Sieci neuronowe i sztuczna inteligencja – laboratorium 8 Monika Błyszcz, 236623

Kod:

```
1 # import os utilities
 2 import os
 3
 4 # import numpy
 5 import numpy as np
 7 # images
 8 import skimage
 9 from skimage import transform
10 from skimage.color import rgb2gray
11
12 # Import the `pyplot` module
13 import matplotlib.pyplot as plt
14
15
16 # function to load data
17 def load_data(data_directory):
18
       """Loads sign images data from their folder.
19
20
       Returns:
21
           images: list of images, i.e., signs
22
           labels: list of labels, i.e., signs IDs
23
24
      # We need back labels and the row images
25
      images = []
26
      labels = []
27
28
      # We have one folder per sign type
29
       directories = []
30
      for d in os.listdir(data_directory):
           if os.path.isdir(os.path.join(data_directory, d)):
31
32
               directories.append(d)
33
34
       # In each foder there are not only images but also csv
    description
      # files
35
36
       for d in directories:
37
           label_directory = os.path.join(data_directory, d)
38
           file_names = [
39
               os.path.join(label_directory, f)
40
               for f in os.listdir(label_directory)
41
               if f.endswith(".ppm")
42
           1
43
           for f in file_names:
44
45
               images.append(skimage.io.imread(f))
```

```
46
               labels.append(int(d))
47
48
       return images, labels
49
50
51 ROOT_PATH = os.getcwd()
52
53 # Download training data
54
55 train_data_directory = os.path.join(ROOT_PATH, "Training")
56 test_data_directory = os.path.join(ROOT_PATH, "Testing")
57
58 images, labels = load_data(train_data_directory)
59 test_images, test_labels = load_data(test_data_directory)
60
61 print(labels)
62
63 ## The following commented lines were reported in the
   DataCamp materials
64 ## but they does not work here
65 print(images.ndim)
66 print(images.size)
67 \text{ images}[0]
68 print(len(images))
69 print(len(labels))
70
71 # this should be a bar plot but an histogram with the same
   number of
72 # bins that that unique levels of the labels list should
   be fine :-)
73 unique_labels = set(labels)
74 n_labels = max(unique_labels) + 1
75
76 # Make a histogram with 62 bins of the `labels` data
77 plt.hist(labels, n_labels)
78
79 # Show the plot
80 plt.show()
81
82 # Determine the (random) indexes of the images that you
   want to see
83 traffic_signs = [300, 2250, 3650, 4000]
84
85 # Fill out the subplots with the random images that you
   defined
86 for i in range(len(traffic_signs)):
```

```
87
        plt.subplot(1, 4, i + 1)
 88
        plt.axis('off')
 89
        plt.imshow(images[traffic_signs[i]])
 90
        plt.subplots_adjust(wspace=0.5)
 91
        plt.show()
 92
        print(
            "shape: {0}, min value: {1}, max value: {2}".
 93
    format(
 94
                images[traffic_signs[i]].shape,
 95
                images[traffic_signs[i]].min(),
 96
                images[traffic_signs[i]].max()
 97
            )
 98
        )
 99
100 # Plot a grid with a sample of all the signs
101 plt.figure(figsize=(15, 15))
102
103 i = 1
104
105 for label in unique labels:
        # pick the first image for the label.
106
107
        #
        # The index() method searches an element in the list
108
    and returns its
109
        # index. In simple terms, index() method finds the
    given element in
110
        # a list and returns its position. However, if the
    same element is
        # present more than once, index() method returns its
111
    smallest/first
112
        # position.
        image = images[labels.index(label)]
113
114
115
        # We have 62 images. Hence, define a 64 grid sub-
   plots
116
        plt.subplot(8, 8, i)
117
        # Don't include axes
118
        plt.axis('off')
119
120
121
        # Add a title to each subplot
122
        # The count() method returns the number of elements
123
    with the
124
        # specified value.
        plt.title("Label {0} ({1})".format(label, labels.
125
```

```
125 count(label)))
126
127
        # Add 1 to the counter
128
        i += 1
129
        # Plot this first image
130
        plt.imshow(image)
131
132
133 plt.show()
134
135 # To tackle the differing image sizes, you're going to
    rescale the images
136 images_{28} = [
        transform.resize(image, (28, 28))
137
138
        for image in images
139 ]
140
141 # Convert `images28` to an array
142 images_28 = np.array(images_28)
143
144 # Convert `images28` to grayscale
145 images_28 = rgb2gray(images_28)
146
147 for i in range(len(traffic_signs)):
        plt.subplot(1, 4, i + 1)
149
        plt.axis('off')
150
        plt.imshow(images_28[traffic_signs[i]], cmap="gray")
        plt.subplots_adjust(wspace=0.5)
151
152
153 plt.show()
154
155 # Test set
156 # Transform the images to 28 by 28 pixels
157 \text{ test_images}_28 = [
        transform.resize(image, (28, 28))
158
159
        for image in test_images
160 ]
161 # Convert to grayscale
162 test_images_28 = rgb2gray(np.array(test_images_28))
163
```

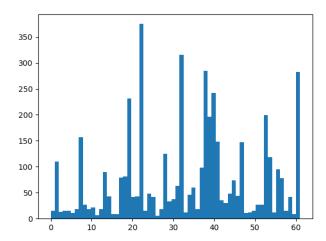
Komentarz:

W trakcie wykonywania zadania, w pierwszym kroku wczytano dane testowe i dane treningowe. Uzyskano dzięki temu zbiór obrazów i ich indeksów (etykiet). Następnie utworzono histogram obrazujący ilość znaków danej kategorii. Potem wyszukano 4 losowe kategorie obrazów. Po czym wykorzystując cały zbiór pokazano po 1 przykładzie z każdej kategorii. Na końcu 4 obrazy z klas wybranych wcześniej ukazano w odcieniach szarości

Wyniki:

```
C:\Users\Mo\AppData\loca\Microsoft\WindowsApps\python5.10.exe C:\Users\Mo\Documents\Studia\SN\Tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2823_SN\\tab\SN\PMR_2
```

Histogram:



Znaki:



Skala szarości:







