export(decision_tree)
           891 if return_string:
                                 return exporter.out_file.getvalue()
           892
File ~\anaconda3\lib\site-packages\sklearn\tree\_export.py:452, in_
    → DOTTreeExporter.export(self, decision_tree)
           450 if self.feature_names is not None:
                                 if len(self.feature_names) != decision_tree.n_features_in_:
           451
--> 452
                                            raise ValueError(
           453
                                                         "Length of feature_names, %d does not match number of ___
    ⇔features, %d"
```

```
454 % (len(self.feature_names), decision_tree.n_features_in_)
455 )
456 # each part writes to out_file
457 self.head()

ValueError: Length of feature_names, 4 does not match number of features, 5
```

```
[]: | # X_input = predict_data.drop(columns=['Price'])__
    # Y output = predict data['Price']
    # model = DecisionTreeClassifier()
    # model.fit(X_input.values, Y_output.values)
    # # Y_output = Y_output.astype(str)
    # tree.export_graphviz(model,
                        out_file='price-predictor.dot',
                        feature_names=['Power', 'Engine CC'],
    #
                        class_names=sorted(Y_output.unique()),
    #
                        label='all',
    #
                        rounded=True,
    #
                        filled=True
    #
```