

DenkowskiStanislawlab3TF

October 28, 2022

```
[ ]: import tensorflow as tf
import tensorflow_datasets as tfds
import numpy as np
import matplotlib.pyplot as plt
import time
from PIL import Image
import requests
```

Ze względu na wielkie problemy z pytorchem i pytorchem lightningiem, zdecydowałem się zrobić tego lab'a przy pomocy tensorflow i keras.

Problemy miałem następujące: - dataloader bardzo bardzo spowalniał, jeśli zwiększałem liczbę workerów (nie jest to wielki problem) - Mój model, który moim zdaniem jest dobrze zdefiniowany, (nie wykluczam, że się pomyliłem ale porównywałem do bardzo wielu innych modeli w internecie), bardzo słabo się uczył - Obie kwestie nie gwarantują błędów po stronie pytorch'a, ale biorąc pod uwagę dosyć specyficzną maszynę na której pracuję i liczbę problemów jaką znalazłem wolałem zmienić framework - potencjalnie będę próbował wrócić do pytorch lightning, ale nie mam teraz czasu szukać problemów

```
[ ]: cifar10 = tf.keras.datasets.cifar10
(train_images, train_labels), (val_images, val_labels) = cifar10.load_data()
info = tfds.builder('cifar10').info
info
```

2022-10-27 11:10:12.223962: W

tensorflow/core/platform/cloud/google_auth_provider.cc:184] All attempts to get a Google authentication bearer token failed, returning an empty token.

Retrieving token from files failed with "NOT_FOUND: Could not locate the credentials file.". Retrieving token from GCE failed with "FAILED_PRECONDITION: Error executing an HTTP request: libcurl code 6 meaning 'Couldn't resolve host name', error details: Could not resolve host: metadata".

```
[ ]: tfds.core.DatasetInfo(
    name='cifar10',
    full_name='cifar10/3.0.2',
    description="""
    The CIFAR-10 dataset consists of 60000 32x32 colour images in 10 classes,
    with 6000 images per class. There are 50000 training images and 10000 test
    images.
    """,
```

```

homepage='https://www.cs.toronto.edu/~kriz/cifar.html',
data_path='/var/folders/8p/_cz7tfws6qj3stcp8jwc282r0000gn/T/tmpts2uasditfds',
file_format=tfrecord,
download_size=162.17 MiB,
dataset_size=132.40 MiB,
features=FeaturesDict({
    'id': Text(shape=(), dtype=tf.string),
    'image': Image(shape=(32, 32, 3), dtype=tf.uint8),
    'label': ClassLabel(shape=(), dtype=tf.int64, num_classes=10),
}),
supervised_keys=('image', 'label'),
disable_shuffling=False,
splits={
    'test': <SplitInfo num_examples=10000, num_shards=1>,
    'train': <SplitInfo num_examples=50000, num_shards=1>,
},
citation="""@TECHREPORT{Krizhevsky09learningmultiple,
author = {Alex Krizhevsky},
title = {Learning multiple layers of features from tiny images},
institution = {},
year = {2009}
}""",
)

```

```
[ ]: classes = info.features['label'].names
      classes
```

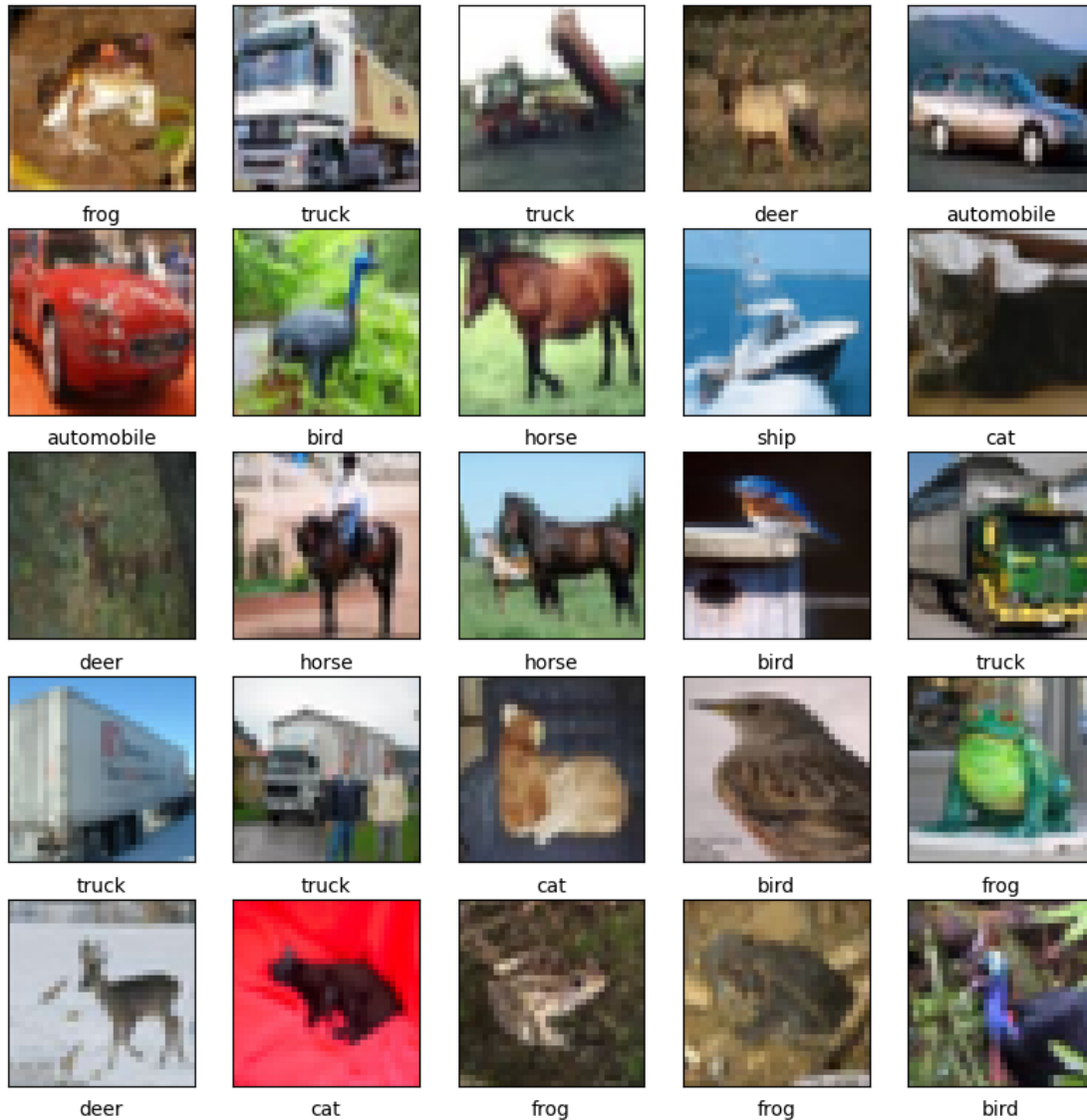
```
[ ]: ['airplane',
      'automobile',
      'bird',
      'cat',
      'deer',
      'dog',
      'frog',
      'horse',
      'ship',
      'truck']
```

```
[ ]: print(train_images.shape, train_labels.shape, val_images.shape, val_labels.
      ↪shape)
```

```
(50000, 32, 32, 3) (50000, 1) (10000, 32, 32, 3) (10000, 1)
```

```
[ ]: plt.figure(figsize=(10,10))
      for i in range(25):
          plt.subplot(5,5,i+1)
          plt.xticks([])
          plt.yticks([])
```

```
plt.grid(False)
plt.imshow(train_images[i])
plt.xlabel(classes[train_labels[i][0]])
plt.show()
```



Bardzo różnorodne zdjęcia.

```
[ ]: train_labels = tf.one_hot(tf.reshape(train_labels, train_labels.shape[0]),  
    ↪ depth=10)  
val_labels = tf.one_hot(tf.reshape(val_labels, val_labels.shape[0]), depth=10)  
train_images = train_images / 255.0  
val_images = val_images / 255.0
```

```
train_images[0].shape
```

Metal device set to: Apple M1 Pro

systemMemory: 32.00 GB

maxCacheSize: 10.67 GB

2022-10-27 11:10:15.855739: I

tensorflow/core/common_runtime/pluggable_device/pluggable_device_factory.cc:306]
Could not identify NUMA node of platform GPU ID 0, defaulting to 0. Your kernel
may not have been built with NUMA support.

2022-10-27 11:10:15.855866: I

tensorflow/core/common_runtime/pluggable_device/pluggable_device_factory.cc:272]
Created TensorFlow device (/job:localhost/replica:0/task:0/device:GPU:0 with 0
MB memory) -> physical PluggableDevice (device: 0, name: METAL, pci bus id:
<undefined>)

```
[ ]: (32, 32, 3)
```

Wykorzystałem funkcję `tf.one_hot` do zmiany kodowania labeli na one hot encoding.
Dodatkowo normalizuję zbiory danych.

```
[ ]: minmodel = tf.keras.Sequential([  
    tf.keras.layers.Conv2D(5, 3, padding='same', activation='sigmoid',  
    ↪ input_shape=train_images[0].shape),  
    tf.keras.layers.Conv2D(5, 3, padding='same', activation='sigmoid'),  
    tf.keras.layers.MaxPool2D(pool_size=(8,8), padding='valid'),  
    tf.keras.layers.Flatten(),  
    tf.keras.layers.Dense(units=10, activation='softmax')  
])
```

```
[ ]: minmodel.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 32, 32, 5)	140
conv2d_1 (Conv2D)	(None, 32, 32, 5)	230
max_pooling2d (MaxPooling2D)	(None, 4, 4, 5)	0
flatten (Flatten)	(None, 80)	0
dense (Dense)	(None, 10)	810

```
=====
Total params: 1,180
Trainable params: 1,180
Non-trainable params: 0
-----
```

Do implementacji minimalnej architektury wykorzystałem, `tf.keras.Sequential` - który gwarantuje nam sekwencyjne działanie zdefiniowanych warstw, oraz warstw - `Conv2D` dwa razy, `MaxPooling2D`, `Flatten`, `Dense`.

Zgodnie z poleceniem zadania.

Taki model ma 1180 parametrów.

```
[ ]: first10 = minmodel(train_images[:10])
print(tf.math.reduce_all(tf.one_hot(tf.math.argmax(first10, axis=1),
    ↪depth=10)==train_labels[:10], 1), '\n', tf.math.argmax(first10, axis=1))
print(tf.math.reduce_max(first10, axis=1))
```

```
tf.Tensor([False  True  True False False False False False False],
shape=(10,), dtype=bool)
tf.Tensor([9 9 9 9 9 9 9 9 9 9], shape=(10,), dtype=int64)
```

Jak widać model daje bardzo słabe wyniki, ma bardzo duży bias do jednej (czasem kilku konkretnych) klasy.

```
[ ]: minmodel.compile(
    optimizer=tf.keras.optimizers.SGD(learning_rate=0.001, momentum=0.9),
    loss=tf.keras.losses.CategoricalCrossentropy(),
    metrics=['accuracy']
)
```

```
[ ]: start_time=time.time()
minhistory = minmodel.fit(
    x=train_images,
    y=train_labels,
    batch_size=64,
    epochs=150,
    verbose=1,
    validation_data=(val_images, val_labels)
)
print(f"time: {time.time()-start_time}")
```

Epoch 1/150

```
2022-10-27 11:10:22.840351: W
tensorflow/core/platform/profile_utils/cpu_utils.cc:128] Failed to get CPU
frequency: 0 Hz
2022-10-27 11:10:23.076174: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.
```

780/782 [=====>.] - ETA: 0s - loss: 2.3099 - accuracy: 0.1015

2022-10-27 11:10:33.350608: I

tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]

Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - 11s 14ms/step - loss: 2.3099 - accuracy: 0.1014 - val_loss: 2.3022 - val_accuracy: 0.0999

Epoch 2/150

782/782 [=====] - 10s 13ms/step - loss: 2.3018 - accuracy: 0.1057 - val_loss: 2.3015 - val_accuracy: 0.1062

Epoch 3/150

782/782 [=====] - 10s 13ms/step - loss: 2.3006 - accuracy: 0.1091 - val_loss: 2.2996 - val_accuracy: 0.1402

Epoch 4/150

782/782 [=====] - 10s 13ms/step - loss: 2.2987 - accuracy: 0.1183 - val_loss: 2.2978 - val_accuracy: 0.1433

Epoch 5/150

782/782 [=====] - 10s 13ms/step - loss: 2.2966 - accuracy: 0.1284 - val_loss: 2.2950 - val_accuracy: 0.1022

Epoch 6/150

782/782 [=====] - 10s 13ms/step - loss: 2.2929 - accuracy: 0.1408 - val_loss: 2.2899 - val_accuracy: 0.1665

Epoch 7/150

782/782 [=====] - 10s 13ms/step - loss: 2.2865 - accuracy: 0.1493 - val_loss: 2.2819 - val_accuracy: 0.1681

Epoch 8/150

782/782 [=====] - 10s 13ms/step - loss: 2.2752 - accuracy: 0.1681 - val_loss: 2.2665 - val_accuracy: 0.1726

Epoch 9/150

782/782 [=====] - 10s 13ms/step - loss: 2.2556 - accuracy: 0.1844 - val_loss: 2.2413 - val_accuracy: 0.1896

Epoch 10/150

782/782 [=====] - 10s 13ms/step - loss: 2.2233 - accuracy: 0.2015 - val_loss: 2.1997 - val_accuracy: 0.2120

Epoch 11/150

782/782 [=====] - 10s 13ms/step - loss: 2.1747 - accuracy: 0.2247 - val_loss: 2.1422 - val_accuracy: 0.2356

Epoch 12/150

782/782 [=====] - 10s 13ms/step - loss: 2.1180 - accuracy: 0.2457 - val_loss: 2.0843 - val_accuracy: 0.2628

Epoch 13/150

782/782 [=====] - 10s 13ms/step - loss: 2.0684 - accuracy: 0.2684 - val_loss: 2.0402 - val_accuracy: 0.2783

Epoch 14/150

782/782 [=====] - 10s 13ms/step - loss: 2.0324 - accuracy: 0.2811 - val_loss: 2.0085 - val_accuracy: 0.2958

Epoch 15/150

782/782 [=====] - 10s 13ms/step - loss: 2.0066 -
accuracy: 0.2902 - val_loss: 1.9840 - val_accuracy: 0.3011
Epoch 16/150
782/782 [=====] - 10s 13ms/step - loss: 1.9859 -
accuracy: 0.2969 - val_loss: 1.9647 - val_accuracy: 0.2990
Epoch 17/150
782/782 [=====] - 11s 14ms/step - loss: 1.9685 -
accuracy: 0.3008 - val_loss: 1.9479 - val_accuracy: 0.3137
Epoch 18/150
782/782 [=====] - 10s 13ms/step - loss: 1.9526 -
accuracy: 0.3089 - val_loss: 1.9319 - val_accuracy: 0.3185
Epoch 19/150
782/782 [=====] - 10s 13ms/step - loss: 1.9374 -
accuracy: 0.3125 - val_loss: 1.9169 - val_accuracy: 0.3307
Epoch 20/150
782/782 [=====] - 10s 13ms/step - loss: 1.9229 -
accuracy: 0.3181 - val_loss: 1.9029 - val_accuracy: 0.3307
Epoch 21/150
782/782 [=====] - 10s 13ms/step - loss: 1.9091 -
accuracy: 0.3217 - val_loss: 1.8899 - val_accuracy: 0.3319
Epoch 22/150
782/782 [=====] - 10s 13ms/step - loss: 1.8956 -
accuracy: 0.3286 - val_loss: 1.8769 - val_accuracy: 0.3356
Epoch 23/150
782/782 [=====] - 10s 13ms/step - loss: 1.8831 -
accuracy: 0.3323 - val_loss: 1.8648 - val_accuracy: 0.3427
Epoch 24/150
782/782 [=====] - 10s 13ms/step - loss: 1.8706 -
accuracy: 0.3380 - val_loss: 1.8539 - val_accuracy: 0.3461
Epoch 25/150
782/782 [=====] - 10s 13ms/step - loss: 1.8586 -
accuracy: 0.3420 - val_loss: 1.8429 - val_accuracy: 0.3495
Epoch 26/150
782/782 [=====] - 10s 13ms/step - loss: 1.8473 -
accuracy: 0.3457 - val_loss: 1.8317 - val_accuracy: 0.3516
Epoch 27/150
782/782 [=====] - 10s 13ms/step - loss: 1.8362 -
accuracy: 0.3515 - val_loss: 1.8210 - val_accuracy: 0.3620
Epoch 28/150
782/782 [=====] - 10s 13ms/step - loss: 1.8259 -
accuracy: 0.3532 - val_loss: 1.8124 - val_accuracy: 0.3584
Epoch 29/150
782/782 [=====] - 10s 13ms/step - loss: 1.8161 -
accuracy: 0.3573 - val_loss: 1.8020 - val_accuracy: 0.3619
Epoch 30/150
782/782 [=====] - 10s 13ms/step - loss: 1.8063 -
accuracy: 0.3600 - val_loss: 1.7928 - val_accuracy: 0.3681
Epoch 31/150

782/782 [=====] - 10s 13ms/step - loss: 1.7971 -
 accuracy: 0.3643 - val_loss: 1.7854 - val_accuracy: 0.3689
 Epoch 32/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7887 -
 accuracy: 0.3658 - val_loss: 1.7802 - val_accuracy: 0.3720
 Epoch 33/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7805 -
 accuracy: 0.3690 - val_loss: 1.7699 - val_accuracy: 0.3774
 Epoch 34/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7725 -
 accuracy: 0.3721 - val_loss: 1.7621 - val_accuracy: 0.3758
 Epoch 35/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7647 -
 accuracy: 0.3737 - val_loss: 1.7560 - val_accuracy: 0.3807
 Epoch 36/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7575 -
 accuracy: 0.3770 - val_loss: 1.7475 - val_accuracy: 0.3835
 Epoch 37/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7500 -
 accuracy: 0.3789 - val_loss: 1.7404 - val_accuracy: 0.3852
 Epoch 38/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7429 -
 accuracy: 0.3826 - val_loss: 1.7354 - val_accuracy: 0.3886
 Epoch 39/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7362 -
 accuracy: 0.3850 - val_loss: 1.7274 - val_accuracy: 0.3886
 Epoch 40/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7291 -
 accuracy: 0.3861 - val_loss: 1.7210 - val_accuracy: 0.3898
 Epoch 41/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7222 -
 accuracy: 0.3892 - val_loss: 1.7139 - val_accuracy: 0.3958
 Epoch 42/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7154 -
 accuracy: 0.3914 - val_loss: 1.7066 - val_accuracy: 0.3953
 Epoch 43/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7086 -
 accuracy: 0.3940 - val_loss: 1.7009 - val_accuracy: 0.3952
 Epoch 44/150
 782/782 [=====] - 10s 13ms/step - loss: 1.7020 -
 accuracy: 0.3976 - val_loss: 1.6940 - val_accuracy: 0.3979
 Epoch 45/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6953 -
 accuracy: 0.3995 - val_loss: 1.6876 - val_accuracy: 0.4010
 Epoch 46/150
 782/782 [=====] - 11s 13ms/step - loss: 1.6884 -
 accuracy: 0.4021 - val_loss: 1.6804 - val_accuracy: 0.4042
 Epoch 47/150

782/782 [=====] - 10s 13ms/step - loss: 1.6813 -
 accuracy: 0.4045 - val_loss: 1.6745 - val_accuracy: 0.4060
 Epoch 48/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6742 -
 accuracy: 0.4081 - val_loss: 1.6681 - val_accuracy: 0.4056
 Epoch 49/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6672 -
 accuracy: 0.4099 - val_loss: 1.6598 - val_accuracy: 0.4105
 Epoch 50/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6603 -
 accuracy: 0.4128 - val_loss: 1.6544 - val_accuracy: 0.4141
 Epoch 51/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6536 -
 accuracy: 0.4147 - val_loss: 1.6469 - val_accuracy: 0.4162
 Epoch 52/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6465 -
 accuracy: 0.4170 - val_loss: 1.6402 - val_accuracy: 0.4171
 Epoch 53/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6404 -
 accuracy: 0.4199 - val_loss: 1.6341 - val_accuracy: 0.4178
 Epoch 54/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6331 -
 accuracy: 0.4230 - val_loss: 1.6274 - val_accuracy: 0.4236
 Epoch 55/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6270 -
 accuracy: 0.4251 - val_loss: 1.6214 - val_accuracy: 0.4226
 Epoch 56/150
 782/782 [=====] - 11s 13ms/step - loss: 1.6207 -
 accuracy: 0.4280 - val_loss: 1.6150 - val_accuracy: 0.4260
 Epoch 57/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6143 -
 accuracy: 0.4306 - val_loss: 1.6106 - val_accuracy: 0.4287
 Epoch 58/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6088 -
 accuracy: 0.4318 - val_loss: 1.6045 - val_accuracy: 0.4301
 Epoch 59/150
 782/782 [=====] - 10s 13ms/step - loss: 1.6028 -
 accuracy: 0.4344 - val_loss: 1.5995 - val_accuracy: 0.4303
 Epoch 60/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5976 -
 accuracy: 0.4351 - val_loss: 1.5946 - val_accuracy: 0.4323
 Epoch 61/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5922 -
 accuracy: 0.4377 - val_loss: 1.5891 - val_accuracy: 0.4350
 Epoch 62/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5868 -
 accuracy: 0.4396 - val_loss: 1.5843 - val_accuracy: 0.4359
 Epoch 63/150

782/782 [=====] - 10s 13ms/step - loss: 1.5819 -
accuracy: 0.4427 - val_loss: 1.5801 - val_accuracy: 0.4371
Epoch 64/150
782/782 [=====] - 10s 13ms/step - loss: 1.5774 -
accuracy: 0.4453 - val_loss: 1.5756 - val_accuracy: 0.4362
Epoch 65/150
782/782 [=====] - 10s 13ms/step - loss: 1.5725 -
accuracy: 0.4453 - val_loss: 1.5709 - val_accuracy: 0.4421
Epoch 66/150
782/782 [=====] - 10s 13ms/step - loss: 1.5680 -
accuracy: 0.4479 - val_loss: 1.5669 - val_accuracy: 0.4449
Epoch 67/150
782/782 [=====] - 10s 13ms/step - loss: 1.5639 -
accuracy: 0.4499 - val_loss: 1.5637 - val_accuracy: 0.4458
Epoch 68/150
782/782 [=====] - 10s 13ms/step - loss: 1.5596 -
accuracy: 0.4506 - val_loss: 1.5594 - val_accuracy: 0.4465
Epoch 69/150
782/782 [=====] - 10s 13ms/step - loss: 1.5558 -
accuracy: 0.4532 - val_loss: 1.5551 - val_accuracy: 0.4504
Epoch 70/150
782/782 [=====] - 10s 13ms/step - loss: 1.5514 -
accuracy: 0.4538 - val_loss: 1.5518 - val_accuracy: 0.4497
Epoch 71/150
782/782 [=====] - 10s 13ms/step - loss: 1.5477 -
accuracy: 0.4551 - val_loss: 1.5483 - val_accuracy: 0.4495
Epoch 72/150
782/782 [=====] - 10s 13ms/step - loss: 1.5439 -
accuracy: 0.4565 - val_loss: 1.5446 - val_accuracy: 0.4494
Epoch 73/150
782/782 [=====] - 10s 13ms/step - loss: 1.5405 -
accuracy: 0.4578 - val_loss: 1.5420 - val_accuracy: 0.4509
Epoch 74/150
782/782 [=====] - 10s 13ms/step - loss: 1.5368 -
accuracy: 0.4581 - val_loss: 1.5384 - val_accuracy: 0.4562
Epoch 75/150
782/782 [=====] - 10s 13ms/step - loss: 1.5336 -
accuracy: 0.4591 - val_loss: 1.5369 - val_accuracy: 0.4547
Epoch 76/150
782/782 [=====] - 10s 13ms/step - loss: 1.5308 -
accuracy: 0.4606 - val_loss: 1.5329 - val_accuracy: 0.4557
Epoch 77/150
782/782 [=====] - 10s 13ms/step - loss: 1.5275 -
accuracy: 0.4607 - val_loss: 1.5307 - val_accuracy: 0.4582
Epoch 78/150
782/782 [=====] - 10s 13ms/step - loss: 1.5244 -
accuracy: 0.4615 - val_loss: 1.5272 - val_accuracy: 0.4600
Epoch 79/150

782/782 [=====] - 10s 13ms/step - loss: 1.5215 -
 accuracy: 0.4621 - val_loss: 1.5264 - val_accuracy: 0.4594
 Epoch 80/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5184 -
 accuracy: 0.4636 - val_loss: 1.5224 - val_accuracy: 0.4612
 Epoch 81/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5158 -
 accuracy: 0.4644 - val_loss: 1.5216 - val_accuracy: 0.4602
 Epoch 82/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5132 -
 accuracy: 0.4651 - val_loss: 1.5160 - val_accuracy: 0.4623
 Epoch 83/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5107 -
 accuracy: 0.4675 - val_loss: 1.5150 - val_accuracy: 0.4657
 Epoch 84/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5083 -
 accuracy: 0.4670 - val_loss: 1.5115 - val_accuracy: 0.4662
 Epoch 85/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5056 -
 accuracy: 0.4674 - val_loss: 1.5098 - val_accuracy: 0.4666
 Epoch 86/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5034 -
 accuracy: 0.4692 - val_loss: 1.5065 - val_accuracy: 0.4674
 Epoch 87/150
 782/782 [=====] - 10s 13ms/step - loss: 1.5010 -
 accuracy: 0.4687 - val_loss: 1.5050 - val_accuracy: 0.4659
 Epoch 88/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4989 -
 accuracy: 0.4700 - val_loss: 1.5030 - val_accuracy: 0.4695
 Epoch 89/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4965 -
 accuracy: 0.4710 - val_loss: 1.5010 - val_accuracy: 0.4693
 Epoch 90/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4944 -
 accuracy: 0.4720 - val_loss: 1.4997 - val_accuracy: 0.4701
 Epoch 91/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4923 -
 accuracy: 0.4717 - val_loss: 1.4967 - val_accuracy: 0.4702
 Epoch 92/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4904 -
 accuracy: 0.4736 - val_loss: 1.4952 - val_accuracy: 0.4746
 Epoch 93/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4883 -
 accuracy: 0.4738 - val_loss: 1.4933 - val_accuracy: 0.4738
 Epoch 94/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4862 -
 accuracy: 0.4744 - val_loss: 1.4908 - val_accuracy: 0.4735
 Epoch 95/150

782/782 [=====] - 10s 13ms/step - loss: 1.4844 -
 accuracy: 0.4748 - val_loss: 1.4891 - val_accuracy: 0.4762
 Epoch 96/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4826 -
 accuracy: 0.4749 - val_loss: 1.4870 - val_accuracy: 0.4779
 Epoch 97/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4809 -
 accuracy: 0.4745 - val_loss: 1.4861 - val_accuracy: 0.4771
 Epoch 98/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4790 -
 accuracy: 0.4765 - val_loss: 1.4856 - val_accuracy: 0.4776
 Epoch 99/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4773 -
 accuracy: 0.4768 - val_loss: 1.4832 - val_accuracy: 0.4761
 Epoch 100/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4757 -
 accuracy: 0.4773 - val_loss: 1.4812 - val_accuracy: 0.4793
 Epoch 101/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4740 -
 accuracy: 0.4782 - val_loss: 1.4794 - val_accuracy: 0.4770
 Epoch 102/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4724 -
 accuracy: 0.4779 - val_loss: 1.4780 - val_accuracy: 0.4789
 Epoch 103/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4706 -
 accuracy: 0.4789 - val_loss: 1.4759 - val_accuracy: 0.4774
 Epoch 104/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4693 -
 accuracy: 0.4794 - val_loss: 1.4748 - val_accuracy: 0.4824
 Epoch 105/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4676 -
 accuracy: 0.4807 - val_loss: 1.4726 - val_accuracy: 0.4803
 Epoch 106/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4662 -
 accuracy: 0.4809 - val_loss: 1.4716 - val_accuracy: 0.4798
 Epoch 107/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4643 -
 accuracy: 0.4805 - val_loss: 1.4690 - val_accuracy: 0.4811
 Epoch 108/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4631 -
 accuracy: 0.4816 - val_loss: 1.4687 - val_accuracy: 0.4836
 Epoch 109/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4614 -
 accuracy: 0.4831 - val_loss: 1.4682 - val_accuracy: 0.4835
 Epoch 110/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4600 -
 accuracy: 0.4824 - val_loss: 1.4662 - val_accuracy: 0.4848
 Epoch 111/150

782/782 [=====] - 10s 13ms/step - loss: 1.4587 -
 accuracy: 0.4835 - val_loss: 1.4641 - val_accuracy: 0.4849
 Epoch 112/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4568 -
 accuracy: 0.4839 - val_loss: 1.4623 - val_accuracy: 0.4831
 Epoch 113/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4555 -
 accuracy: 0.4850 - val_loss: 1.4617 - val_accuracy: 0.4846
 Epoch 114/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4542 -
 accuracy: 0.4856 - val_loss: 1.4600 - val_accuracy: 0.4831
 Epoch 115/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4527 -
 accuracy: 0.4863 - val_loss: 1.4583 - val_accuracy: 0.4832
 Epoch 116/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4515 -
 accuracy: 0.4859 - val_loss: 1.4581 - val_accuracy: 0.4847
 Epoch 117/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4500 -
 accuracy: 0.4858 - val_loss: 1.4593 - val_accuracy: 0.4820
 Epoch 118/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4488 -
 accuracy: 0.4875 - val_loss: 1.4548 - val_accuracy: 0.4850
 Epoch 119/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4472 -
 accuracy: 0.4889 - val_loss: 1.4545 - val_accuracy: 0.4877
 Epoch 120/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4465 -
 accuracy: 0.4891 - val_loss: 1.4520 - val_accuracy: 0.4854
 Epoch 121/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4447 -
 accuracy: 0.4888 - val_loss: 1.4513 - val_accuracy: 0.4855
 Epoch 122/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4435 -
 accuracy: 0.4892 - val_loss: 1.4497 - val_accuracy: 0.4869
 Epoch 123/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4423 -
 accuracy: 0.4906 - val_loss: 1.4476 - val_accuracy: 0.4849
 Epoch 124/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4408 -
 accuracy: 0.4913 - val_loss: 1.4472 - val_accuracy: 0.4876
 Epoch 125/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4399 -
 accuracy: 0.4906 - val_loss: 1.4457 - val_accuracy: 0.4886
 Epoch 126/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4384 -
 accuracy: 0.4922 - val_loss: 1.4442 - val_accuracy: 0.4904
 Epoch 127/150

782/782 [=====] - 10s 13ms/step - loss: 1.4370 -
 accuracy: 0.4922 - val_loss: 1.4442 - val_accuracy: 0.4893
 Epoch 128/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4359 -
 accuracy: 0.4927 - val_loss: 1.4420 - val_accuracy: 0.4891
 Epoch 129/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4350 -
 accuracy: 0.4947 - val_loss: 1.4416 - val_accuracy: 0.4908
 Epoch 130/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4338 -
 accuracy: 0.4939 - val_loss: 1.4427 - val_accuracy: 0.4868
 Epoch 131/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4324 -
 accuracy: 0.4957 - val_loss: 1.4391 - val_accuracy: 0.4917
 Epoch 132/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4311 -
 accuracy: 0.4958 - val_loss: 1.4420 - val_accuracy: 0.4893
 Epoch 133/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4302 -
 accuracy: 0.4954 - val_loss: 1.4384 - val_accuracy: 0.4888
 Epoch 134/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4290 -
 accuracy: 0.4961 - val_loss: 1.4353 - val_accuracy: 0.4938
 Epoch 135/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4277 -
 accuracy: 0.4961 - val_loss: 1.4344 - val_accuracy: 0.4931
 Epoch 136/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4266 -
 accuracy: 0.4967 - val_loss: 1.4337 - val_accuracy: 0.4918
 Epoch 137/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4253 -
 accuracy: 0.4974 - val_loss: 1.4326 - val_accuracy: 0.4956
 Epoch 138/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4239 -
 accuracy: 0.4982 - val_loss: 1.4335 - val_accuracy: 0.4925
 Epoch 139/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4231 -
 accuracy: 0.4987 - val_loss: 1.4298 - val_accuracy: 0.4942
 Epoch 140/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4221 -
 accuracy: 0.4992 - val_loss: 1.4295 - val_accuracy: 0.4950
 Epoch 141/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4210 -
 accuracy: 0.4998 - val_loss: 1.4284 - val_accuracy: 0.4961
 Epoch 142/150
 782/782 [=====] - 10s 13ms/step - loss: 1.4195 -
 accuracy: 0.5002 - val_loss: 1.4308 - val_accuracy: 0.4938
 Epoch 143/150

```

782/782 [=====] - 10s 13ms/step - loss: 1.4185 -
accuracy: 0.5007 - val_loss: 1.4274 - val_accuracy: 0.4949
Epoch 144/150
782/782 [=====] - 10s 13ms/step - loss: 1.4177 -
accuracy: 0.5016 - val_loss: 1.4255 - val_accuracy: 0.4961
Epoch 145/150
782/782 [=====] - 10s 13ms/step - loss: 1.4164 -
accuracy: 0.5013 - val_loss: 1.4243 - val_accuracy: 0.4943
Epoch 146/150
782/782 [=====] - 10s 13ms/step - loss: 1.4159 -
accuracy: 0.5021 - val_loss: 1.4252 - val_accuracy: 0.4956
Epoch 147/150
782/782 [=====] - 10s 13ms/step - loss: 1.4146 -
accuracy: 0.5022 - val_loss: 1.4220 - val_accuracy: 0.4957
Epoch 148/150
782/782 [=====] - 10s 13ms/step - loss: 1.4131 -
accuracy: 0.5028 - val_loss: 1.4215 - val_accuracy: 0.4963
Epoch 149/150
782/782 [=====] - 10s 13ms/step - loss: 1.4125 -
accuracy: 0.5036 - val_loss: 1.4198 - val_accuracy: 0.4966
Epoch 150/150
782/782 [=====] - 10s 13ms/step - loss: 1.4113 -
accuracy: 0.5043 - val_loss: 1.4192 - val_accuracy: 0.4969
time: 1541.3710536956787

```

Model liczy mniej więcej 30 min.

Trochę mniej jak nie korzystam z komputera, trochę więcej jeśli np. słucham wykładu.

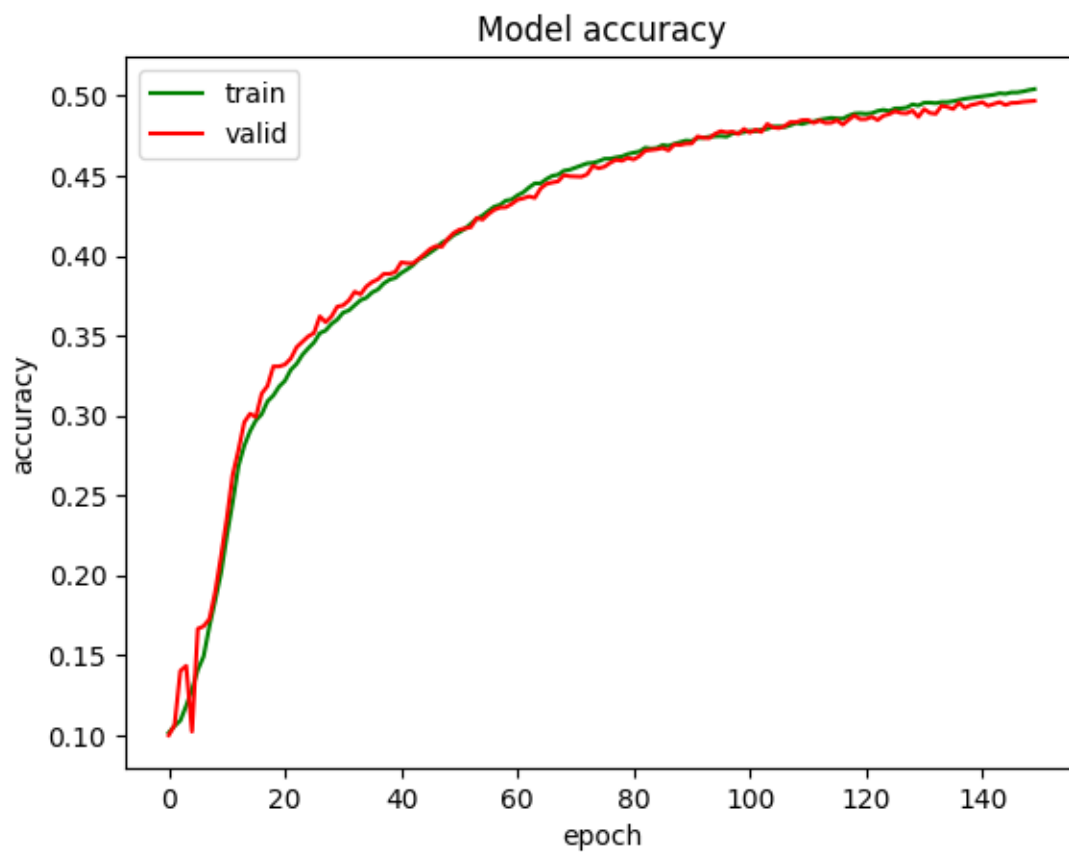
```
[ ]: minhistory.history.keys()
```

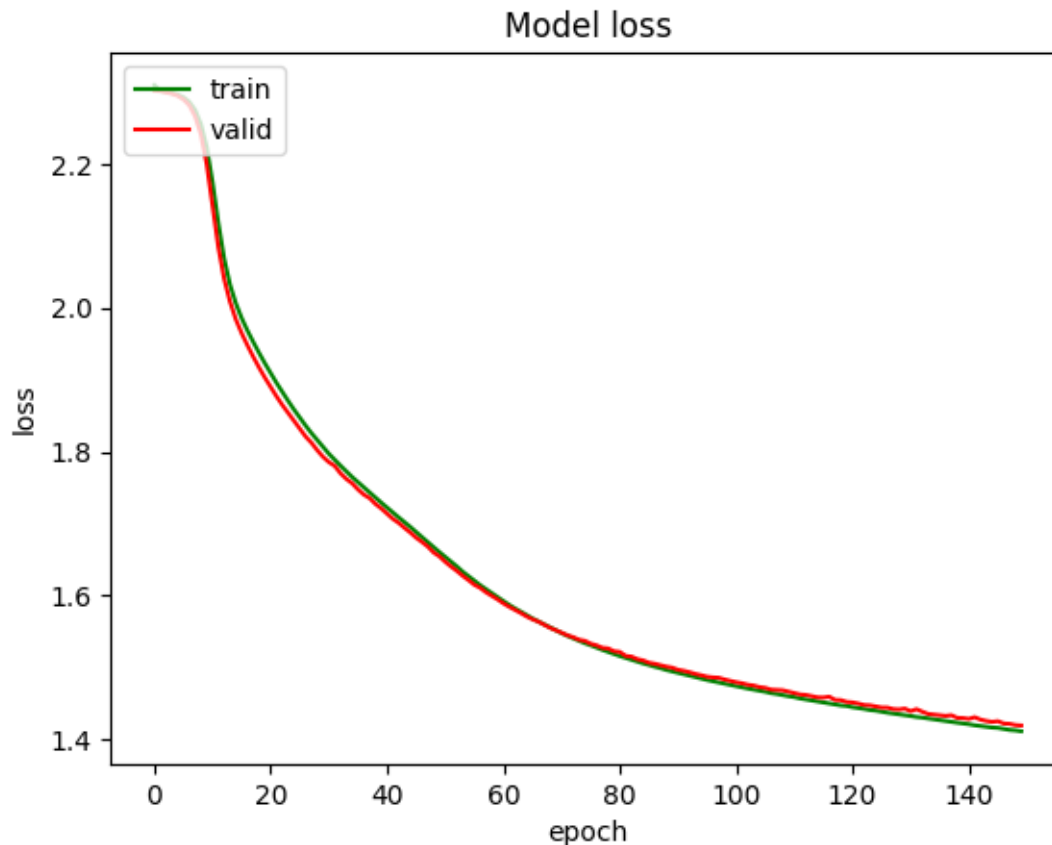
```
[ ]: dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy'])
```

```
[ ]: def print_train_and_val(history, gtype):
    plt.plot(history[gtype], 'g')
    plt.plot(history['val_'+gtype], 'r')
    plt.title('Model '+gtype)
    plt.ylabel(gtype)
    plt.xlabel('epoch')
    plt.legend(['train', 'valid'], loc='upper left')
    plt.show()

def print_history(history):
    print_train_and_val(history, 'accuracy')
    print_train_and_val(history, 'loss')

print_history(minhistory.history)
```





```
[ ]: first10 = minmodel(train_images[:10])
print(tf.math.reduce_all(tf.one_hot(tf.math.argmax(first10, axis=1),
↳ depth=10)==train_labels[:10], 1), '\n', tf.math.argmax(first10, axis=1))
print(tf.math.reduce_max(first10, axis=1))
```

```
tf.Tensor([False True True False True False False True True False],
shape=(10,), dtype=bool)
tf.Tensor([7 9 9 6 1 7 7 8 7], shape=(10,), dtype=int64)
tf.Tensor(
[0.28862807 0.52612484 0.197822 0.41355667 0.37701195 0.28073245
0.50594556 0.3302498 0.7759927 0.25490594], shape=(10,), dtype=float32)
```

Jak widać, po uczeniu, nasz model znacznie lepiej rozpoznaje klasy, choć dalej nieperfekcyjnie i wciąż możemy zaobserwować pewne biasy co do niektórych klas.

Widać, że często przy pomyłkach, pewność modelu est stosunkowo niska.

```
[ ]: minmodel2 = tf.keras.Sequential([
    tf.keras.layers.Conv2D(20, 3, padding='same', activation='sigmoid',
↳ input_shape=train_images[0].shape),
    tf.keras.layers.Conv2D(20, 3, padding='same', activation='sigmoid'),
```

```

tf.keras.layers.MaxPool2D(pool_size=(8,8), padding='valid'),
tf.keras.layers.Flatten(),
tf.keras.layers.Dense(units=10, activation='softmax')
])

```

```
[ ]: minmodel2.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #
conv2d_2 (Conv2D)	(None, 32, 32, 20)	560
conv2d_3 (Conv2D)	(None, 32, 32, 20)	3620
max_pooling2d_1 (MaxPooling 2D)	(None, 4, 4, 20)	0
flatten_1 (Flatten)	(None, 320)	0
dense_1 (Dense)	(None, 10)	3210

=====
 Total params: 7,390
 Trainable params: 7,390
 Non-trainable params: 0
 =====

Tym razem mamy 7,390 parametrów, czyli mniej więcej 7 razy więcej niż poprzednim razem.

```

[ ]: minmodel2.compile(
    optimizer=tf.keras.optimizers.SGD(learning_rate=0.001, momentum=0.9),
    loss=tf.keras.losses.CategoricalCrossentropy(),
    metrics=['accuracy']
)

```

```

[ ]: start_time = time.time()
history = minmodel2.fit(
    x=train_images,
    y=train_labels,
    batch_size=64,
    epochs=150,
    verbose=1,
    validation_data=(val_images, val_labels)
)
print(f"time: {time.time()-start_time}")

```

Epoch 1/150

```

2022-10-27 12:41:41.915278: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - ETA: 0s - loss: 2.3068 - accuracy:
0.1018

2022-10-27 12:41:54.545722: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - 14s 17ms/step - loss: 2.3068 -
accuracy: 0.1018 - val_loss: 2.3013 - val_accuracy: 0.1490
Epoch 2/150
782/782 [=====] - 13s 16ms/step - loss: 2.3016 -
accuracy: 0.1101 - val_loss: 2.2999 - val_accuracy: 0.1167
Epoch 3/150
782/782 [=====] - 13s 16ms/step - loss: 2.2967 -
accuracy: 0.1194 - val_loss: 2.2991 - val_accuracy: 0.1000
Epoch 4/150
782/782 [=====] - 13s 16ms/step - loss: 2.2886 -
accuracy: 0.1386 - val_loss: 2.2833 - val_accuracy: 0.1334
Epoch 5/150
782/782 [=====] - 13s 16ms/step - loss: 2.2717 -
accuracy: 0.1657 - val_loss: 2.2554 - val_accuracy: 0.2324
Epoch 6/150
782/782 [=====] - 13s 16ms/step - loss: 2.2398 -
accuracy: 0.1991 - val_loss: 2.2133 - val_accuracy: 0.2430
Epoch 7/150
782/782 [=====] - 13s 16ms/step - loss: 2.1872 -
accuracy: 0.2312 - val_loss: 2.1525 - val_accuracy: 0.2518
Epoch 8/150
782/782 [=====] - 13s 16ms/step - loss: 2.1253 -
accuracy: 0.2524 - val_loss: 2.0901 - val_accuracy: 0.2585
Epoch 9/150
782/782 [=====] - 13s 16ms/step - loss: 2.0687 -
accuracy: 0.2763 - val_loss: 2.0337 - val_accuracy: 0.2911
Epoch 10/150
782/782 [=====] - 13s 17ms/step - loss: 2.0236 -
accuracy: 0.2894 - val_loss: 1.9916 - val_accuracy: 0.3034
Epoch 11/150
782/782 [=====] - 13s 16ms/step - loss: 1.9854 -
accuracy: 0.3038 - val_loss: 1.9599 - val_accuracy: 0.3199
Epoch 12/150
782/782 [=====] - 13s 16ms/step - loss: 1.9551 -
accuracy: 0.3136 - val_loss: 1.9291 - val_accuracy: 0.3252
Epoch 13/150
782/782 [=====] - 13s 16ms/step - loss: 1.9285 -
accuracy: 0.3221 - val_loss: 1.9061 - val_accuracy: 0.3350

```

Epoch 14/150
782/782 [=====] - 13s 16ms/step - loss: 1.9058 -
accuracy: 0.3285 - val_loss: 1.8836 - val_accuracy: 0.3395
Epoch 15/150
782/782 [=====] - 13s 16ms/step - loss: 1.8854 -
accuracy: 0.3364 - val_loss: 1.8644 - val_accuracy: 0.3460
Epoch 16/150
782/782 [=====] - 13s 17ms/step - loss: 1.8667 -
accuracy: 0.3413 - val_loss: 1.8464 - val_accuracy: 0.3529
Epoch 17/150
782/782 [=====] - 13s 16ms/step - loss: 1.8502 -
accuracy: 0.3468 - val_loss: 1.8279 - val_accuracy: 0.3571
Epoch 18/150
782/782 [=====] - 13s 17ms/step - loss: 1.8346 -
accuracy: 0.3521 - val_loss: 1.8153 - val_accuracy: 0.3610
Epoch 19/150
782/782 [=====] - 13s 17ms/step - loss: 1.8191 -
accuracy: 0.3576 - val_loss: 1.8016 - val_accuracy: 0.3597
Epoch 20/150
782/782 [=====] - 13s 16ms/step - loss: 1.8054 -
accuracy: 0.3616 - val_loss: 1.7890 - val_accuracy: 0.3717
Epoch 21/150
782/782 [=====] - 13s 16ms/step - loss: 1.7915 -
accuracy: 0.3670 - val_loss: 1.7780 - val_accuracy: 0.3713
Epoch 22/150
782/782 [=====] - 13s 16ms/step - loss: 1.7785 -
accuracy: 0.3719 - val_loss: 1.7637 - val_accuracy: 0.3766
Epoch 23/150
782/782 [=====] - 13s 17ms/step - loss: 1.7659 -
accuracy: 0.3759 - val_loss: 1.7492 - val_accuracy: 0.3855
Epoch 24/150
782/782 [=====] - 13s 17ms/step - loss: 1.7529 -
accuracy: 0.3793 - val_loss: 1.7429 - val_accuracy: 0.3838
Epoch 25/150
782/782 [=====] - 13s 17ms/step - loss: 1.7408 -
accuracy: 0.3846 - val_loss: 1.7330 - val_accuracy: 0.3899
Epoch 26/150
782/782 [=====] - 13s 17ms/step - loss: 1.7296 -
accuracy: 0.3892 - val_loss: 1.7152 - val_accuracy: 0.3953
Epoch 27/150
782/782 [=====] - 13s 16ms/step - loss: 1.7185 -
accuracy: 0.3943 - val_loss: 1.7065 - val_accuracy: 0.4009
Epoch 28/150
782/782 [=====] - 13s 16ms/step - loss: 1.7074 -
accuracy: 0.3974 - val_loss: 1.6950 - val_accuracy: 0.4015
Epoch 29/150
782/782 [=====] - 13s 16ms/step - loss: 1.6967 -
accuracy: 0.4024 - val_loss: 1.6842 - val_accuracy: 0.4109

Epoch 30/150
782/782 [=====] - 13s 16ms/step - loss: 1.6877 - accuracy: 0.4064 - val_loss: 1.6761 - val_accuracy: 0.4149

Epoch 31/150
782/782 [=====] - 13s 16ms/step - loss: 1.6779 - accuracy: 0.4095 - val_loss: 1.6672 - val_accuracy: 0.4190

Epoch 32/150
782/782 [=====] - 13s 16ms/step - loss: 1.6682 - accuracy: 0.4135 - val_loss: 1.6573 - val_accuracy: 0.4227

Epoch 33/150
782/782 [=====] - 13s 16ms/step - loss: 1.6599 - accuracy: 0.4166 - val_loss: 1.6501 - val_accuracy: 0.4220

Epoch 34/150
782/782 [=====] - 13s 16ms/step - loss: 1.6517 - accuracy: 0.4190 - val_loss: 1.6411 - val_accuracy: 0.4256

Epoch 35/150
782/782 [=====] - 13s 17ms/step - loss: 1.6438 - accuracy: 0.4209 - val_loss: 1.6353 - val_accuracy: 0.4279

Epoch 36/150
782/782 [=====] - 13s 16ms/step - loss: 1.6356 - accuracy: 0.4259 - val_loss: 1.6258 - val_accuracy: 0.4313

Epoch 37/150
782/782 [=====] - 13s 16ms/step - loss: 1.6280 - accuracy: 0.4285 - val_loss: 1.6189 - val_accuracy: 0.4331

Epoch 38/150
782/782 [=====] - 13s 16ms/step - loss: 1.6213 - accuracy: 0.4316 - val_loss: 1.6109 - val_accuracy: 0.4364

Epoch 39/150
782/782 [=====] - 13s 16ms/step - loss: 1.6132 - accuracy: 0.4346 - val_loss: 1.6064 - val_accuracy: 0.4376

Epoch 40/150
782/782 [=====] - 13s 16ms/step - loss: 1.6065 - accuracy: 0.4364 - val_loss: 1.5961 - val_accuracy: 0.4429

Epoch 41/150
782/782 [=====] - 13s 16ms/step - loss: 1.5988 - accuracy: 0.4387 - val_loss: 1.5919 - val_accuracy: 0.4402

Epoch 42/150
782/782 [=====] - 13s 16ms/step - loss: 1.5920 - accuracy: 0.4417 - val_loss: 1.5831 - val_accuracy: 0.4439

Epoch 43/150
782/782 [=====] - 13s 16ms/step - loss: 1.5854 - accuracy: 0.4451 - val_loss: 1.5801 - val_accuracy: 0.4419

Epoch 44/150
782/782 [=====] - 13s 16ms/step - loss: 1.5788 - accuracy: 0.4458 - val_loss: 1.5701 - val_accuracy: 0.4465

Epoch 45/150
782/782 [=====] - 13s 16ms/step - loss: 1.5725 - accuracy: 0.4477 - val_loss: 1.5636 - val_accuracy: 0.4482

Epoch 46/150
782/782 [=====] - 13s 16ms/step - loss: 1.5659 - accuracy: 0.4495 - val_loss: 1.5564 - val_accuracy: 0.4541
Epoch 47/150
782/782 [=====] - 13s 16ms/step - loss: 1.5599 - accuracy: 0.4517 - val_loss: 1.5533 - val_accuracy: 0.4549
Epoch 48/150
782/782 [=====] - 13s 16ms/step - loss: 1.5540 - accuracy: 0.4528 - val_loss: 1.5483 - val_accuracy: 0.4550
Epoch 49/150
782/782 [=====] - 13s 17ms/step - loss: 1.5481 - accuracy: 0.4550 - val_loss: 1.5396 - val_accuracy: 0.4556
Epoch 50/150
782/782 [=====] - 13s 16ms/step - loss: 1.5422 - accuracy: 0.4574 - val_loss: 1.5334 - val_accuracy: 0.4595
Epoch 51/150
782/782 [=====] - 13s 16ms/step - loss: 1.5368 - accuracy: 0.4584 - val_loss: 1.5263 - val_accuracy: 0.4627
Epoch 52/150
782/782 [=====] - 13s 16ms/step - loss: 1.5315 - accuracy: 0.4598 - val_loss: 1.5259 - val_accuracy: 0.4636
Epoch 53/150
782/782 [=====] - 13s 17ms/step - loss: 1.5263 - accuracy: 0.4626 - val_loss: 1.5186 - val_accuracy: 0.4663
Epoch 54/150
782/782 [=====] - 13s 16ms/step - loss: 1.5206 - accuracy: 0.4640 - val_loss: 1.5112 - val_accuracy: 0.4676
Epoch 55/150
782/782 [=====] - 13s 16ms/step - loss: 1.5154 - accuracy: 0.4653 - val_loss: 1.5103 - val_accuracy: 0.4678
Epoch 56/150
782/782 [=====] - 13s 16ms/step - loss: 1.5107 - accuracy: 0.4669 - val_loss: 1.5034 - val_accuracy: 0.4714
Epoch 57/150
782/782 [=====] - 13s 17ms/step - loss: 1.5059 - accuracy: 0.4680 - val_loss: 1.5007 - val_accuracy: 0.4711
Epoch 58/150
782/782 [=====] - 13s 17ms/step - loss: 1.5005 - accuracy: 0.4707 - val_loss: 1.4930 - val_accuracy: 0.4749
Epoch 59/150
782/782 [=====] - 13s 16ms/step - loss: 1.4959 - accuracy: 0.4712 - val_loss: 1.4922 - val_accuracy: 0.4725
Epoch 60/150
782/782 [=====] - 13s 16ms/step - loss: 1.4917 - accuracy: 0.4723 - val_loss: 1.4834 - val_accuracy: 0.4777
Epoch 61/150
782/782 [=====] - 13s 16ms/step - loss: 1.4867 - accuracy: 0.4738 - val_loss: 1.4805 - val_accuracy: 0.4740

Epoch 62/150
782/782 [=====] - 13s 16ms/step - loss: 1.4823 - accuracy: 0.4754 - val_loss: 1.4736 - val_accuracy: 0.4809
Epoch 63/150
782/782 [=====] - 13s 16ms/step - loss: 1.4780 - accuracy: 0.4778 - val_loss: 1.4717 - val_accuracy: 0.4820
Epoch 64/150
782/782 [=====] - 13s 16ms/step - loss: 1.4727 - accuracy: 0.4796 - val_loss: 1.4659 - val_accuracy: 0.4828
Epoch 65/150
782/782 [=====] - 13s 16ms/step - loss: 1.4679 - accuracy: 0.4812 - val_loss: 1.4613 - val_accuracy: 0.4831
Epoch 66/150
782/782 [=====] - 13s 16ms/step - loss: 1.4637 - accuracy: 0.4822 - val_loss: 1.4549 - val_accuracy: 0.4880
Epoch 67/150
782/782 [=====] - 13s 16ms/step - loss: 1.4593 - accuracy: 0.4843 - val_loss: 1.4527 - val_accuracy: 0.4870
Epoch 68/150
782/782 [=====] - 13s 16ms/step - loss: 1.4545 - accuracy: 0.4864 - val_loss: 1.4479 - val_accuracy: 0.4899
Epoch 69/150
782/782 [=====] - 13s 16ms/step - loss: 1.4502 - accuracy: 0.4884 - val_loss: 1.4435 - val_accuracy: 0.4929
Epoch 70/150
782/782 [=====] - 13s 16ms/step - loss: 1.4459 - accuracy: 0.4884 - val_loss: 1.4398 - val_accuracy: 0.4951
Epoch 71/150
782/782 [=====] - 13s 16ms/step - loss: 1.4411 - accuracy: 0.4900 - val_loss: 1.4373 - val_accuracy: 0.4939
Epoch 72/150
782/782 [=====] - 13s 16ms/step - loss: 1.4365 - accuracy: 0.4928 - val_loss: 1.4329 - val_accuracy: 0.4945
Epoch 73/150
782/782 [=====] - 13s 16ms/step - loss: 1.4321 - accuracy: 0.4934 - val_loss: 1.4290 - val_accuracy: 0.4996
Epoch 74/150
782/782 [=====] - 13s 16ms/step - loss: 1.4277 - accuracy: 0.4952 - val_loss: 1.4297 - val_accuracy: 0.4940
Epoch 75/150
782/782 [=====] - 13s 16ms/step - loss: 1.4235 - accuracy: 0.4964 - val_loss: 1.4194 - val_accuracy: 0.5005
Epoch 76/150
782/782 [=====] - 13s 16ms/step - loss: 1.4190 - accuracy: 0.4980 - val_loss: 1.4185 - val_accuracy: 0.5000
Epoch 77/150
782/782 [=====] - 13s 16ms/step - loss: 1.4149 - accuracy: 0.5000 - val_loss: 1.4126 - val_accuracy: 0.5021

Epoch 78/150
782/782 [=====] - 13s 16ms/step - loss: 1.4111 -
accuracy: 0.5023 - val_loss: 1.4084 - val_accuracy: 0.5036
Epoch 79/150
782/782 [=====] - 13s 17ms/step - loss: 1.4064 -
accuracy: 0.5030 - val_loss: 1.4047 - val_accuracy: 0.5065
Epoch 80/150
782/782 [=====] - 13s 16ms/step - loss: 1.4017 -
accuracy: 0.5047 - val_loss: 1.3982 - val_accuracy: 0.5075
Epoch 81/150
782/782 [=====] - 13s 16ms/step - loss: 1.3973 -
accuracy: 0.5066 - val_loss: 1.3970 - val_accuracy: 0.5074
Epoch 82/150
782/782 [=====] - 13s 16ms/step - loss: 1.3933 -
accuracy: 0.5091 - val_loss: 1.3928 - val_accuracy: 0.5076
Epoch 83/150
782/782 [=====] - 13s 16ms/step - loss: 1.3892 -
accuracy: 0.5091 - val_loss: 1.3877 - val_accuracy: 0.5112
Epoch 84/150
782/782 [=====] - 13s 16ms/step - loss: 1.3841 -
accuracy: 0.5120 - val_loss: 1.3825 - val_accuracy: 0.5131
Epoch 85/150
782/782 [=====] - 13s 17ms/step - loss: 1.3799 -
accuracy: 0.5127 - val_loss: 1.3815 - val_accuracy: 0.5137
Epoch 86/150
782/782 [=====] - 13s 16ms/step - loss: 1.3760 -
accuracy: 0.5141 - val_loss: 1.3761 - val_accuracy: 0.5136
Epoch 87/150
782/782 [=====] - 13s 16ms/step - loss: 1.3718 -
accuracy: 0.5152 - val_loss: 1.3722 - val_accuracy: 0.5167
Epoch 88/150
782/782 [=====] - 13s 16ms/step - loss: 1.3677 -
accuracy: 0.5192 - val_loss: 1.3702 - val_accuracy: 0.5169
Epoch 89/150
782/782 [=====] - 13s 17ms/step - loss: 1.3641 -
accuracy: 0.5191 - val_loss: 1.3649 - val_accuracy: 0.5192
Epoch 90/150
782/782 [=====] - 13s 17ms/step - loss: 1.3598 -
accuracy: 0.5206 - val_loss: 1.3653 - val_accuracy: 0.5188
Epoch 91/150
782/782 [=====] - 13s 16ms/step - loss: 1.3560 -
accuracy: 0.5223 - val_loss: 1.3556 - val_accuracy: 0.5231
Epoch 92/150
782/782 [=====] - 13s 16ms/step - loss: 1.3521 -
accuracy: 0.5237 - val_loss: 1.3521 - val_accuracy: 0.5226
Epoch 93/150
782/782 [=====] - 13s 17ms/step - loss: 1.3483 -
accuracy: 0.5239 - val_loss: 1.3495 - val_accuracy: 0.5254

Epoch 94/150
782/782 [=====] - 13s 17ms/step - loss: 1.3443 - accuracy: 0.5262 - val_loss: 1.3448 - val_accuracy: 0.5299
Epoch 95/150
782/782 [=====] - 13s 16ms/step - loss: 1.3402 - accuracy: 0.5290 - val_loss: 1.3423 - val_accuracy: 0.5299
Epoch 96/150
782/782 [=====] - 13s 16ms/step - loss: 1.3370 - accuracy: 0.5282 - val_loss: 1.3425 - val_accuracy: 0.5277
Epoch 97/150
782/782 [=====] - 13s 17ms/step - loss: 1.3335 - accuracy: 0.5297 - val_loss: 1.3376 - val_accuracy: 0.5277
Epoch 98/150
782/782 [=====] - 13s 17ms/step - loss: 1.3298 - accuracy: 0.5300 - val_loss: 1.3313 - val_accuracy: 0.5319
Epoch 99/150
782/782 [=====] - 13s 16ms/step - loss: 1.3261 - accuracy: 0.5330 - val_loss: 1.3287 - val_accuracy: 0.5311
Epoch 100/150
782/782 [=====] - 13s 16ms/step - loss: 1.3226 - accuracy: 0.5327 - val_loss: 1.3286 - val_accuracy: 0.5322
Epoch 101/150
782/782 [=====] - 13s 16ms/step - loss: 1.3195 - accuracy: 0.5352 - val_loss: 1.3223 - val_accuracy: 0.5362
Epoch 102/150
782/782 [=====] - 13s 17ms/step - loss: 1.3160 - accuracy: 0.5361 - val_loss: 1.3196 - val_accuracy: 0.5356
Epoch 103/150
782/782 [=====] - 13s 16ms/step - loss: 1.3130 - accuracy: 0.5379 - val_loss: 1.3158 - val_accuracy: 0.5361
Epoch 104/150
782/782 [=====] - 13s 17ms/step - loss: 1.3097 - accuracy: 0.5391 - val_loss: 1.3220 - val_accuracy: 0.5309
Epoch 105/150
782/782 [=====] - 13s 16ms/step - loss: 1.3064 - accuracy: 0.5403 - val_loss: 1.3109 - val_accuracy: 0.5354
Epoch 106/150
782/782 [=====] - 13s 16ms/step - loss: 1.3034 - accuracy: 0.5413 - val_loss: 1.3098 - val_accuracy: 0.5384
Epoch 107/150
782/782 [=====] - 13s 16ms/step - loss: 1.3002 - accuracy: 0.5431 - val_loss: 1.3065 - val_accuracy: 0.5391
Epoch 108/150
782/782 [=====] - 13s 16ms/step - loss: 1.2974 - accuracy: 0.5431 - val_loss: 1.3052 - val_accuracy: 0.5394
Epoch 109/150
782/782 [=====] - 13s 17ms/step - loss: 1.2942 - accuracy: 0.5459 - val_loss: 1.3003 - val_accuracy: 0.5421

Epoch 110/150
782/782 [=====] - 13s 16ms/step - loss: 1.2920 - accuracy: 0.5456 - val_loss: 1.3023 - val_accuracy: 0.5414

Epoch 111/150
782/782 [=====] - 13s 16ms/step - loss: 1.2890 - accuracy: 0.5473 - val_loss: 1.2975 - val_accuracy: 0.5403

Epoch 112/150
782/782 [=====] - 13s 16ms/step - loss: 1.2859 - accuracy: 0.5471 - val_loss: 1.3005 - val_accuracy: 0.5448

Epoch 113/150
782/782 [=====] - 13s 17ms/step - loss: 1.2833 - accuracy: 0.5482 - val_loss: 1.2893 - val_accuracy: 0.5464

Epoch 114/150
782/782 [=====] - 13s 16ms/step - loss: 1.2810 - accuracy: 0.5505 - val_loss: 1.2868 - val_accuracy: 0.5452

Epoch 115/150
782/782 [=====] - 13s 16ms/step - loss: 1.2781 - accuracy: 0.5500 - val_loss: 1.2853 - val_accuracy: 0.5460

Epoch 116/150
782/782 [=====] - 13s 16ms/step - loss: 1.2759 - accuracy: 0.5518 - val_loss: 1.2839 - val_accuracy: 0.5475

Epoch 117/150
782/782 [=====] - 13s 16ms/step - loss: 1.2733 - accuracy: 0.5525 - val_loss: 1.2832 - val_accuracy: 0.5484

Epoch 118/150
782/782 [=====] - 13s 16ms/step - loss: 1.2704 - accuracy: 0.5554 - val_loss: 1.2774 - val_accuracy: 0.5525

Epoch 119/150
782/782 [=====] - 13s 16ms/step - loss: 1.2686 - accuracy: 0.5543 - val_loss: 1.2779 - val_accuracy: 0.5498

Epoch 120/150
782/782 [=====] - 13s 17ms/step - loss: 1.2660 - accuracy: 0.5544 - val_loss: 1.2758 - val_accuracy: 0.5540

Epoch 121/150
782/782 [=====] - 13s 17ms/step - loss: 1.2635 - accuracy: 0.5564 - val_loss: 1.2727 - val_accuracy: 0.5555

Epoch 122/150
782/782 [=====] - 13s 17ms/step - loss: 1.2612 - accuracy: 0.5564 - val_loss: 1.2711 - val_accuracy: 0.5496

Epoch 123/150
782/782 [=====] - 13s 16ms/step - loss: 1.2586 - accuracy: 0.5587 - val_loss: 1.2684 - val_accuracy: 0.5539

Epoch 124/150
782/782 [=====] - 13s 17ms/step - loss: 1.2563 - accuracy: 0.5592 - val_loss: 1.2678 - val_accuracy: 0.5556

Epoch 125/150
782/782 [=====] - 13s 16ms/step - loss: 1.2546 - accuracy: 0.5594 - val_loss: 1.2653 - val_accuracy: 0.5545

Epoch 126/150
782/782 [=====] - 13s 17ms/step - loss: 1.2528 - accuracy: 0.5602 - val_loss: 1.2600 - val_accuracy: 0.5570

Epoch 127/150
782/782 [=====] - 13s 17ms/step - loss: 1.2498 - accuracy: 0.5610 - val_loss: 1.2617 - val_accuracy: 0.5549

Epoch 128/150
782/782 [=====] - 13s 17ms/step - loss: 1.2479 - accuracy: 0.5615 - val_loss: 1.2578 - val_accuracy: 0.5584

Epoch 129/150
782/782 [=====] - 13s 16ms/step - loss: 1.2463 - accuracy: 0.5627 - val_loss: 1.2602 - val_accuracy: 0.5559

Epoch 130/150
782/782 [=====] - 13s 16ms/step - loss: 1.2434 - accuracy: 0.5631 - val_loss: 1.2550 - val_accuracy: 0.5595

Epoch 131/150
782/782 [=====] - 13s 17ms/step - loss: 1.2413 - accuracy: 0.5649 - val_loss: 1.2530 - val_accuracy: 0.5604

Epoch 132/150
782/782 [=====] - 13s 17ms/step - loss: 1.2390 - accuracy: 0.5644 - val_loss: 1.2490 - val_accuracy: 0.5625

Epoch 133/150
782/782 [=====] - 13s 17ms/step - loss: 1.2369 - accuracy: 0.5677 - val_loss: 1.2503 - val_accuracy: 0.5612

Epoch 134/150
782/782 [=====] - 13s 16ms/step - loss: 1.2347 - accuracy: 0.5650 - val_loss: 1.2489 - val_accuracy: 0.5596

Epoch 135/150
782/782 [=====] - 13s 16ms/step - loss: 1.2328 - accuracy: 0.5675 - val_loss: 1.2462 - val_accuracy: 0.5632

Epoch 136/150
782/782 [=====] - 13s 17ms/step - loss: 1.2301 - accuracy: 0.5692 - val_loss: 1.2432 - val_accuracy: 0.5638

Epoch 137/150
782/782 [=====] - 13s 16ms/step - loss: 1.2285 - accuracy: 0.5683 - val_loss: 1.2419 - val_accuracy: 0.5651

Epoch 138/150
782/782 [=====] - 13s 16ms/step - loss: 1.2265 - accuracy: 0.5691 - val_loss: 1.2418 - val_accuracy: 0.5636

Epoch 139/150
782/782 [=====] - 13s 17ms/step - loss: 1.2249 - accuracy: 0.5699 - val_loss: 1.2386 - val_accuracy: 0.5660

Epoch 140/150
782/782 [=====] - 13s 17ms/step - loss: 1.2221 - accuracy: 0.5715 - val_loss: 1.2357 - val_accuracy: 0.5661

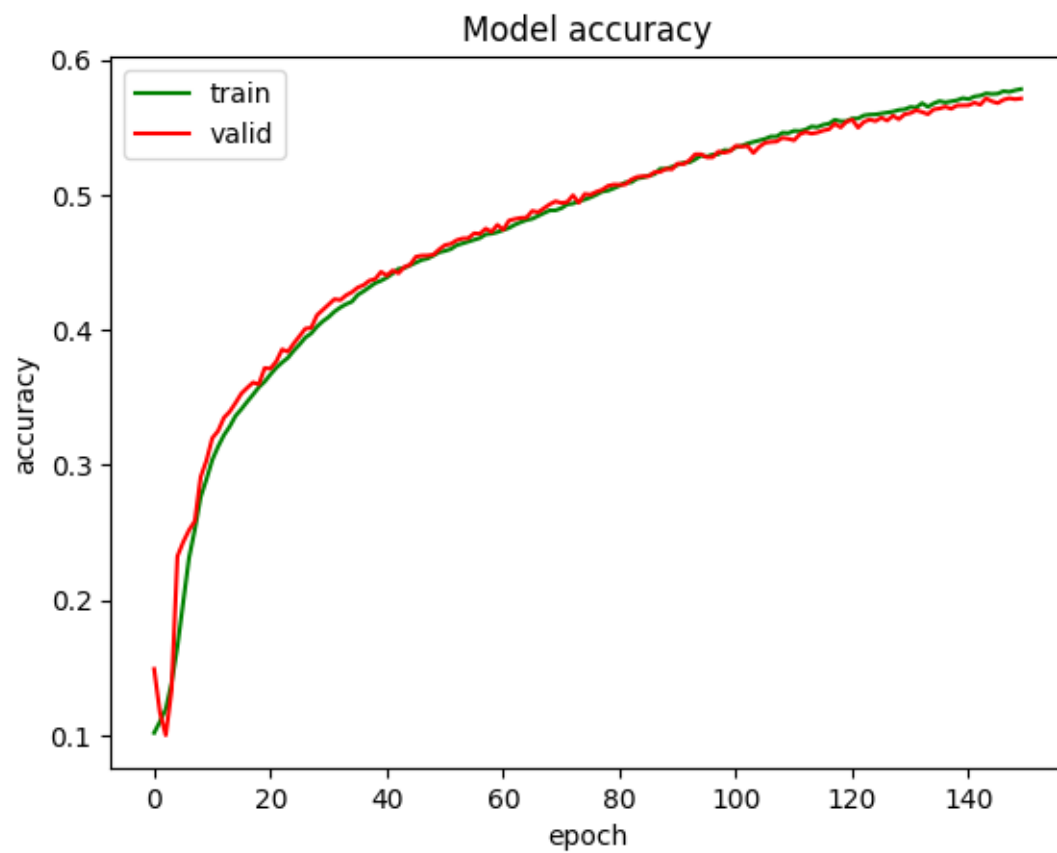
Epoch 141/150
782/782 [=====] - 13s 17ms/step - loss: 1.2207 - accuracy: 0.5707 - val_loss: 1.2345 - val_accuracy: 0.5662

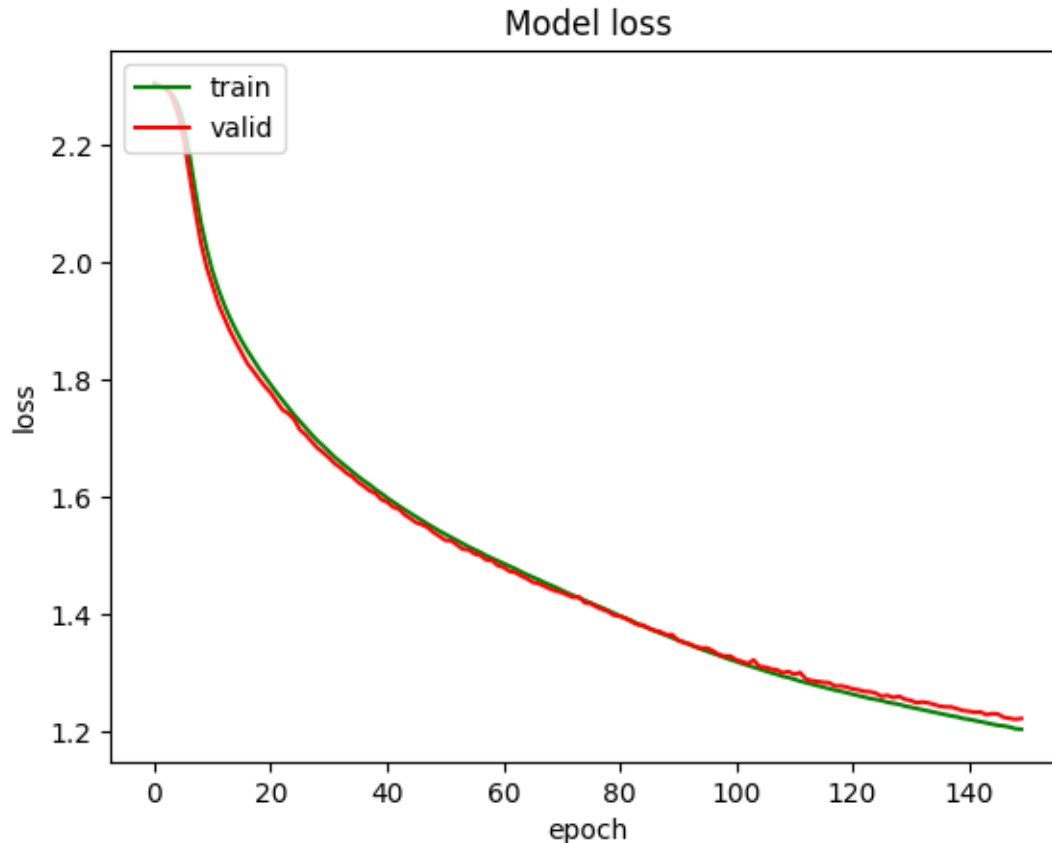
```

Epoch 142/150
782/782 [=====] - 13s 17ms/step - loss: 1.2188 -
accuracy: 0.5727 - val_loss: 1.2328 - val_accuracy: 0.5681
Epoch 143/150
782/782 [=====] - 13s 16ms/step - loss: 1.2168 -
accuracy: 0.5733 - val_loss: 1.2333 - val_accuracy: 0.5666
Epoch 144/150
782/782 [=====] - 13s 17ms/step - loss: 1.2146 -
accuracy: 0.5750 - val_loss: 1.2285 - val_accuracy: 0.5713
Epoch 145/150
782/782 [=====] - 13s 16ms/step - loss: 1.2124 -
accuracy: 0.5745 - val_loss: 1.2305 - val_accuracy: 0.5691
Epoch 146/150
782/782 [=====] - 13s 16ms/step - loss: 1.2106 -
accuracy: 0.5748 - val_loss: 1.2297 - val_accuracy: 0.5678
Epoch 147/150
782/782 [=====] - 13s 16ms/step - loss: 1.2094 -
accuracy: 0.5766 - val_loss: 1.2241 - val_accuracy: 0.5703
Epoch 148/150
782/782 [=====] - 13s 16ms/step - loss: 1.2075 -
accuracy: 0.5761 - val_loss: 1.2224 - val_accuracy: 0.5714
Epoch 149/150
782/782 [=====] - 13s 16ms/step - loss: 1.2046 -
accuracy: 0.5773 - val_loss: 1.2209 - val_accuracy: 0.5707
Epoch 150/150
782/782 [=====] - 13s 16ms/step - loss: 1.2038 -
accuracy: 0.5782 - val_loss: 1.2221 - val_accuracy: 0.5712
time: 1935.4221241474152

```

```
[ ]: print_history(history.history)
```





Czas trenowania wzrósł nieznacznie, jedynie mniej więcej 3 sekundy na każdej epoce, w sumie ok. 400 sekund.

Rezultaty wzrosły o prawie 10 punktów procentowych.

Jestem zdziwiony że tak drastyczna zmiana liczby parametrów (ok 7 razy więcej), tak nieznacznie wpłynęła na czas trwania epoki.

```
[ ]: def block(channels, activation='relu', input_shape=None,
    ↪ batch_normalisation=False, dropout=None, gap=False):
    if input_shape:
        conv1 = tf.keras.layers.Conv2D(channels, 3, padding='same',
    ↪ activation=activation, input_shape=input_shape)
    else:
        conv1 = tf.keras.layers.Conv2D(channels, 3, padding='same',
    ↪ activation=activation)
        conv2 = tf.keras.layers.Conv2D(channels, 3, padding='same',
    ↪ activation=activation)
    if batch_normalisation:
        batch_norm1 = tf.keras.layers.BatchNormalization()
        batch_norm2 = tf.keras.layers.BatchNormalization()
    if dropout:
```

```

        drops = tf.keras.layers.Dropout(dropout)
    if gap:
        pool = tf.keras.layers.
↪GlobalAveragePooling2D(data_format='channels_last', keepdims=False)
    else:
        pool = tf.keras.layers.MaxPool2D(pool_size=(2,2), padding='valid')

    if dropout:
        return [conv1, batch_norm1, conv2, batch_norm2, pool, drops]
    elif batch_normalisation:
        return [conv1, batch_norm1, conv2, batch_norm2, pool]
    else:
        return [conv1, conv2, pool]

```

Przygotowałem funkcję którą przygotowuje nam taki blok, zwraca go jako listę, więc mogę go dodać do innego bloku, bądź listy innych warst i tf.keras.Sequential nie będzie mieć problemów z formatem

```

[ ]: model1 = tf.keras.Sequential(
    block(channels=20, activation='sigmoid', input_shape=train_images[0].shape)↵
    ↪+
    block(channels=40, activation='sigmoid') + [
        tf.keras.layers.Flatten(),
        tf.keras.layers.Dense(units=10,activation='softmax')
    ])

```

```

[ ]: model1.summary()

```

Model: "sequential_8"

Layer (type)	Output Shape	Param #
=====		
conv2d_31 (Conv2D)	(None, 32, 32, 20)	560
conv2d_32 (Conv2D)	(None, 32, 32, 20)	3620
max_pooling2d_15 (MaxPoolin g2D)	(None, 16, 16, 20)	0
conv2d_33 (Conv2D)	(None, 16, 16, 40)	7240
conv2d_34 (Conv2D)	(None, 16, 16, 40)	14440
max_pooling2d_16 (MaxPoolin g2D)	(None, 8, 8, 40)	0
flatten_8 (Flatten)	(None, 2560)	0
dense_8 (Dense)	(None, 10)	25610

```
=====
Total params: 51,470
Trainable params: 51,470
Non-trainable params: 0
-----
```

```
[ ]: model1.compile(
    optimizer=tf.keras.optimizers.SGD(learning_rate=0.001, momentum=0.9),
    loss=tf.keras.losses.CategoricalCrossentropy(),
    metrics=['accuracy']
)
```

```
[ ]: history1 = model1.fit(
    x=train_images,
    y=train_labels,
    batch_size=64,
    epochs=15,
    verbose=1,
    validation_data=(val_images, val_labels)
)
```

Epoch 1/15

```
2022-10-27 17:02:56.479463: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - ETA: 0s - loss: 2.3263 - accuracy:
0.0988
```

```
2022-10-27 17:03:06.321068: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.
```

```
782/782 [=====] - 11s 13ms/step - loss: 2.3263 -
accuracy: 0.0988 - val_loss: 2.3235 - val_accuracy: 0.1000
Epoch 2/15
```

```
782/782 [=====] - 9s 12ms/step - loss: 2.3207 -
accuracy: 0.1001 - val_loss: 2.3315 - val_accuracy: 0.1000
```

Epoch 3/15

```
782/782 [=====] - 10s 13ms/step - loss: 2.3157 -
accuracy: 0.1010 - val_loss: 2.3152 - val_accuracy: 0.1000
```

Epoch 4/15

```
782/782 [=====] - 10s 12ms/step - loss: 2.3130 -
accuracy: 0.0996 - val_loss: 2.3067 - val_accuracy: 0.1000
```

Epoch 5/15

```
782/782 [=====] - 10s 12ms/step - loss: 2.3097 -
accuracy: 0.1021 - val_loss: 2.3042 - val_accuracy: 0.1000
```

Epoch 6/15


```

782/782 [=====] - 9s 12ms/step - loss: 2.3084 -
accuracy: 0.1012 - val_loss: 2.3053 - val_accuracy: 0.1000
Epoch 7/15
782/782 [=====] - 9s 12ms/step - loss: 2.3074 -
accuracy: 0.1002 - val_loss: 2.3120 - val_accuracy: 0.1000
Epoch 8/15
782/782 [=====] - 10s 12ms/step - loss: 2.3075 -
accuracy: 0.0980 - val_loss: 2.3070 - val_accuracy: 0.1000
Epoch 9/15
782/782 [=====] - 10s 12ms/step - loss: 2.3071 -
accuracy: 0.0985 - val_loss: 2.3076 - val_accuracy: 0.1000
Epoch 10/15
782/782 [=====] - 10s 12ms/step - loss: 2.3057 -
accuracy: 0.1003 - val_loss: 2.3071 - val_accuracy: 0.1000
Epoch 11/15
782/782 [=====] - 10s 12ms/step - loss: 2.3058 -
accuracy: 0.0996 - val_loss: 2.3039 - val_accuracy: 0.1000
Epoch 12/15
782/782 [=====] - 10s 13ms/step - loss: 2.3052 -
accuracy: 0.1000 - val_loss: 2.3041 - val_accuracy: 0.1000
Epoch 13/15
782/782 [=====] - 10s 12ms/step - loss: 2.3048 -
accuracy: 0.1012 - val_loss: 2.3045 - val_accuracy: 0.1012
Epoch 14/15
782/782 [=====] - 10s 13ms/step - loss: 2.3047 -
accuracy: 0.1013 - val_loss: 2.3053 - val_accuracy: 0.1000
Epoch 15/15
782/782 [=====] - 10s 13ms/step - loss: 2.3046 -
accuracy: 0.1014 - val_loss: 2.3035 - val_accuracy: 0.1000

```

Wygląda to bardzo nieobiecująco, loss i accuracy praktycznie pozostały takie same...

```
[ ]: model2 = tf.keras.Sequential(
    block(channels=20, activation='relu', input_shape=train_images[0].shape) +
    block(channels=40, activation='relu') + [
        tf.keras.layers.Flatten(),
        tf.keras.layers.Dense(units=10,activation='softmax')
    ])

```

```
[ ]: model2.summary()
```

Model: "sequential_10"

Layer (type)	Output Shape	Param #
conv2d_39 (Conv2D)	(None, 32, 32, 20)	560
conv2d_40 (Conv2D)	(None, 32, 32, 20)	3620

max_pooling2d_19 (MaxPoolin g2D)	(None, 16, 16, 20)	0
conv2d_41 (Conv2D)	(None, 16, 16, 40)	7240
conv2d_42 (Conv2D)	(None, 16, 16, 40)	14440
max_pooling2d_20 (MaxPoolin g2D)	(None, 8, 8, 40)	0
flatten_10 (Flatten)	(None, 2560)	0
dense_10 (Dense)	(None, 10)	25610

```
=====
Total params: 51,470
Trainable params: 51,470
Non-trainable params: 0
-----
```

```
[ ]: model2.compile(
    optimizer=tf.keras.optimizers.SGD(learning_rate=0.001, momentum=0.9),
    loss=tf.keras.losses.CategoricalCrossentropy(),
    metrics=['accuracy']
)
```

```
[ ]: start_time = time.time()
hisotry2 = model2.fit(
    x=train_images,
    y=train_labels,
    batch_size=64,
    epochs=150,
    verbose=1,
    validation_data=(val_images, val_labels)
)
print(f"time: {time.time()-start_time}")
```

Epoch 1/150

```
2022-10-27 17:13:46.260185: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - ETA: 0s - loss: 2.1764 - accuracy:
0.1938

2022-10-27 17:13:55.343189: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.
```

782/782 [=====] - 10s 13ms/step - loss: 2.1764 -
accuracy: 0.1938 - val_loss: 1.9325 - val_accuracy: 0.3098
Epoch 2/150
782/782 [=====] - 10s 13ms/step - loss: 1.7968 -
accuracy: 0.3655 - val_loss: 1.6535 - val_accuracy: 0.4148
Epoch 3/150
782/782 [=====] - 10s 12ms/step - loss: 1.5625 -
accuracy: 0.4446 - val_loss: 1.4597 - val_accuracy: 0.4846
Epoch 4/150
782/782 [=====] - 10s 12ms/step - loss: 1.4455 -
accuracy: 0.4854 - val_loss: 1.4012 - val_accuracy: 0.5023
Epoch 5/150
782/782 [=====] - 9s 12ms/step - loss: 1.3731 -
accuracy: 0.5122 - val_loss: 1.3722 - val_accuracy: 0.5125
Epoch 6/150
782/782 [=====] - 9s 12ms/step - loss: 1.3032 -
accuracy: 0.5382 - val_loss: 1.2813 - val_accuracy: 0.5449
Epoch 7/150
782/782 [=====] - 10s 12ms/step - loss: 1.2455 -
accuracy: 0.5638 - val_loss: 1.2332 - val_accuracy: 0.5619
Epoch 8/150
782/782 [=====] - 9s 12ms/step - loss: 1.1921 -
accuracy: 0.5786 - val_loss: 1.2683 - val_accuracy: 0.5587
Epoch 9/150
782/782 [=====] - 10s 12ms/step - loss: 1.1488 -
accuracy: 0.5985 - val_loss: 1.1512 - val_accuracy: 0.5962
Epoch 10/150
782/782 [=====] - 10s 12ms/step - loss: 1.1064 -
accuracy: 0.6111 - val_loss: 1.1262 - val_accuracy: 0.6082
Epoch 11/150
782/782 [=====] - 10s 12ms/step - loss: 1.0716 -
accuracy: 0.6248 - val_loss: 1.1510 - val_accuracy: 0.5892
Epoch 12/150
782/782 [=====] - 10s 12ms/step - loss: 1.0385 -
accuracy: 0.6370 - val_loss: 1.0850 - val_accuracy: 0.6189
Epoch 13/150
782/782 [=====] - 10s 12ms/step - loss: 1.0063 -
accuracy: 0.6488 - val_loss: 1.0686 - val_accuracy: 0.6244
Epoch 14/150
782/782 [=====] - 10s 12ms/step - loss: 0.9788 -
accuracy: 0.6585 - val_loss: 1.0482 - val_accuracy: 0.6316
Epoch 15/150
782/782 [=====] - 10s 12ms/step - loss: 0.9537 -
accuracy: 0.6678 - val_loss: 1.0364 - val_accuracy: 0.6413
Epoch 16/150
782/782 [=====] - 9s 12ms/step - loss: 0.9288 -
accuracy: 0.6770 - val_loss: 1.0184 - val_accuracy: 0.6470
Epoch 17/150

782/782 [=====] - 10s 12ms/step - loss: 0.9060 -
accuracy: 0.6853 - val_loss: 1.0047 - val_accuracy: 0.6495
Epoch 18/150
782/782 [=====] - 10s 12ms/step - loss: 0.8855 -
accuracy: 0.6929 - val_loss: 0.9988 - val_accuracy: 0.6589
Epoch 19/150
782/782 [=====] - 10s 12ms/step - loss: 0.8636 -
accuracy: 0.7002 - val_loss: 1.0003 - val_accuracy: 0.6566
Epoch 20/150
782/782 [=====] - 10s 12ms/step - loss: 0.8451 -
accuracy: 0.7076 - val_loss: 0.9741 - val_accuracy: 0.6643
Epoch 21/150
782/782 [=====] - 10s 12ms/step - loss: 0.8313 -
accuracy: 0.7126 - val_loss: 0.9621 - val_accuracy: 0.6729
Epoch 22/150
782/782 [=====] - 10s 12ms/step - loss: 0.8128 -
accuracy: 0.7192 - val_loss: 0.9445 - val_accuracy: 0.6762
Epoch 23/150
782/782 [=====] - 10s 12ms/step - loss: 0.7955 -
accuracy: 0.7265 - val_loss: 0.9531 - val_accuracy: 0.6747
Epoch 24/150
782/782 [=====] - 10s 12ms/step - loss: 0.7841 -
accuracy: 0.7277 - val_loss: 0.9334 - val_accuracy: 0.6811
Epoch 25/150
782/782 [=====] - 10s 12ms/step - loss: 0.7691 -
accuracy: 0.7345 - val_loss: 0.9512 - val_accuracy: 0.6756
Epoch 26/150
782/782 [=====] - 10s 12ms/step - loss: 0.7510 -
accuracy: 0.7413 - val_loss: 0.9635 - val_accuracy: 0.6761
Epoch 27/150
782/782 [=====] - 10s 12ms/step - loss: 0.7396 -
accuracy: 0.7452 - val_loss: 0.9370 - val_accuracy: 0.6861
Epoch 28/150
782/782 [=====] - 10s 12ms/step - loss: 0.7218 -
accuracy: 0.7495 - val_loss: 0.9486 - val_accuracy: 0.6801
Epoch 29/150
782/782 [=====] - 10s 12ms/step - loss: 0.7153 -
accuracy: 0.7534 - val_loss: 0.9369 - val_accuracy: 0.6869
Epoch 30/150
782/782 [=====] - 10s 12ms/step - loss: 0.7011 -
accuracy: 0.7576 - val_loss: 0.9384 - val_accuracy: 0.6906
Epoch 31/150
782/782 [=====] - 10s 12ms/step - loss: 0.6892 -
accuracy: 0.7620 - val_loss: 0.9391 - val_accuracy: 0.6857
Epoch 32/150
782/782 [=====] - 10s 12ms/step - loss: 0.6764 -
accuracy: 0.7668 - val_loss: 0.9538 - val_accuracy: 0.6800
Epoch 33/150

782/782 [=====] - 10s 12ms/step - loss: 0.6639 -
accuracy: 0.7726 - val_loss: 0.9433 - val_accuracy: 0.6840
Epoch 34/150
782/782 [=====] - 10s 12ms/step - loss: 0.6527 -
accuracy: 0.7746 - val_loss: 0.9399 - val_accuracy: 0.6898
Epoch 35/150
782/782 [=====] - 9s 12ms/step - loss: 0.6407 -
accuracy: 0.7788 - val_loss: 0.9632 - val_accuracy: 0.6848
Epoch 36/150
782/782 [=====] - 10s 12ms/step - loss: 0.6333 -
accuracy: 0.7812 - val_loss: 0.9992 - val_accuracy: 0.6737
Epoch 37/150
782/782 [=====] - 10s 12ms/step - loss: 0.6201 -
accuracy: 0.7868 - val_loss: 0.9371 - val_accuracy: 0.6916
Epoch 38/150
782/782 [=====] - 9s 12ms/step - loss: 0.6112 -
accuracy: 0.7881 - val_loss: 0.9672 - val_accuracy: 0.6895
Epoch 39/150
782/782 [=====] - 10s 12ms/step - loss: 0.6048 -
accuracy: 0.7897 - val_loss: 0.9757 - val_accuracy: 0.6864
Epoch 40/150
782/782 [=====] - 10s 12ms/step - loss: 0.5957 -
accuracy: 0.7932 - val_loss: 0.9826 - val_accuracy: 0.6825
Epoch 41/150
782/782 [=====] - 10s 12ms/step - loss: 0.5833 -
accuracy: 0.7987 - val_loss: 0.9721 - val_accuracy: 0.6924
Epoch 42/150
782/782 [=====] - 10s 12ms/step - loss: 0.5716 -
accuracy: 0.8036 - val_loss: 0.9630 - val_accuracy: 0.6880
Epoch 43/150
782/782 [=====] - 10s 12ms/step - loss: 0.5658 -
accuracy: 0.8036 - val_loss: 0.9894 - val_accuracy: 0.6881
Epoch 44/150
782/782 [=====] - 10s 12ms/step - loss: 0.5566 -
accuracy: 0.8071 - val_loss: 0.9798 - val_accuracy: 0.6850
Epoch 45/150
782/782 [=====] - 10s 12ms/step - loss: 0.5504 -
accuracy: 0.8085 - val_loss: 0.9742 - val_accuracy: 0.6910
Epoch 46/150
782/782 [=====] - 10s 12ms/step - loss: 0.5433 -
accuracy: 0.8106 - val_loss: 0.9940 - val_accuracy: 0.6882
Epoch 47/150
782/782 [=====] - 10s 12ms/step - loss: 0.5310 -
accuracy: 0.8146 - val_loss: 0.9870 - val_accuracy: 0.6944
Epoch 48/150
782/782 [=====] - 10s 12ms/step - loss: 0.5266 -
accuracy: 0.8162 - val_loss: 1.0158 - val_accuracy: 0.6895
Epoch 49/150

782/782 [=====] - 10s 12ms/step - loss: 0.5158 -
 accuracy: 0.8206 - val_loss: 1.0042 - val_accuracy: 0.6898
 Epoch 50/150
 782/782 [=====] - 10s 12ms/step - loss: 0.5108 -
 accuracy: 0.8222 - val_loss: 1.0017 - val_accuracy: 0.6887
 Epoch 51/150
 782/782 [=====] - 9s 12ms/step - loss: 0.5006 -
 accuracy: 0.8250 - val_loss: 1.0551 - val_accuracy: 0.6870
 Epoch 52/150
 782/782 [=====] - 9s 12ms/step - loss: 0.4942 -
 accuracy: 0.8276 - val_loss: 1.0436 - val_accuracy: 0.6843
 Epoch 53/150
 782/782 [=====] - 9s 12ms/step - loss: 0.4864 -
 accuracy: 0.8302 - val_loss: 1.0286 - val_accuracy: 0.6918
 Epoch 54/150
 782/782 [=====] - 9s 12ms/step - loss: 0.4809 -
 accuracy: 0.8316 - val_loss: 1.0505 - val_accuracy: 0.6904
 Epoch 55/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4708 -
 accuracy: 0.8369 - val_loss: 1.0623 - val_accuracy: 0.6878
 Epoch 56/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4604 -
 accuracy: 0.8404 - val_loss: 1.0736 - val_accuracy: 0.6832
 Epoch 57/150
 782/782 [=====] - 9s 12ms/step - loss: 0.4544 -
 accuracy: 0.8403 - val_loss: 1.1103 - val_accuracy: 0.6791
 Epoch 58/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4500 -
 accuracy: 0.8428 - val_loss: 1.1045 - val_accuracy: 0.6860
 Epoch 59/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4458 -
 accuracy: 0.8448 - val_loss: 1.0834 - val_accuracy: 0.6929
 Epoch 60/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4376 -
 accuracy: 0.8465 - val_loss: 1.1157 - val_accuracy: 0.6878
 Epoch 61/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4300 -
 accuracy: 0.8495 - val_loss: 1.1512 - val_accuracy: 0.6812
 Epoch 62/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4247 -
 accuracy: 0.8524 - val_loss: 1.1393 - val_accuracy: 0.6874
 Epoch 63/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4198 -
 accuracy: 0.8530 - val_loss: 1.1576 - val_accuracy: 0.6819
 Epoch 64/150
 782/782 [=====] - 10s 12ms/step - loss: 0.4122 -
 accuracy: 0.8545 - val_loss: 1.1698 - val_accuracy: 0.6803
 Epoch 65/150

782/782 [=====] - 10s 12ms/step - loss: 0.4047 -
accuracy: 0.8587 - val_loss: 1.1707 - val_accuracy: 0.6874
Epoch 66/150
782/782 [=====] - 10s 12ms/step - loss: 0.3996 -
accuracy: 0.8597 - val_loss: 1.1836 - val_accuracy: 0.6845
Epoch 67/150
782/782 [=====] - 10s 12ms/step - loss: 0.3985 -
accuracy: 0.8603 - val_loss: 1.1773 - val_accuracy: 0.6837
Epoch 68/150
782/782 [=====] - 9s 12ms/step - loss: 0.3886 -
accuracy: 0.8632 - val_loss: 1.2086 - val_accuracy: 0.6817
Epoch 69/150
782/782 [=====] - 10s 12ms/step - loss: 0.3824 -
accuracy: 0.8644 - val_loss: 1.2417 - val_accuracy: 0.6820
Epoch 70/150
782/782 [=====] - 9s 12ms/step - loss: 0.3733 -
accuracy: 0.8687 - val_loss: 1.2322 - val_accuracy: 0.6818
Epoch 71/150
782/782 [=====] - 10s 12ms/step - loss: 0.3715 -
accuracy: 0.8676 - val_loss: 1.2627 - val_accuracy: 0.6793
Epoch 72/150
782/782 [=====] - 9s 12ms/step - loss: 0.3619 -
accuracy: 0.8726 - val_loss: 1.2703 - val_accuracy: 0.6812
Epoch 73/150
782/782 [=====] - 9s 12ms/step - loss: 0.3561 -
accuracy: 0.8743 - val_loss: 1.2887 - val_accuracy: 0.6712
Epoch 74/150
782/782 [=====] - 9s 12ms/step - loss: 0.3532 -
accuracy: 0.8752 - val_loss: 1.3037 - val_accuracy: 0.6816
Epoch 75/150
782/782 [=====] - 9s 12ms/step - loss: 0.3472 -
accuracy: 0.8783 - val_loss: 1.3359 - val_accuracy: 0.6774
Epoch 76/150
782/782 [=====] - 9s 12ms/step - loss: 0.3441 -
accuracy: 0.8777 - val_loss: 1.3303 - val_accuracy: 0.6792
Epoch 77/150
782/782 [=====] - 9s 12ms/step - loss: 0.3349 -
accuracy: 0.8833 - val_loss: 1.3360 - val_accuracy: 0.6819
Epoch 78/150
782/782 [=====] - 9s 12ms/step - loss: 0.3274 -
accuracy: 0.8852 - val_loss: 1.3708 - val_accuracy: 0.6785
Epoch 79/150
782/782 [=====] - 9s 12ms/step - loss: 0.3272 -
accuracy: 0.8845 - val_loss: 1.3963 - val_accuracy: 0.6803
Epoch 80/150
782/782 [=====] - 9s 12ms/step - loss: 0.3242 -
accuracy: 0.8840 - val_loss: 1.3869 - val_accuracy: 0.6743
Epoch 81/150

782/782 [=====] - 9s 12ms/step - loss: 0.3166 -
accuracy: 0.8877 - val_loss: 1.4146 - val_accuracy: 0.6766
Epoch 82/150
782/782 [=====] - 9s 12ms/step - loss: 0.3114 -
accuracy: 0.8894 - val_loss: 1.4435 - val_accuracy: 0.6740
Epoch 83/150
782/782 [=====] - 9s 12ms/step - loss: 0.3048 -
accuracy: 0.8912 - val_loss: 1.4482 - val_accuracy: 0.6743
Epoch 84/150
782/782 [=====] - 9s 12ms/step - loss: 0.3058 -
accuracy: 0.8915 - val_loss: 1.4765 - val_accuracy: 0.6692
Epoch 85/150
782/782 [=====] - 9s 12ms/step - loss: 0.2996 -
accuracy: 0.8938 - val_loss: 1.4991 - val_accuracy: 0.6730
Epoch 86/150
782/782 [=====] - 9s 12ms/step - loss: 0.2935 -
accuracy: 0.8957 - val_loss: 1.5373 - val_accuracy: 0.6742
Epoch 87/150
782/782 [=====] - 9s 12ms/step - loss: 0.2867 -
accuracy: 0.8986 - val_loss: 1.5678 - val_accuracy: 0.6674
Epoch 88/150
782/782 [=====] - 9s 12ms/step - loss: 0.2813 -
accuracy: 0.8999 - val_loss: 1.5570 - val_accuracy: 0.6769
Epoch 89/150
782/782 [=====] - 9s 12ms/step - loss: 0.2773 -
accuracy: 0.9019 - val_loss: 1.5878 - val_accuracy: 0.6675
Epoch 90/150
782/782 [=====] - 9s 12ms/step - loss: 0.2758 -
accuracy: 0.9016 - val_loss: 1.6030 - val_accuracy: 0.6732
Epoch 91/150
782/782 [=====] - 9s 12ms/step - loss: 0.2726 -
accuracy: 0.9027 - val_loss: 1.5743 - val_accuracy: 0.6740
Epoch 92/150
782/782 [=====] - 9s 12ms/step - loss: 0.2661 -
accuracy: 0.9040 - val_loss: 1.6248 - val_accuracy: 0.6733
Epoch 93/150
782/782 [=====] - 9s 12ms/step - loss: 0.2623 -
accuracy: 0.9060 - val_loss: 1.6553 - val_accuracy: 0.6669
Epoch 94/150
782/782 [=====] - 9s 12ms/step - loss: 0.2629 -
accuracy: 0.9040 - val_loss: 1.6822 - val_accuracy: 0.6721
Epoch 95/150
782/782 [=====] - 9s 12ms/step - loss: 0.2611 -
accuracy: 0.9065 - val_loss: 1.6824 - val_accuracy: 0.6715
Epoch 96/150
782/782 [=====] - 9s 12ms/step - loss: 0.2492 -
accuracy: 0.9100 - val_loss: 1.7615 - val_accuracy: 0.6631
Epoch 97/150

782/782 [=====] - 9s 12ms/step - loss: 0.2520 -
 accuracy: 0.9095 - val_loss: 1.7134 - val_accuracy: 0.6677
 Epoch 98/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2416 -
 accuracy: 0.9134 - val_loss: 1.8051 - val_accuracy: 0.6635
 Epoch 99/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2431 -
 accuracy: 0.9109 - val_loss: 1.8132 - val_accuracy: 0.6645
 Epoch 100/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2367 -
 accuracy: 0.9137 - val_loss: 1.7885 - val_accuracy: 0.6678
 Epoch 101/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2330 -
 accuracy: 0.9157 - val_loss: 1.8140 - val_accuracy: 0.6641
 Epoch 102/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2258 -
 accuracy: 0.9190 - val_loss: 1.8558 - val_accuracy: 0.6639
 Epoch 103/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2255 -
 accuracy: 0.9187 - val_loss: 1.8618 - val_accuracy: 0.6662
 Epoch 104/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2281 -
 accuracy: 0.9171 - val_loss: 1.9025 - val_accuracy: 0.6685
 Epoch 105/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2236 -
 accuracy: 0.9176 - val_loss: 1.9217 - val_accuracy: 0.6646
 Epoch 106/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2184 -
 accuracy: 0.9206 - val_loss: 1.9467 - val_accuracy: 0.6587
 Epoch 107/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2142 -
 accuracy: 0.9227 - val_loss: 1.9616 - val_accuracy: 0.6684
 Epoch 108/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2064 -
 accuracy: 0.9260 - val_loss: 1.9792 - val_accuracy: 0.6586
 Epoch 109/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2012 -
 accuracy: 0.9265 - val_loss: 2.0113 - val_accuracy: 0.6683
 Epoch 110/150
 782/782 [=====] - 9s 12ms/step - loss: 0.2017 -
 accuracy: 0.9265 - val_loss: 2.0679 - val_accuracy: 0.6605
 Epoch 111/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1994 -
 accuracy: 0.9265 - val_loss: 2.0906 - val_accuracy: 0.6599
 Epoch 112/150
 782/782 [=====] - 10s 13ms/step - loss: 0.1954 -
 accuracy: 0.9278 - val_loss: 2.1239 - val_accuracy: 0.6667
 Epoch 113/150

782/782 [=====] - 10s 12ms/step - loss: 0.1971 - accuracy: 0.9271 - val_loss: 2.1215 - val_accuracy: 0.6631
Epoch 114/150
782/782 [=====] - 10s 12ms/step - loss: 0.1878 - accuracy: 0.9310 - val_loss: 2.1457 - val_accuracy: 0.6598
Epoch 115/150
782/782 [=====] - 10s 12ms/step - loss: 0.1938 - accuracy: 0.9282 - val_loss: 2.1546 - val_accuracy: 0.6586
Epoch 116/150
782/782 [=====] - 10s 12ms/step - loss: 0.1942 - accuracy: 0.9281 - val_loss: 2.1509 - val_accuracy: 0.6614
Epoch 117/150
782/782 [=====] - 10s 12ms/step - loss: 0.1838 - accuracy: 0.9315 - val_loss: 2.1966 - val_accuracy: 0.6564
Epoch 118/150
782/782 [=====] - 10s 12ms/step - loss: 0.1811 - accuracy: 0.9332 - val_loss: 2.2334 - val_accuracy: 0.6620
Epoch 119/150
782/782 [=====] - 10s 12ms/step - loss: 0.1776 - accuracy: 0.9344 - val_loss: 2.2444 - val_accuracy: 0.6582
Epoch 120/150
782/782 [=====] - 10s 12ms/step - loss: 0.1771 - accuracy: 0.9337 - val_loss: 2.3007 - val_accuracy: 0.6556
Epoch 121/150
782/782 [=====] - 10s 12ms/step - loss: 0.1702 - accuracy: 0.9369 - val_loss: 2.3485 - val_accuracy: 0.6602
Epoch 122/150
782/782 [=====] - 10s 12ms/step - loss: 0.1685 - accuracy: 0.9385 - val_loss: 2.3378 - val_accuracy: 0.6612
Epoch 123/150
782/782 [=====] - 10s 12ms/step - loss: 0.1634 - accuracy: 0.9387 - val_loss: 2.4258 - val_accuracy: 0.6585
Epoch 124/150
782/782 [=====] - 10s 12ms/step - loss: 0.1700 - accuracy: 0.9370 - val_loss: 2.3814 - val_accuracy: 0.6587
Epoch 125/150
782/782 [=====] - 10s 12ms/step - loss: 0.1587 - accuracy: 0.9400 - val_loss: 2.4350 - val_accuracy: 0.6537
Epoch 126/150
782/782 [=====] - 10s 12ms/step - loss: 0.1583 - accuracy: 0.9417 - val_loss: 2.4299 - val_accuracy: 0.6617
Epoch 127/150
782/782 [=====] - 10s 12ms/step - loss: 0.1552 - accuracy: 0.9418 - val_loss: 2.5343 - val_accuracy: 0.6602
Epoch 128/150
782/782 [=====] - 10s 12ms/step - loss: 0.1554 - accuracy: 0.9427 - val_loss: 2.4799 - val_accuracy: 0.6639
Epoch 129/150

782/782 [=====] - 10s 12ms/step - loss: 0.1588 -
 accuracy: 0.9409 - val_loss: 2.4807 - val_accuracy: 0.6602
 Epoch 130/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1487 -
 accuracy: 0.9447 - val_loss: 2.5458 - val_accuracy: 0.6542
 Epoch 131/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1516 -
 accuracy: 0.9437 - val_loss: 2.5567 - val_accuracy: 0.6596
 Epoch 132/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1550 -
 accuracy: 0.9417 - val_loss: 2.6246 - val_accuracy: 0.6581
 Epoch 133/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1410 -
 accuracy: 0.9483 - val_loss: 2.6703 - val_accuracy: 0.6520
 Epoch 134/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1472 -
 accuracy: 0.9446 - val_loss: 2.6596 - val_accuracy: 0.6542
 Epoch 135/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1413 -
 accuracy: 0.9471 - val_loss: 2.6953 - val_accuracy: 0.6603
 Epoch 136/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1421 -
 accuracy: 0.9472 - val_loss: 2.7888 - val_accuracy: 0.6476
 Epoch 137/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1438 -
 accuracy: 0.9458 - val_loss: 2.7017 - val_accuracy: 0.6582
 Epoch 138/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1313 -
 accuracy: 0.9521 - val_loss: 2.7591 - val_accuracy: 0.6551
 Epoch 139/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1543 -
 accuracy: 0.9416 - val_loss: 2.7353 - val_accuracy: 0.6580
 Epoch 140/150
 782/782 [=====] - 10s 13ms/step - loss: 0.1401 -
 accuracy: 0.9481 - val_loss: 2.8006 - val_accuracy: 0.6576
 Epoch 141/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1364 -
 accuracy: 0.9482 - val_loss: 2.8311 - val_accuracy: 0.6556
 Epoch 142/150
 782/782 [=====] - 10s 13ms/step - loss: 0.1398 -
 accuracy: 0.9468 - val_loss: 2.8509 - val_accuracy: 0.6534
 Epoch 143/150
 782/782 [=====] - 10s 12ms/step - loss: 0.1270 -
 accuracy: 0.9536 - val_loss: 2.8740 - val_accuracy: 0.6567
 Epoch 144/150
 782/782 [=====] - 10s 13ms/step - loss: 0.1234 -
 accuracy: 0.9536 - val_loss: 2.9041 - val_accuracy: 0.6560
 Epoch 145/150

```

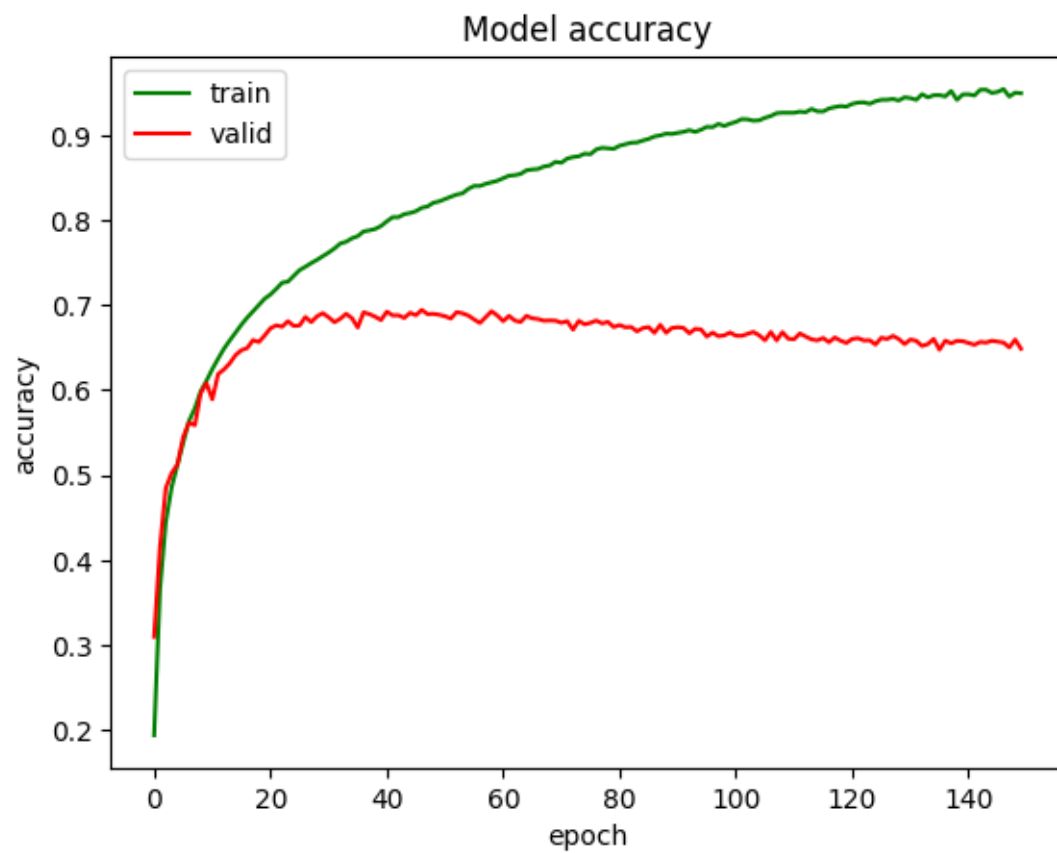
782/782 [=====] - 10s 12ms/step - loss: 0.1347 -
accuracy: 0.9498 - val_loss: 2.8561 - val_accuracy: 0.6581
Epoch 146/150
782/782 [=====] - 10s 12ms/step - loss: 0.1330 -
accuracy: 0.9511 - val_loss: 2.9528 - val_accuracy: 0.6571
Epoch 147/150
782/782 [=====] - 10s 12ms/step - loss: 0.1228 -
accuracy: 0.9544 - val_loss: 2.9438 - val_accuracy: 0.6560
Epoch 148/150
782/782 [=====] - 10s 12ms/step - loss: 0.1467 -
accuracy: 0.9455 - val_loss: 2.9854 - val_accuracy: 0.6503
Epoch 149/150
782/782 [=====] - 10s 12ms/step - loss: 0.1333 -
accuracy: 0.9501 - val_loss: 2.9757 - val_accuracy: 0.6595
Epoch 150/150
782/782 [=====] - 10s 12ms/step - loss: 0.1348 -
accuracy: 0.9494 - val_loss: 3.0778 - val_accuracy: 0.6488
time: 1426.2418642044067

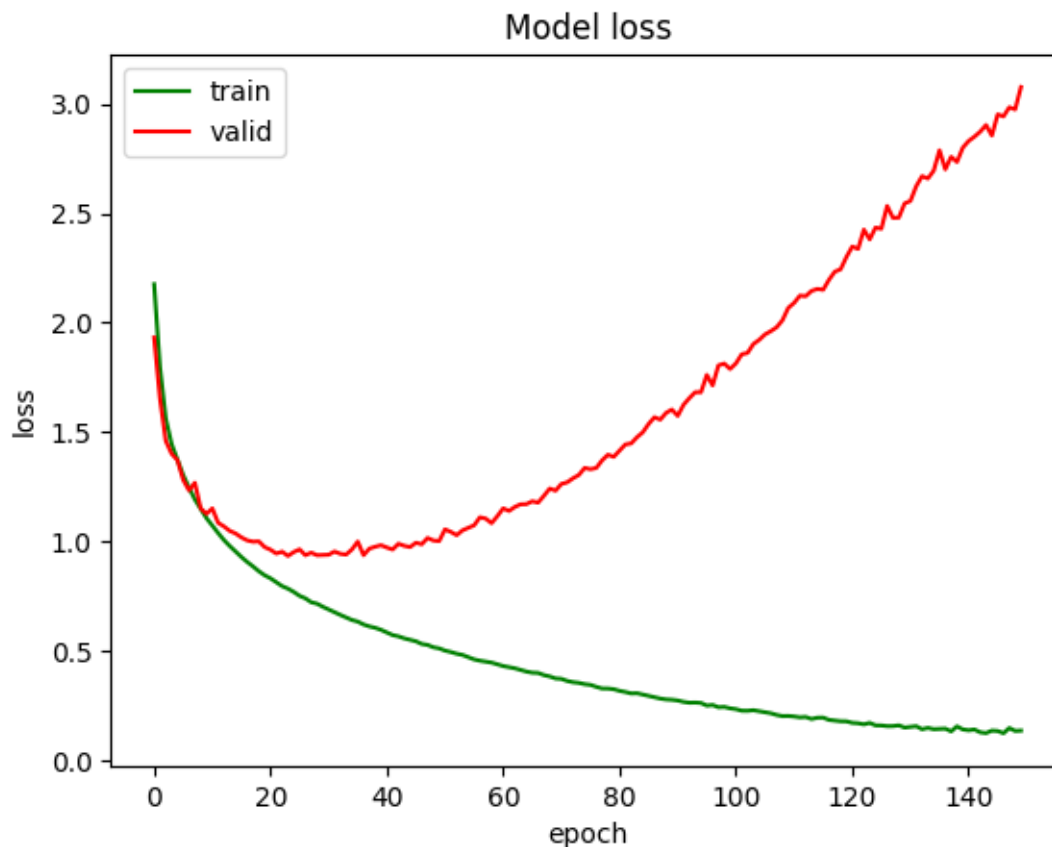
```

Model liczył się bardzo szybko, szybciej nawet niż nasz pierwotny model, który jest dużo prostszy i ma o wiele mniej parametrów.

Możemy zauważyć, że accuracy jest bardzo wysokie, a loss bardzo niski, ale niestety tylko dla zbioru treningowego, zbiór walidacyjny ma lepsze wyniki niż poprzednie modele (jeśli chodzi o accuracy, jeśli chodzi o loss, to jest on dużo wyższy niż początkowy), ale bardzo mocno odstające od zbioru treningowego, co sugeruje overfitting.

```
[ ]: print_history(hisotry2.history)
```





```
[ ]: model3 = tf.keras.Sequential(
    block(channels=20, activation='relu', input_shape=train_images[0].shape) +
    block(channels=40, activation='relu') +
    block(channels=80, activation='relu') +
    block(channels=160, activation='relu') + [
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(units=10,activation='softmax')
    ])
```

```
[ ]: model3.summary()
```

Model: "sequential_17"

Layer (type)	Output Shape	Param #
conv2d_91 (Conv2D)	(None, 32, 32, 20)	560
conv2d_92 (Conv2D)	(None, 32, 32, 20)	3620
max_pooling2d_45 (MaxPoolin	(None, 16, 16, 20)	0

```

g2D)

conv2d_93 (Conv2D)          (None, 16, 16, 40)      7240

conv2d_94 (Conv2D)          (None, 16, 16, 40)     14440

max_pooling2d_46 (MaxPoolin (None, 8, 8, 40)      0
g2D)

conv2d_95 (Conv2D)          (None, 8, 8, 80)       28880

conv2d_96 (Conv2D)          (None, 8, 8, 80)       57680

max_pooling2d_47 (MaxPoolin (None, 4, 4, 80)      0
g2D)

conv2d_97 (Conv2D)          (None, 4, 4, 160)      115360

conv2d_98 (Conv2D)          (None, 4, 4, 160)     230560

max_pooling2d_48 (MaxPoolin (None, 2, 2, 160)      0
g2D)

flatten_17 (Flatten)        (None, 640)            0

dense_17 (Dense)            (None, 10)             6410

```

```

=====
Total params: 464,750
Trainable params: 464,750
Non-trainable params: 0
-----

```

```
[ ]: model3.compile(
    optimizer=tf.keras.optimizers.SGD(learning_rate=0.001, momentum=0.9),
    loss=tf.keras.losses.CategoricalCrossentropy(),
    metrics=['accuracy']
)
```

```
[ ]: start_time = time.time()
hisotry3 = model3.fit(
    x=train_images,
    y=train_labels,
    batch_size=64,
    epochs=150,
    verbose=1,
    validation_data=(val_images, val_labels)
)
```

```
)  
print(f"time: {time.time()-start_time}")
```

Epoch 1/150

2022-10-27 20:10:25.753797: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.
782/782 [=====] - ETA: 0s - loss: 2.2530 - accuracy:
0.1568

2022-10-27 20:10:38.747768: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - 14s 17ms/step - loss: 2.2530 -
accuracy: 0.1568 - val_loss: 2.0130 - val_accuracy: 0.2606

Epoch 2/150

782/782 [=====] - 13s 17ms/step - loss: 1.9138 -
accuracy: 0.3023 - val_loss: 1.7708 - val_accuracy: 0.3632

Epoch 3/150

782/782 [=====] - 13s 17ms/step - loss: 1.6901 -
accuracy: 0.3857 - val_loss: 1.7128 - val_accuracy: 0.3930

Epoch 4/150

782/782 [=====] - 13s 17ms/step - loss: 1.5419 -
accuracy: 0.4435 - val_loss: 1.4859 - val_accuracy: 0.4646

Epoch 5/150

782/782 [=====] - 13s 17ms/step - loss: 1.4403 -
accuracy: 0.4821 - val_loss: 1.4032 - val_accuracy: 0.4954

Epoch 6/150

782/782 [=====] - 13s 16ms/step - loss: 1.3609 -
accuracy: 0.5137 - val_loss: 1.3121 - val_accuracy: 0.5320

Epoch 7/150

782/782 [=====] - 13s 17ms/step - loss: 1.2796 -
accuracy: 0.5440 - val_loss: 1.2709 - val_accuracy: 0.5454

Epoch 8/150

782/782 [=====] - 13s 17ms/step - loss: 1.2075 -
accuracy: 0.5712 - val_loss: 1.2667 - val_accuracy: 0.5463

Epoch 9/150

782/782 [=====] - 13s 17ms/step - loss: 1.1437 -
accuracy: 0.5970 - val_loss: 1.1683 - val_accuracy: 0.5835

Epoch 10/150

782/782 [=====] - 13s 17ms/step - loss: 1.0727 -
accuracy: 0.6245 - val_loss: 1.1763 - val_accuracy: 0.5854

Epoch 11/150

782/782 [=====] - 13s 17ms/step - loss: 1.0061 -
accuracy: 0.6473 - val_loss: 1.1088 - val_accuracy: 0.6118

Epoch 12/150

782/782 [=====] - 13s 17ms/step - loss: 0.9445 -

accuracy: 0.6704 - val_loss: 1.0566 - val_accuracy: 0.6300
 Epoch 13/150
 782/782 [=====] - 13s 17ms/step - loss: 0.8851 -
 accuracy: 0.6933 - val_loss: 1.0650 - val_accuracy: 0.6301
 Epoch 14/150
 782/782 [=====] - 13s 17ms/step - loss: 0.8225 -
 accuracy: 0.7134 - val_loss: 1.0454 - val_accuracy: 0.6408
 Epoch 15/150
 782/782 [=====] - 13s 17ms/step - loss: 0.7657 -
 accuracy: 0.7329 - val_loss: 1.0137 - val_accuracy: 0.6461
 Epoch 16/150
 782/782 [=====] - 13s 17ms/step - loss: 0.7167 -
 accuracy: 0.7502 - val_loss: 1.0779 - val_accuracy: 0.6494
 Epoch 17/150
 782/782 [=====] - 13s 17ms/step - loss: 0.6591 -
 accuracy: 0.7682 - val_loss: 1.0487 - val_accuracy: 0.6516
 Epoch 18/150
 782/782 [=====] - 13s 16ms/step - loss: 0.6011 -
 accuracy: 0.7902 - val_loss: 1.0893 - val_accuracy: 0.6440
 Epoch 19/150
 782/782 [=====] - 13s 17ms/step - loss: 0.5422 -
 accuracy: 0.8101 - val_loss: 1.0952 - val_accuracy: 0.6614
 Epoch 20/150
 782/782 [=====] - 13s 17ms/step - loss: 0.4944 -
 accuracy: 0.8279 - val_loss: 1.1397 - val_accuracy: 0.6529
 Epoch 21/150
 782/782 [=====] - 13s 17ms/step - loss: 0.4365 -
 accuracy: 0.8483 - val_loss: 1.2416 - val_accuracy: 0.6450
 Epoch 22/150
 782/782 [=====] - 13s 17ms/step - loss: 0.3858 -
 accuracy: 0.8643 - val_loss: 1.2674 - val_accuracy: 0.6555
 Epoch 23/150
 782/782 [=====] - 13s 17ms/step - loss: 0.3461 -
 accuracy: 0.8773 - val_loss: 1.3653 - val_accuracy: 0.6519
 Epoch 24/150
 782/782 [=====] - 13s 17ms/step - loss: 0.3080 -
 accuracy: 0.8897 - val_loss: 1.3998 - val_accuracy: 0.6534
 Epoch 25/150
 782/782 [=====] - 13s 16ms/step - loss: 0.2656 -
 accuracy: 0.9056 - val_loss: 1.5213 - val_accuracy: 0.6409
 Epoch 26/150
 782/782 [=====] - 13s 17ms/step - loss: 0.2221 -
 accuracy: 0.9216 - val_loss: 1.5691 - val_accuracy: 0.6581
 Epoch 27/150
 782/782 [=====] - 13s 17ms/step - loss: 0.2103 -
 accuracy: 0.9247 - val_loss: 1.6293 - val_accuracy: 0.6568
 Epoch 28/150
 782/782 [=====] - 13s 17ms/step - loss: 0.1962 -

accuracy: 0.9290 - val_loss: 1.7684 - val_accuracy: 0.6442
Epoch 29/150
782/782 [=====] - 14s 18ms/step - loss: 0.1770 -
accuracy: 0.9380 - val_loss: 1.7525 - val_accuracy: 0.6580
Epoch 30/150
782/782 [=====] - 14s 17ms/step - loss: 0.1432 -
accuracy: 0.9489 - val_loss: 1.9447 - val_accuracy: 0.6453
Epoch 31/150
782/782 [=====] - 13s 17ms/step - loss: 0.1410 -
accuracy: 0.9502 - val_loss: 1.8869 - val_accuracy: 0.6415
Epoch 32/150
782/782 [=====] - 13s 16ms/step - loss: 0.1239 -
accuracy: 0.9560 - val_loss: 2.0502 - val_accuracy: 0.6475
Epoch 33/150
782/782 [=====] - 13s 16ms/step - loss: 0.1256 -
accuracy: 0.9562 - val_loss: 1.9736 - val_accuracy: 0.6625
Epoch 34/150
782/782 [=====] - 13s 17ms/step - loss: 0.1084 -
accuracy: 0.9611 - val_loss: 2.1262 - val_accuracy: 0.6449
Epoch 35/150
782/782 [=====] - 13s 17ms/step - loss: 0.0984 -
accuracy: 0.9638 - val_loss: 2.1979 - val_accuracy: 0.6535
Epoch 36/150
782/782 [=====] - 13s 17ms/step - loss: 0.0958 -
accuracy: 0.9662 - val_loss: 2.0748 - val_accuracy: 0.6578
Epoch 37/150
782/782 [=====] - 13s 17ms/step - loss: 0.0836 -
accuracy: 0.9702 - val_loss: 2.1868 - val_accuracy: 0.6554
Epoch 38/150
782/782 [=====] - 13s 17ms/step - loss: 0.0750 -
accuracy: 0.9740 - val_loss: 2.3209 - val_accuracy: 0.6593
Epoch 39/150
782/782 [=====] - 14s 17ms/step - loss: 0.0869 -
accuracy: 0.9706 - val_loss: 2.2995 - val_accuracy: 0.6621
Epoch 40/150
782/782 [=====] - 14s 18ms/step - loss: 0.0631 -
accuracy: 0.9783 - val_loss: 2.3279 - val_accuracy: 0.6546
Epoch 41/150
782/782 [=====] - 14s 17ms/step - loss: 0.0596 -
accuracy: 0.9792 - val_loss: 2.3856 - val_accuracy: 0.6588
Epoch 42/150
782/782 [=====] - 14s 17ms/step - loss: 0.0556 -
accuracy: 0.9808 - val_loss: 2.4468 - val_accuracy: 0.6540
Epoch 43/150
782/782 [=====] - 13s 17ms/step - loss: 0.0649 -
accuracy: 0.9771 - val_loss: 2.5763 - val_accuracy: 0.6555
Epoch 44/150
782/782 [=====] - 13s 17ms/step - loss: 0.0532 -

accuracy: 0.9816 - val_loss: 2.7291 - val_accuracy: 0.6496
 Epoch 45/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0414 -
 accuracy: 0.9854 - val_loss: 2.6668 - val_accuracy: 0.6513
 Epoch 46/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0550 -
 accuracy: 0.9807 - val_loss: 2.5888 - val_accuracy: 0.6540
 Epoch 47/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0588 -
 accuracy: 0.9804 - val_loss: 2.6018 - val_accuracy: 0.6511
 Epoch 48/150
 782/782 [=====] - 14s 18ms/step - loss: 0.0539 -
 accuracy: 0.9809 - val_loss: 2.5863 - val_accuracy: 0.6631
 Epoch 49/150
 782/782 [=====] - 14s 17ms/step - loss: 0.0397 -
 accuracy: 0.9864 - val_loss: 2.8490 - val_accuracy: 0.6607
 Epoch 50/150
 782/782 [=====] - 14s 17ms/step - loss: 0.0487 -
 accuracy: 0.9823 - val_loss: 2.6761 - val_accuracy: 0.6665
 Epoch 51/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0456 -
 accuracy: 0.9845 - val_loss: 2.5743 - val_accuracy: 0.6640
 Epoch 52/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0326 -
 accuracy: 0.9885 - val_loss: 2.9693 - val_accuracy: 0.6697
 Epoch 53/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0269 -
 accuracy: 0.9912 - val_loss: 2.8219 - val_accuracy: 0.6629
 Epoch 54/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0351 -
 accuracy: 0.9878 - val_loss: 2.7480 - val_accuracy: 0.6650
 Epoch 55/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0315 -
 accuracy: 0.9887 - val_loss: 2.9182 - val_accuracy: 0.6572
 Epoch 56/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0297 -
 accuracy: 0.9898 - val_loss: 2.8996 - val_accuracy: 0.6629
 Epoch 57/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0341 -
 accuracy: 0.9882 - val_loss: 2.9604 - val_accuracy: 0.6587
 Epoch 58/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0305 -
 accuracy: 0.9895 - val_loss: 3.0200 - val_accuracy: 0.6647
 Epoch 59/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0302 -
 accuracy: 0.9894 - val_loss: 3.0581 - val_accuracy: 0.6659
 Epoch 60/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0273 -

accuracy: 0.9912 - val_loss: 2.9402 - val_accuracy: 0.6660
Epoch 61/150
782/782 [=====] - 13s 17ms/step - loss: 0.0418 -
accuracy: 0.9862 - val_loss: 3.0426 - val_accuracy: 0.6476
Epoch 62/150
782/782 [=====] - 13s 17ms/step - loss: 0.0279 -
accuracy: 0.9911 - val_loss: 2.9498 - val_accuracy: 0.6658
Epoch 63/150
782/782 [=====] - 13s 17ms/step - loss: 0.0252 -
accuracy: 0.9918 - val_loss: 3.1412 - val_accuracy: 0.6596
Epoch 64/150
782/782 [=====] - 13s 17ms/step - loss: 0.0309 -
accuracy: 0.9898 - val_loss: 2.9614 - val_accuracy: 0.6682
Epoch 65/150
782/782 [=====] - 13s 16ms/step - loss: 0.0092 -
accuracy: 0.9972 - val_loss: 3.3190 - val_accuracy: 0.6708
Epoch 66/150
782/782 [=====] - 13s 17ms/step - loss: 0.0054 -
accuracy: 0.9981 - val_loss: 3.3942 - val_accuracy: 0.6642
Epoch 67/150
782/782 [=====] - 13s 17ms/step - loss: 0.0222 -
accuracy: 0.9923 - val_loss: 3.2285 - val_accuracy: 0.6559
Epoch 68/150
782/782 [=====] - 13s 17ms/step - loss: 0.0251 -
accuracy: 0.9916 - val_loss: 3.0740 - val_accuracy: 0.6685
Epoch 69/150
782/782 [=====] - 13s 17ms/step - loss: 0.0234 -
accuracy: 0.9921 - val_loss: 3.1688 - val_accuracy: 0.6644
Epoch 70/150
782/782 [=====] - 13s 17ms/step - loss: 0.0242 -
accuracy: 0.9918 - val_loss: 3.2028 - val_accuracy: 0.6697
Epoch 71/150
782/782 [=====] - 13s 17ms/step - loss: 0.0152 -
accuracy: 0.9948 - val_loss: 3.2338 - val_accuracy: 0.6667
Epoch 72/150
782/782 [=====] - 14s 17ms/step - loss: 0.0193 -
accuracy: 0.9936 - val_loss: 3.2737 - val_accuracy: 0.6682
Epoch 73/150
782/782 [=====] - 13s 17ms/step - loss: 0.0360 -
accuracy: 0.9881 - val_loss: 3.1430 - val_accuracy: 0.6564
Epoch 74/150
782/782 [=====] - 13s 17ms/step - loss: 0.0337 -
accuracy: 0.9881 - val_loss: 3.0751 - val_accuracy: 0.6639
Epoch 75/150
782/782 [=====] - 13s 17ms/step - loss: 0.0210 -
accuracy: 0.9930 - val_loss: 3.2226 - val_accuracy: 0.6556
Epoch 76/150
782/782 [=====] - 13s 17ms/step - loss: 0.0108 -

accuracy: 0.9962 - val_loss: 3.2641 - val_accuracy: 0.6664
 Epoch 77/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0200 -
 accuracy: 0.9935 - val_loss: 3.2389 - val_accuracy: 0.6679
 Epoch 78/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0204 -
 accuracy: 0.9930 - val_loss: 3.2275 - val_accuracy: 0.6717
 Epoch 79/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0284 -
 accuracy: 0.9903 - val_loss: 3.0991 - val_accuracy: 0.6698
 Epoch 80/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0086 -
 accuracy: 0.9974 - val_loss: 3.3241 - val_accuracy: 0.6674
 Epoch 81/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0067 -
 accuracy: 0.9978 - val_loss: 3.2919 - val_accuracy: 0.6676
 Epoch 82/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0054 -
 accuracy: 0.9982 - val_loss: 3.4064 - val_accuracy: 0.6650
 Epoch 83/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0139 -
 accuracy: 0.9956 - val_loss: 3.5175 - val_accuracy: 0.6558
 Epoch 84/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0187 -
 accuracy: 0.9935 - val_loss: 3.4662 - val_accuracy: 0.6655
 Epoch 85/150
 782/782 [=====] - 13s 16ms/step - loss: 0.0181 -
 accuracy: 0.9939 - val_loss: 3.3165 - val_accuracy: 0.6664
 Epoch 86/150
 782/782 [=====] - 13s 16ms/step - loss: 0.0260 -
 accuracy: 0.9919 - val_loss: 3.2668 - val_accuracy: 0.6645
 Epoch 87/150
 782/782 [=====] - 13s 16ms/step - loss: 0.0194 -
 accuracy: 0.9939 - val_loss: 3.3585 - val_accuracy: 0.6682
 Epoch 88/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0178 -
 accuracy: 0.9939 - val_loss: 3.3700 - val_accuracy: 0.6675
 Epoch 89/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0211 -
 accuracy: 0.9933 - val_loss: 3.3404 - val_accuracy: 0.6663
 Epoch 90/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0294 -
 accuracy: 0.9902 - val_loss: 3.3598 - val_accuracy: 0.6663
 Epoch 91/150
 782/782 [=====] - 13s 16ms/step - loss: 0.0192 -
 accuracy: 0.9937 - val_loss: 3.3081 - val_accuracy: 0.6642
 Epoch 92/150
 782/782 [=====] - 13s 17ms/step - loss: 0.0102 -

accuracy: 0.9964 - val_loss: 3.3996 - val_accuracy: 0.6679
Epoch 93/150
782/782 [=====] - 13s 17ms/step - loss: 0.0125 -
accuracy: 0.9958 - val_loss: 3.5905 - val_accuracy: 0.6556
Epoch 94/150
782/782 [=====] - 13s 16ms/step - loss: 0.0188 -
accuracy: 0.9935 - val_loss: 3.5541 - val_accuracy: 0.6679
Epoch 95/150
782/782 [=====] - 13s 17ms/step - loss: 0.0082 -
accuracy: 0.9974 - val_loss: 3.4231 - val_accuracy: 0.6783
Epoch 96/150
782/782 [=====] - 13s 17ms/step - loss: 0.0025 -
accuracy: 0.9994 - val_loss: 3.5242 - val_accuracy: 0.6783
Epoch 97/150
782/782 [=====] - 13s 16ms/step - loss: 7.0897e-04 -
accuracy: 0.9998 - val_loss: 3.6393 - val_accuracy: 0.6761
Epoch 98/150
782/782 [=====] - 13s 17ms/step - loss: 3.1175e-04 -
accuracy: 0.9999 - val_loss: 3.6098 - val_accuracy: 0.6845
Epoch 99/150
782/782 [=====] - 13s 17ms/step - loss: 6.2531e-05 -
accuracy: 1.0000 - val_loss: 3.6367 - val_accuracy: 0.6851
Epoch 100/150
782/782 [=====] - 14s 17ms/step - loss: 4.4727e-05 -
accuracy: 1.0000 - val_loss: 3.6590 - val_accuracy: 0.6852
Epoch 101/150
782/782 [=====] - 13s 17ms/step - loss: 3.8122e-05 -
accuracy: 1.0000 - val_loss: 3.6794 - val_accuracy: 0.6854
Epoch 102/150
782/782 [=====] - 13s 17ms/step - loss: 3.3713e-05 -
accuracy: 1.0000 - val_loss: 3.6958 - val_accuracy: 0.6856
Epoch 103/150
782/782 [=====] - 13s 17ms/step - loss: 3.0533e-05 -
accuracy: 1.0000 - val_loss: 3.7122 - val_accuracy: 0.6851
Epoch 104/150
782/782 [=====] - 13s 17ms/step - loss: 2.7980e-05 -
accuracy: 1.0000 - val_loss: 3.7270 - val_accuracy: 0.6851
Epoch 105/150
782/782 [=====] - 13s 17ms/step - loss: 2.5887e-05 -
accuracy: 1.0000 - val_loss: 3.7397 - val_accuracy: 0.6854
Epoch 106/150
782/782 [=====] - 13s 17ms/step - loss: 2.4168e-05 -
accuracy: 1.0000 - val_loss: 3.7517 - val_accuracy: 0.6855
Epoch 107/150
782/782 [=====] - 13s 17ms/step - loss: 2.2679e-05 -
accuracy: 1.0000 - val_loss: 3.7631 - val_accuracy: 0.6852
Epoch 108/150
782/782 [=====] - 13s 17ms/step - loss: 2.1392e-05 -

accuracy: 1.0000 - val_loss: 3.7741 - val_accuracy: 0.6855
Epoch 109/150
782/782 [=====] - 13s 17ms/step - loss: 2.0255e-05 -
accuracy: 1.0000 - val_loss: 3.7844 - val_accuracy: 0.6855
Epoch 110/150
782/782 [=====] - 13s 17ms/step - loss: 1.9265e-05 -
accuracy: 1.0000 - val_loss: 3.7942 - val_accuracy: 0.6852
Epoch 111/150
782/782 [=====] - 13s 17ms/step - loss: 1.8371e-05 -
accuracy: 1.0000 - val_loss: 3.8040 - val_accuracy: 0.6849
Epoch 112/150
782/782 [=====] - 13s 17ms/step - loss: 1.7574e-05 -
accuracy: 1.0000 - val_loss: 3.8132 - val_accuracy: 0.6850
Epoch 113/150
782/782 [=====] - 13s 17ms/step - loss: 1.6845e-05 -
accuracy: 1.0000 - val_loss: 3.8214 - val_accuracy: 0.6846
Epoch 114/150
782/782 [=====] - 13s 17ms/step - loss: 1.6183e-05 -
accuracy: 1.0000 - val_loss: 3.8295 - val_accuracy: 0.6846
Epoch 115/150
782/782 [=====] - 13s 17ms/step - loss: 1.5580e-05 -
accuracy: 1.0000 - val_loss: 3.8377 - val_accuracy: 0.6845
Epoch 116/150
782/782 [=====] - 13s 16ms/step - loss: 1.5017e-05 -
accuracy: 1.0000 - val_loss: 3.8451 - val_accuracy: 0.6845
Epoch 117/150
782/782 [=====] - 13s 16ms/step - loss: 1.4502e-05 -
accuracy: 1.0000 - val_loss: 3.8524 - val_accuracy: 0.6846
Epoch 118/150
782/782 [=====] - 13s 17ms/step - loss: 1.4022e-05 -
accuracy: 1.0000 - val_loss: 3.8594 - val_accuracy: 0.6849
Epoch 119/150
782/782 [=====] - 13s 17ms/step - loss: 1.3579e-05 -
accuracy: 1.0000 - val_loss: 3.8664 - val_accuracy: 0.6847
Epoch 120/150
782/782 [=====] - 13s 17ms/step - loss: 1.3167e-05 -
accuracy: 1.0000 - val_loss: 3.8731 - val_accuracy: 0.6843
Epoch 121/150
782/782 [=====] - 13s 17ms/step - loss: 1.2779e-05 -
accuracy: 1.0000 - val_loss: 3.8796 - val_accuracy: 0.6843
Epoch 122/150
782/782 [=====] - 13s 17ms/step - loss: 1.2418e-05 -
accuracy: 1.0000 - val_loss: 3.8859 - val_accuracy: 0.6842
Epoch 123/150
782/782 [=====] - 13s 17ms/step - loss: 1.2077e-05 -
accuracy: 1.0000 - val_loss: 3.8921 - val_accuracy: 0.6841
Epoch 124/150
782/782 [=====] - 13s 17ms/step - loss: 1.1757e-05 -

```

accuracy: 1.0000 - val_loss: 3.8980 - val_accuracy: 0.6842
Epoch 125/150
782/782 [=====] - 13s 17ms/step - loss: 1.1455e-05 -
accuracy: 1.0000 - val_loss: 3.9039 - val_accuracy: 0.6842
Epoch 126/150
782/782 [=====] - 13s 17ms/step - loss: 1.1168e-05 -
accuracy: 1.0000 - val_loss: 3.9094 - val_accuracy: 0.6841
Epoch 127/150
782/782 [=====] - 13s 17ms/step - loss: 1.0894e-05 -
accuracy: 1.0000 - val_loss: 3.9151 - val_accuracy: 0.6840
Epoch 128/150
782/782 [=====] - 13s 17ms/step - loss: 1.0636e-05 -
accuracy: 1.0000 - val_loss: 3.9205 - val_accuracy: 0.6842
Epoch 129/150
782/782 [=====] - 13s 17ms/step - loss: 1.0393e-05 -
accuracy: 1.0000 - val_loss: 3.9259 - val_accuracy: 0.6842
Epoch 130/150
782/782 [=====] - 13s 17ms/step - loss: 1.0160e-05 -
accuracy: 1.0000 - val_loss: 3.9310 - val_accuracy: 0.6841
Epoch 131/150
782/782 [=====] - 13s 17ms/step - loss: 9.9389e-06 -
accuracy: 1.0000 - val_loss: 3.9361 - val_accuracy: 0.6842
Epoch 132/150
782/782 [=====] - 13s 17ms/step - loss: 9.7254e-06 -
accuracy: 1.0000 - val_loss: 3.9409 - val_accuracy: 0.6842
Epoch 133/150
782/782 [=====] - 13s 17ms/step - loss: 9.5227e-06 -
accuracy: 1.0000 - val_loss: 3.9457 - val_accuracy: 0.6843
Epoch 134/150
782/782 [=====] - 13s 17ms/step - loss: 9.3284e-06 -
accuracy: 1.0000 - val_loss: 3.9504 - val_accuracy: 0.6844
Epoch 135/150
782/782 [=====] - 13s 17ms/step - loss: 9.1424e-06 -
accuracy: 1.0000 - val_loss: 3.9551 - val_accuracy: 0.6846
Epoch 136/150
782/782 [=====] - 13s 17ms/step - loss: 8.9649e-06 -
accuracy: 1.0000 - val_loss: 3.9595 - val_accuracy: 0.6846
Epoch 137/150
782/782 [=====] - 13s 17ms/step - loss: 8.7935e-06 -
accuracy: 1.0000 - val_loss: 3.9641 - val_accuracy: 0.6845
Epoch 138/150
782/782 [=====] - 13s 17ms/step - loss: 8.6290e-06 -
accuracy: 1.0000 - val_loss: 3.9684 - val_accuracy: 0.6845
Epoch 139/150
782/782 [=====] - 13s 17ms/step - loss: 8.4730e-06 -
accuracy: 1.0000 - val_loss: 3.9729 - val_accuracy: 0.6846
Epoch 140/150
782/782 [=====] - 13s 17ms/step - loss: 8.3204e-06 -

```



```

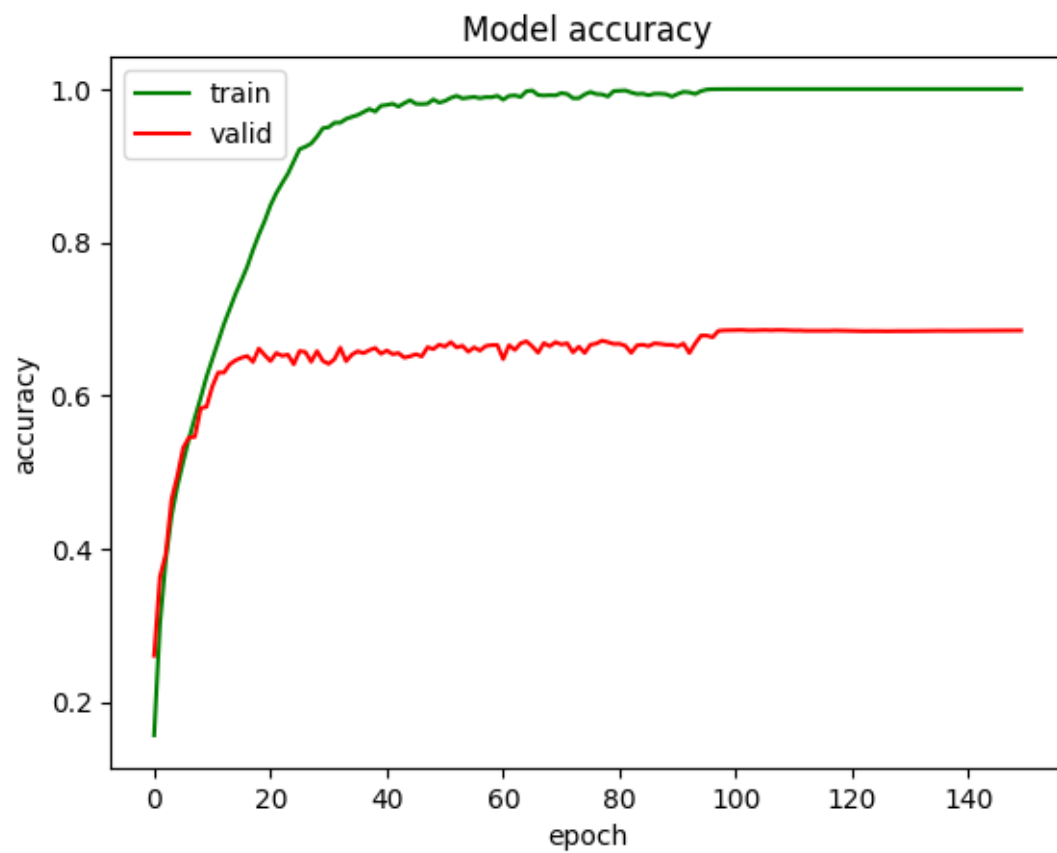
accuracy: 1.0000 - val_loss: 3.9771 - val_accuracy: 0.6846
Epoch 141/150
782/782 [=====] - 13s 17ms/step - loss: 8.1746e-06 -
accuracy: 1.0000 - val_loss: 3.9812 - val_accuracy: 0.6847
Epoch 142/150
782/782 [=====] - 13s 17ms/step - loss: 8.0346e-06 -
accuracy: 1.0000 - val_loss: 3.9853 - val_accuracy: 0.6847
Epoch 143/150
782/782 [=====] - 13s 17ms/step - loss: 7.8985e-06 -
accuracy: 1.0000 - val_loss: 3.9895 - val_accuracy: 0.6847
Epoch 144/150
782/782 [=====] - 13s 17ms/step - loss: 7.7677e-06 -
accuracy: 1.0000 - val_loss: 3.9934 - val_accuracy: 0.6847
Epoch 145/150
782/782 [=====] - 13s 17ms/step - loss: 7.6416e-06 -
accuracy: 1.0000 - val_loss: 3.9973 - val_accuracy: 0.6848
Epoch 146/150
782/782 [=====] - 13s 17ms/step - loss: 7.5192e-06 -
accuracy: 1.0000 - val_loss: 4.0011 - val_accuracy: 0.6849
Epoch 147/150
782/782 [=====] - 13s 17ms/step - loss: 7.4024e-06 -
accuracy: 1.0000 - val_loss: 4.0050 - val_accuracy: 0.6849
Epoch 148/150
782/782 [=====] - 13s 17ms/step - loss: 7.2875e-06 -
accuracy: 1.0000 - val_loss: 4.0085 - val_accuracy: 0.6850
Epoch 149/150
782/782 [=====] - 13s 17ms/step - loss: 7.1777e-06 -
accuracy: 1.0000 - val_loss: 4.0124 - val_accuracy: 0.6850
Epoch 150/150
782/782 [=====] - 13s 17ms/step - loss: 7.0706e-06 -
accuracy: 1.0000 - val_loss: 4.0160 - val_accuracy: 0.6850
time: 1967.3622839450836

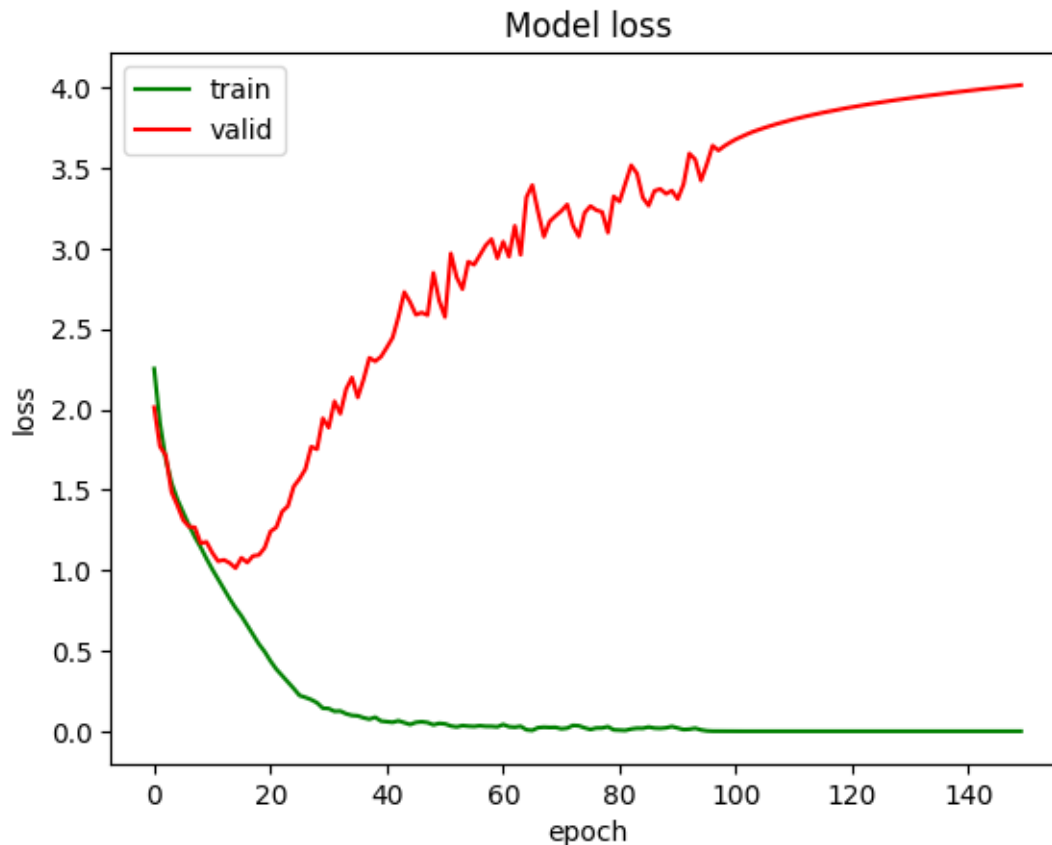
```

Tym razem model liczył się odrobinę wolniej od poprzedniego modelu, dalej jestem zaskoczony, że tak duża różnica w rozmiarach modelu liczy się tak niewiele wolniej.

Mamy bardzo typowy przykład overfittingu, na danych treningowych model działa wyśmienicie, ale na walidacyjnych radzi sobie znacznie gorzej(choć wciąż lepiej od poprzednich modeli). Co ciekawe, wartość loss jest większa niż startowa, dla danych walidacyjnych. Nasz model nauczył się wszystkich zdjęć ze zbioru treningowego na pamięć.

```
[ ]: print_history(hisotry3.history)
```





1 Batch norm

Dodałem batch normalization dzięki funkcji batchnormalization, z kerasa.

Dodałem te warstwy normalizujące po każdej warstwie konwolucyjnej, zgodnie z poleceniem.

```
[ ]: model = tf.keras.Sequential(
    block(channels=20, activation='relu', input_shape=train_images[0].shape,
    ↪ batch_normalisation=True) +
    block(channels=40, activation='relu', batch_normalisation=True) +
    block(channels=80, activation='relu', batch_normalisation=True) +
    block(channels=160, activation='relu', batch_normalisation=True) + [
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(units=10, activation='softmax')
])
```

```
[ ]: model.summary()
```

Model: "sequential_18"

Layer (type)	Output Shape	Param #
--------------	--------------	---------

=====		
conv2d_105 (Conv2D)	(None, 32, 32, 20)	560
batch_normalization_30 (Batch Normalization)	(None, 32, 32, 20)	80
conv2d_106 (Conv2D)	(None, 32, 32, 20)	3620
batch_normalization_31 (Batch Normalization)	(None, 32, 32, 20)	80
max_pooling2d_52 (MaxPooling2D)	(None, 16, 16, 20)	0
conv2d_107 (Conv2D)	(None, 16, 16, 40)	7240
batch_normalization_32 (Batch Normalization)	(None, 16, 16, 40)	160
conv2d_108 (Conv2D)	(None, 16, 16, 40)	14440
batch_normalization_33 (Batch Normalization)	(None, 16, 16, 40)	160
max_pooling2d_53 (MaxPooling2D)	(None, 8, 8, 40)	0
conv2d_109 (Conv2D)	(None, 8, 8, 80)	28880
batch_normalization_34 (Batch Normalization)	(None, 8, 8, 80)	320
conv2d_110 (Conv2D)	(None, 8, 8, 80)	57680
batch_normalization_35 (Batch Normalization)	(None, 8, 8, 80)	320
max_pooling2d_54 (MaxPooling2D)	(None, 4, 4, 80)	0
conv2d_111 (Conv2D)	(None, 4, 4, 160)	115360
batch_normalization_36 (Batch Normalization)	(None, 4, 4, 160)	640
conv2d_112 (Conv2D)	(None, 4, 4, 160)	230560
batch_normalization_37 (Batch Normalization)	(None, 4, 4, 160)	640

```

chNormalization)

max_pooling2d_55 (MaxPoolin (None, 2, 2, 160)      0
g2D)

flatten_18 (Flatten)      (None, 640)      0

dense_18 (Dense)      (None, 10)      6410

```

```

=====
Total params: 467,150
Trainable params: 465,950
Non-trainable params: 1,200
-----

```

```

[ ]: model.compile(
    optimizer=tf.keras.optimizers.SGD(learning_rate=0.001, momentum=0.9),
    loss=tf.keras.losses.CategoricalCrossentropy(),
    metrics=['accuracy']
)

```

```

[ ]: start_time = time.time()
hisotry = model.fit(
    x=train_images,
    y=train_labels,
    batch_size=64,
    epochs=150,
    verbose=1,
    validation_data=(val_images, val_labels)
)
print(f"time: {time.time()-start_time}")

```

Epoch 1/150

```

2022-10-27 22:04:04.992124: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - ETA: 0s - loss: 1.5026 - accuracy:
0.4813

```

```

2022-10-27 22:04:20.253288: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

```

```

782/782 [=====] - 17s 21ms/step - loss: 1.5026 -
accuracy: 0.4813 - val_loss: 1.4780 - val_accuracy: 0.4960
Epoch 2/150

```

```

782/782 [=====] - 16s 20ms/step - loss: 1.0254 -
accuracy: 0.6400 - val_loss: 1.2252 - val_accuracy: 0.5861

```

Epoch 3/150
782/782 [=====] - 16s 20ms/step - loss: 0.8234 - accuracy: 0.7124 - val_loss: 0.9440 - val_accuracy: 0.6751

Epoch 4/150
782/782 [=====] - 17s 22ms/step - loss: 0.6751 - accuracy: 0.7665 - val_loss: 0.8876 - val_accuracy: 0.6969

Epoch 5/150
782/782 [=====] - 16s 20ms/step - loss: 0.5589 - accuracy: 0.8056 - val_loss: 0.9135 - val_accuracy: 0.6931

Epoch 6/150
782/782 [=====] - 16s 20ms/step - loss: 0.4569 - accuracy: 0.8430 - val_loss: 0.8722 - val_accuracy: 0.7106

Epoch 7/150
782/782 [=====] - 15s 20ms/step - loss: 0.3600 - accuracy: 0.8783 - val_loss: 0.9563 - val_accuracy: 0.6991

Epoch 8/150
782/782 [=====] - 15s 20ms/step - loss: 0.2790 - accuracy: 0.9098 - val_loss: 0.9088 - val_accuracy: 0.7189

Epoch 9/150
782/782 [=====] - 16s 20ms/step - loss: 0.2035 - accuracy: 0.9390 - val_loss: 0.9961 - val_accuracy: 0.7136

Epoch 10/150
782/782 [=====] - 15s 20ms/step - loss: 0.1463 - accuracy: 0.9586 - val_loss: 0.9580 - val_accuracy: 0.7262

Epoch 11/150
782/782 [=====] - 16s 20ms/step - loss: 0.0986 - accuracy: 0.9767 - val_loss: 1.0420 - val_accuracy: 0.7195

Epoch 12/150
782/782 [=====] - 15s 20ms/step - loss: 0.0623 - accuracy: 0.9891 - val_loss: 1.0272 - val_accuracy: 0.7322

Epoch 13/150
782/782 [=====] - 15s 20ms/step - loss: 0.0414 - accuracy: 0.9949 - val_loss: 1.0408 - val_accuracy: 0.7322

Epoch 14/150
782/782 [=====] - 16s 20ms/step - loss: 0.0266 - accuracy: 0.9980 - val_loss: 1.0706 - val_accuracy: 0.7361

Epoch 15/150
782/782 [=====] - 16s 20ms/step - loss: 0.0176 - accuracy: 0.9994 - val_loss: 1.0947 - val_accuracy: 0.7333

Epoch 16/150
782/782 [=====] - 15s 20ms/step - loss: 0.0135 - accuracy: 0.9995 - val_loss: 1.1058 - val_accuracy: 0.7368

Epoch 17/150
782/782 [=====] - 15s 20ms/step - loss: 0.0104 - accuracy: 0.9997 - val_loss: 1.1203 - val_accuracy: 0.7388

Epoch 18/150
782/782 [=====] - 16s 20ms/step - loss: 0.0086 - accuracy: 0.9998 - val_loss: 1.1328 - val_accuracy: 0.7382

Epoch 19/150
782/782 [=====] - 16s 20ms/step - loss: 0.0080 - accuracy: 0.9998 - val_loss: 1.1325 - val_accuracy: 0.7377

Epoch 20/150
782/782 [=====] - 16s 20ms/step - loss: 0.0064 - accuracy: 0.9999 - val_loss: 1.1496 - val_accuracy: 0.7387

Epoch 21/150
782/782 [=====] - 16s 20ms/step - loss: 0.0056 - accuracy: 0.9999 - val_loss: 1.1673 - val_accuracy: 0.7393

Epoch 22/150
782/782 [=====] - 16s 20ms/step - loss: 0.0066 - accuracy: 0.9998 - val_loss: 1.1798 - val_accuracy: 0.7352

Epoch 23/150
782/782 [=====] - 16s 20ms/step - loss: 0.0048 - accuracy: 1.0000 - val_loss: 1.1765 - val_accuracy: 0.7397

Epoch 24/150
782/782 [=====] - 16s 20ms/step - loss: 0.0043 - accuracy: 0.9999 - val_loss: 1.1921 - val_accuracy: 0.7393

Epoch 25/150
782/782 [=====] - 16s 20ms/step - loss: 0.0039 - accuracy: 0.9999 - val_loss: 1.2036 - val_accuracy: 0.7399

Epoch 26/150
782/782 [=====] - 16s 20ms/step - loss: 0.0036 - accuracy: 1.0000 - val_loss: 1.2080 - val_accuracy: 0.7400

Epoch 27/150
782/782 [=====] - 16s 20ms/step - loss: 0.0031 - accuracy: 1.0000 - val_loss: 1.2164 - val_accuracy: 0.7386

Epoch 28/150
782/782 [=====] - 16s 20ms/step - loss: 0.0028 - accuracy: 1.0000 - val_loss: 1.2198 - val_accuracy: 0.7421

Epoch 29/150
782/782 [=====] - 16s 20ms/step - loss: 0.0029 - accuracy: 0.9999 - val_loss: 1.2251 - val_accuracy: 0.7389

Epoch 30/150
782/782 [=====] - 16s 20ms/step - loss: 0.0033 - accuracy: 0.9999 - val_loss: 1.2458 - val_accuracy: 0.7409

Epoch 31/150
782/782 [=====] - 15s 20ms/step - loss: 0.0025 - accuracy: 1.0000 - val_loss: 1.2428 - val_accuracy: 0.7422

Epoch 32/150
782/782 [=====] - 16s 20ms/step - loss: 0.0023 - accuracy: 1.0000 - val_loss: 1.2516 - val_accuracy: 0.7388

Epoch 33/150
782/782 [=====] - 16s 20ms/step - loss: 0.0024 - accuracy: 1.0000 - val_loss: 1.2559 - val_accuracy: 0.7410

Epoch 34/150
782/782 [=====] - 16s 20ms/step - loss: 0.0022 - accuracy: 1.0000 - val_loss: 1.2627 - val_accuracy: 0.7395

Epoch 35/150
782/782 [=====] - 16s 20ms/step - loss: 0.0020 - accuracy: 1.0000 - val_loss: 1.2664 - val_accuracy: 0.7386

Epoch 36/150
782/782 [=====] - 16s 20ms/step - loss: 0.0018 - accuracy: 1.0000 - val_loss: 1.2705 - val_accuracy: 0.7396

Epoch 37/150
782/782 [=====] - 16s 20ms/step - loss: 0.0018 - accuracy: 1.0000 - val_loss: 1.2727 - val_accuracy: 0.7396

Epoch 38/150
782/782 [=====] - 16s 20ms/step - loss: 0.0018 - accuracy: 1.0000 - val_loss: 1.2797 - val_accuracy: 0.7397

Epoch 39/150
782/782 [=====] - 16s 20ms/step - loss: 0.0018 - accuracy: 1.0000 - val_loss: 1.2919 - val_accuracy: 0.7375

Epoch 40/150
782/782 [=====] - 16s 20ms/step - loss: 0.0016 - accuracy: 1.0000 - val_loss: 1.2872 - val_accuracy: 0.7412

Epoch 41/150
782/782 [=====] - 16s 20ms/step - loss: 0.0017 - accuracy: 0.9999 - val_loss: 1.2951 - val_accuracy: 0.7393

Epoch 42/150
782/782 [=====] - 16s 20ms/step - loss: 0.0016 - accuracy: 1.0000 - val_loss: 1.2991 - val_accuracy: 0.7405

Epoch 43/150
782/782 [=====] - 16s 20ms/step - loss: 0.0014 - accuracy: 1.0000 - val_loss: 1.2975 - val_accuracy: 0.7415

Epoch 44/150
782/782 [=====] - 16s 20ms/step - loss: 0.0013 - accuracy: 1.0000 - val_loss: 1.3045 - val_accuracy: 0.7415

Epoch 45/150
782/782 [=====] - 16s 20ms/step - loss: 0.0013 - accuracy: 1.0000 - val_loss: 1.3044 - val_accuracy: 0.7413

Epoch 46/150
782/782 [=====] - 16s 20ms/step - loss: 0.0013 - accuracy: 1.0000 - val_loss: 1.3104 - val_accuracy: 0.7412

Epoch 47/150
782/782 [=====] - 15s 20ms/step - loss: 0.0013 - accuracy: 1.0000 - val_loss: 1.3126 - val_accuracy: 0.7405

Epoch 48/150
782/782 [=====] - 15s 20ms/step - loss: 0.0011 - accuracy: 1.0000 - val_loss: 1.3180 - val_accuracy: 0.7415

Epoch 49/150
782/782 [=====] - 15s 20ms/step - loss: 0.0012 - accuracy: 1.0000 - val_loss: 1.3224 - val_accuracy: 0.7416

Epoch 50/150
782/782 [=====] - 15s 19ms/step - loss: 0.0012 - accuracy: 1.0000 - val_loss: 1.3262 - val_accuracy: 0.7419

Epoch 51/150
782/782 [=====] - 15s 20ms/step - loss: 0.0011 -
accuracy: 1.0000 - val_loss: 1.3350 - val_accuracy: 0.7413

Epoch 52/150
782/782 [=====] - 15s 19ms/step - loss: 0.0011 -
accuracy: 1.0000 - val_loss: 1.3319 - val_accuracy: 0.7412

Epoch 53/150
782/782 [=====] - 15s 20ms/step - loss: 0.0011 -
accuracy: 1.0000 - val_loss: 1.3381 - val_accuracy: 0.7414

Epoch 54/150
782/782 [=====] - 15s 19ms/step - loss: 0.0010 -
accuracy: 1.0000 - val_loss: 1.3363 - val_accuracy: 0.7397

Epoch 55/150
782/782 [=====] - 15s 19ms/step - loss: 9.3960e-04 -
accuracy: 1.0000 - val_loss: 1.3433 - val_accuracy: 0.7410

Epoch 56/150
782/782 [=====] - 15s 19ms/step - loss: 0.0010 -
accuracy: 1.0000 - val_loss: 1.3478 - val_accuracy: 0.7394

Epoch 57/150
782/782 [=====] - 15s 20ms/step - loss: 9.2894e-04 -
accuracy: 1.0000 - val_loss: 1.3498 - val_accuracy: 0.7406

Epoch 58/150
782/782 [=====] - 16s 20ms/step - loss: 9.2239e-04 -
accuracy: 1.0000 - val_loss: 1.3553 - val_accuracy: 0.7418

Epoch 59/150
782/782 [=====] - 15s 19ms/step - loss: 9.9525e-04 -
accuracy: 1.0000 - val_loss: 1.3763 - val_accuracy: 0.7359

Epoch 60/150
782/782 [=====] - 15s 19ms/step - loss: 0.0055 -
accuracy: 0.9992 - val_loss: 1.3676 - val_accuracy: 0.7329

Epoch 61/150
782/782 [=====] - 15s 20ms/step - loss: 0.0021 -
accuracy: 0.9999 - val_loss: 1.3699 - val_accuracy: 0.7389

Epoch 62/150
782/782 [=====] - 15s 19ms/step - loss: 0.0014 -
accuracy: 1.0000 - val_loss: 1.3694 - val_accuracy: 0.7400

Epoch 63/150
782/782 [=====] - 15s 19ms/step - loss: 0.0011 -
accuracy: 1.0000 - val_loss: 1.3760 - val_accuracy: 0.7403

Epoch 64/150
782/782 [=====] - 15s 19ms/step - loss: 0.0010 -
accuracy: 1.0000 - val_loss: 1.3762 - val_accuracy: 0.7409

Epoch 65/150
782/782 [=====] - 15s 19ms/step - loss: 9.8151e-04 -
accuracy: 1.0000 - val_loss: 1.3760 - val_accuracy: 0.7416

Epoch 66/150
782/782 [=====] - 15s 19ms/step - loss: 8.8301e-04 -
accuracy: 1.0000 - val_loss: 1.3747 - val_accuracy: 0.7413

Epoch 67/150
782/782 [=====] - 15s 19ms/step - loss: 7.9803e-04 -
accuracy: 1.0000 - val_loss: 1.3815 - val_accuracy: 0.7421

Epoch 68/150
782/782 [=====] - 15s 19ms/step - loss: 0.0012 -
accuracy: 1.0000 - val_loss: 1.3820 - val_accuracy: 0.7399

Epoch 69/150
782/782 [=====] - 15s 19ms/step - loss: 9.2804e-04 -
accuracy: 1.0000 - val_loss: 1.3838 - val_accuracy: 0.7387

Epoch 70/150
782/782 [=====] - 15s 19ms/step - loss: 8.6036e-04 -
accuracy: 1.0000 - val_loss: 1.3897 - val_accuracy: 0.7403

Epoch 71/150
782/782 [=====] - 15s 20ms/step - loss: 7.9090e-04 -
accuracy: 1.0000 - val_loss: 1.3910 - val_accuracy: 0.7390

Epoch 72/150
782/782 [=====] - 15s 19ms/step - loss: 8.7235e-04 -
accuracy: 1.0000 - val_loss: 1.3934 - val_accuracy: 0.7397

Epoch 73/150
782/782 [=====] - 15s 19ms/step - loss: 7.2716e-04 -
accuracy: 1.0000 - val_loss: 1.3927 - val_accuracy: 0.7412

Epoch 74/150
782/782 [=====] - 15s 19ms/step - loss: 7.4694e-04 -
accuracy: 1.0000 - val_loss: 1.4010 - val_accuracy: 0.7401

Epoch 75/150
782/782 [=====] - 15s 19ms/step - loss: 7.8145e-04 -
accuracy: 1.0000 - val_loss: 1.3993 - val_accuracy: 0.7433

Epoch 76/150
782/782 [=====] - 15s 19ms/step - loss: 7.2696e-04 -
accuracy: 1.0000 - val_loss: 1.4020 - val_accuracy: 0.7410

Epoch 77/150
782/782 [=====] - 15s 19ms/step - loss: 7.1300e-04 -
accuracy: 1.0000 - val_loss: 1.4082 - val_accuracy: 0.7414

Epoch 78/150
782/782 [=====] - 15s 20ms/step - loss: 6.6163e-04 -
accuracy: 1.0000 - val_loss: 1.4103 - val_accuracy: 0.7411

Epoch 79/150
782/782 [=====] - 15s 19ms/step - loss: 0.0012 -
accuracy: 0.9999 - val_loss: 1.4296 - val_accuracy: 0.7351

Epoch 80/150
782/782 [=====] - 15s 19ms/step - loss: 0.0047 -
accuracy: 0.9993 - val_loss: 1.4034 - val_accuracy: 0.7374

Epoch 81/150
782/782 [=====] - 15s 19ms/step - loss: 0.0012 -
accuracy: 1.0000 - val_loss: 1.4067 - val_accuracy: 0.7371

Epoch 82/150
782/782 [=====] - 15s 19ms/step - loss: 0.0011 -
accuracy: 1.0000 - val_loss: 1.4094 - val_accuracy: 0.7364

Epoch 83/150
782/782 [=====] - 15s 19ms/step - loss: 8.5697e-04 - accuracy: 1.0000 - val_loss: 1.4144 - val_accuracy: 0.7382

Epoch 84/150
782/782 [=====] - 15s 19ms/step - loss: 9.9288e-04 - accuracy: 1.0000 - val_loss: 1.4189 - val_accuracy: 0.7375

Epoch 85/150
782/782 [=====] - 15s 19ms/step - loss: 7.7958e-04 - accuracy: 1.0000 - val_loss: 1.4198 - val_accuracy: 0.7373

Epoch 86/150
782/782 [=====] - 15s 19ms/step - loss: 6.7466e-04 - accuracy: 1.0000 - val_loss: 1.4224 - val_accuracy: 0.7402

Epoch 87/150
782/782 [=====] - 15s 20ms/step - loss: 6.9757e-04 - accuracy: 1.0000 - val_loss: 1.4319 - val_accuracy: 0.7392

Epoch 88/150
782/782 [=====] - 15s 20ms/step - loss: 0.0013 - accuracy: 0.9999 - val_loss: 1.4306 - val_accuracy: 0.7377

Epoch 89/150
782/782 [=====] - 15s 19ms/step - loss: 7.6174e-04 - accuracy: 1.0000 - val_loss: 1.4311 - val_accuracy: 0.7367

Epoch 90/150
782/782 [=====] - 15s 20ms/step - loss: 7.2263e-04 - accuracy: 1.0000 - val_loss: 1.4318 - val_accuracy: 0.7383

Epoch 91/150
782/782 [=====] - 15s 19ms/step - loss: 5.8430e-04 - accuracy: 1.0000 - val_loss: 1.4385 - val_accuracy: 0.7379

Epoch 92/150
782/782 [=====] - 15s 20ms/step - loss: 5.9802e-04 - accuracy: 1.0000 - val_loss: 1.4374 - val_accuracy: 0.7389

Epoch 93/150
782/782 [=====] - 15s 20ms/step - loss: 5.7406e-04 - accuracy: 1.0000 - val_loss: 1.4393 - val_accuracy: 0.7386

Epoch 94/150
782/782 [=====] - 15s 19ms/step - loss: 6.2019e-04 - accuracy: 1.0000 - val_loss: 1.4409 - val_accuracy: 0.7392

Epoch 95/150
782/782 [=====] - 15s 20ms/step - loss: 5.5834e-04 - accuracy: 1.0000 - val_loss: 1.4433 - val_accuracy: 0.7387

Epoch 96/150
782/782 [=====] - 15s 20ms/step - loss: 5.0662e-04 - accuracy: 1.0000 - val_loss: 1.4465 - val_accuracy: 0.7397

Epoch 97/150
782/782 [=====] - 15s 20ms/step - loss: 4.8730e-04 - accuracy: 1.0000 - val_loss: 1.4487 - val_accuracy: 0.7379

Epoch 98/150
782/782 [=====] - 15s 20ms/step - loss: 4.9862e-04 - accuracy: 1.0000 - val_loss: 1.4514 - val_accuracy: 0.7398

Epoch 99/150
782/782 [=====] - 15s 20ms/step - loss: 6.0171e-04 - accuracy: 1.0000 - val_loss: 1.4560 - val_accuracy: 0.7380

Epoch 100/150
782/782 [=====] - 15s 20ms/step - loss: 4.8082e-04 - accuracy: 1.0000 - val_loss: 1.4536 - val_accuracy: 0.7386

Epoch 101/150
782/782 [=====] - 15s 20ms/step - loss: 4.4656e-04 - accuracy: 1.0000 - val_loss: 1.4556 - val_accuracy: 0.7393

Epoch 102/150
782/782 [=====] - 15s 20ms/step - loss: 5.7542e-04 - accuracy: 1.0000 - val_loss: 1.4556 - val_accuracy: 0.7385

Epoch 103/150
782/782 [=====] - 15s 20ms/step - loss: 5.0106e-04 - accuracy: 1.0000 - val_loss: 1.4539 - val_accuracy: 0.7397

Epoch 104/150
782/782 [=====] - 15s 20ms/step - loss: 4.6609e-04 - accuracy: 1.0000 - val_loss: 1.4595 - val_accuracy: 0.7394

Epoch 105/150
782/782 [=====] - 15s 19ms/step - loss: 4.2501e-04 - accuracy: 1.0000 - val_loss: 1.4639 - val_accuracy: 0.7386

Epoch 106/150
782/782 [=====] - 15s 19ms/step - loss: 3.9558e-04 - accuracy: 1.0000 - val_loss: 1.4626 - val_accuracy: 0.7387

Epoch 107/150
782/782 [=====] - 15s 20ms/step - loss: 5.1860e-04 - accuracy: 1.0000 - val_loss: 1.4726 - val_accuracy: 0.7389

Epoch 108/150
782/782 [=====] - 15s 20ms/step - loss: 4.2724e-04 - accuracy: 1.0000 - val_loss: 1.4674 - val_accuracy: 0.7385

Epoch 109/150
782/782 [=====] - 15s 19ms/step - loss: 4.4954e-04 - accuracy: 1.0000 - val_loss: 1.4712 - val_accuracy: 0.7391

Epoch 110/150
782/782 [=====] - 15s 20ms/step - loss: 4.6241e-04 - accuracy: 1.0000 - val_loss: 1.4730 - val_accuracy: 0.7380

Epoch 111/150
782/782 [=====] - 15s 19ms/step - loss: 4.3391e-04 - accuracy: 1.0000 - val_loss: 1.4717 - val_accuracy: 0.7387

Epoch 112/150
782/782 [=====] - 15s 19ms/step - loss: 4.2202e-04 - accuracy: 1.0000 - val_loss: 1.4781 - val_accuracy: 0.7388

Epoch 113/150
782/782 [=====] - 15s 20ms/step - loss: 4.2225e-04 - accuracy: 1.0000 - val_loss: 1.4740 - val_accuracy: 0.7401

Epoch 114/150
782/782 [=====] - 15s 20ms/step - loss: 3.8804e-04 - accuracy: 1.0000 - val_loss: 1.4771 - val_accuracy: 0.7389

Epoch 115/150
782/782 [=====] - 15s 20ms/step - loss: 3.5012e-04 - accuracy: 1.0000 - val_loss: 1.4797 - val_accuracy: 0.7396

Epoch 116/150
782/782 [=====] - 15s 20ms/step - loss: 4.0253e-04 - accuracy: 1.0000 - val_loss: 1.4809 - val_accuracy: 0.7399

Epoch 117/150
782/782 [=====] - 15s 19ms/step - loss: 3.6453e-04 - accuracy: 1.0000 - val_loss: 1.4813 - val_accuracy: 0.7398

Epoch 118/150
782/782 [=====] - 15s 20ms/step - loss: 3.6232e-04 - accuracy: 1.0000 - val_loss: 1.4843 - val_accuracy: 0.7388

Epoch 119/150
782/782 [=====] - 15s 20ms/step - loss: 3.9150e-04 - accuracy: 1.0000 - val_loss: 1.4883 - val_accuracy: 0.7398

Epoch 120/150
782/782 [=====] - 15s 20ms/step - loss: 0.0015 - accuracy: 0.9998 - val_loss: 1.4933 - val_accuracy: 0.7402

Epoch 121/150
782/782 [=====] - 15s 19ms/step - loss: 5.6292e-04 - accuracy: 1.0000 - val_loss: 1.4905 - val_accuracy: 0.7402

Epoch 122/150
782/782 [=====] - 15s 20ms/step - loss: 5.1645e-04 - accuracy: 1.0000 - val_loss: 1.4874 - val_accuracy: 0.7416

Epoch 123/150
782/782 [=====] - 15s 20ms/step - loss: 4.7962e-04 - accuracy: 1.0000 - val_loss: 1.4884 - val_accuracy: 0.7402

Epoch 124/150
782/782 [=====] - 15s 20ms/step - loss: 4.0013e-04 - accuracy: 1.0000 - val_loss: 1.4905 - val_accuracy: 0.7403

Epoch 125/150
782/782 [=====] - 15s 20ms/step - loss: 3.6702e-04 - accuracy: 1.0000 - val_loss: 1.4899 - val_accuracy: 0.7399

Epoch 126/150
782/782 [=====] - 15s 20ms/step - loss: 3.9806e-04 - accuracy: 1.0000 - val_loss: 1.4897 - val_accuracy: 0.7400

Epoch 127/150
782/782 [=====] - 15s 20ms/step - loss: 3.6187e-04 - accuracy: 1.0000 - val_loss: 1.4916 - val_accuracy: 0.7402

Epoch 128/150
782/782 [=====] - 15s 19ms/step - loss: 3.8183e-04 - accuracy: 1.0000 - val_loss: 1.4917 - val_accuracy: 0.7400

Epoch 129/150
782/782 [=====] - 16s 20ms/step - loss: 3.2445e-04 - accuracy: 1.0000 - val_loss: 1.4926 - val_accuracy: 0.7397

Epoch 130/150
782/782 [=====] - 16s 21ms/step - loss: 3.3378e-04 - accuracy: 1.0000 - val_loss: 1.4962 - val_accuracy: 0.7403

Epoch 131/150
782/782 [=====] - 16s 20ms/step - loss: 3.2361e-04 - accuracy: 1.0000 - val_loss: 1.4954 - val_accuracy: 0.7409

Epoch 132/150
782/782 [=====] - 16s 20ms/step - loss: 3.3827e-04 - accuracy: 1.0000 - val_loss: 1.4975 - val_accuracy: 0.7411

Epoch 133/150
782/782 [=====] - 16s 20ms/step - loss: 3.2768e-04 - accuracy: 1.0000 - val_loss: 1.4976 - val_accuracy: 0.7421

Epoch 134/150
782/782 [=====] - 16s 20ms/step - loss: 3.4196e-04 - accuracy: 1.0000 - val_loss: 1.4994 - val_accuracy: 0.7403

Epoch 135/150
782/782 [=====] - 16s 20ms/step - loss: 4.0048e-04 - accuracy: 1.0000 - val_loss: 1.5041 - val_accuracy: 0.7386

Epoch 136/150
782/782 [=====] - 16s 20ms/step - loss: 3.2058e-04 - accuracy: 1.0000 - val_loss: 1.5023 - val_accuracy: 0.7396

Epoch 137/150
782/782 [=====] - 16s 20ms/step - loss: 3.3128e-04 - accuracy: 1.0000 - val_loss: 1.5015 - val_accuracy: 0.7396

Epoch 138/150
782/782 [=====] - 16s 20ms/step - loss: 2.8696e-04 - accuracy: 1.0000 - val_loss: 1.5042 - val_accuracy: 0.7402

Epoch 139/150
782/782 [=====] - 16s 20ms/step - loss: 2.8008e-04 - accuracy: 1.0000 - val_loss: 1.5039 - val_accuracy: 0.7391

Epoch 140/150
782/782 [=====] - 16s 20ms/step - loss: 2.8299e-04 - accuracy: 1.0000 - val_loss: 1.5054 - val_accuracy: 0.7411

Epoch 141/150
782/782 [=====] - 16s 20ms/step - loss: 3.2932e-04 - accuracy: 1.0000 - val_loss: 1.5074 - val_accuracy: 0.7389

Epoch 142/150
782/782 [=====] - 16s 20ms/step - loss: 4.5912e-04 - accuracy: 1.0000 - val_loss: 1.5127 - val_accuracy: 0.7405

Epoch 143/150
782/782 [=====] - 16s 20ms/step - loss: 3.5665e-04 - accuracy: 1.0000 - val_loss: 1.5149 - val_accuracy: 0.7391

Epoch 144/150
782/782 [=====] - 16s 20ms/step - loss: 9.0228e-04 - accuracy: 0.9999 - val_loss: 1.5133 - val_accuracy: 0.7410

Epoch 145/150
782/782 [=====] - 16s 20ms/step - loss: 4.2301e-04 - accuracy: 1.0000 - val_loss: 1.5154 - val_accuracy: 0.7423

Epoch 146/150
782/782 [=====] - 15s 20ms/step - loss: 3.5230e-04 - accuracy: 1.0000 - val_loss: 1.5133 - val_accuracy: 0.7414

```

Epoch 147/150
782/782 [=====] - 16s 20ms/step - loss: 3.7816e-04 -
accuracy: 1.0000 - val_loss: 1.5151 - val_accuracy: 0.7411
Epoch 148/150
782/782 [=====] - 16s 20ms/step - loss: 6.1537e-04 -
accuracy: 0.9999 - val_loss: 1.5286 - val_accuracy: 0.7413
Epoch 149/150
782/782 [=====] - 16s 20ms/step - loss: 3.9483e-04 -
accuracy: 1.0000 - val_loss: 1.5237 - val_accuracy: 0.7394
Epoch 150/150
782/782 [=====] - 15s 20ms/step - loss: 3.2838e-04 -
accuracy: 1.0000 - val_loss: 1.5205 - val_accuracy: 0.7408
time: 2317.193153142929

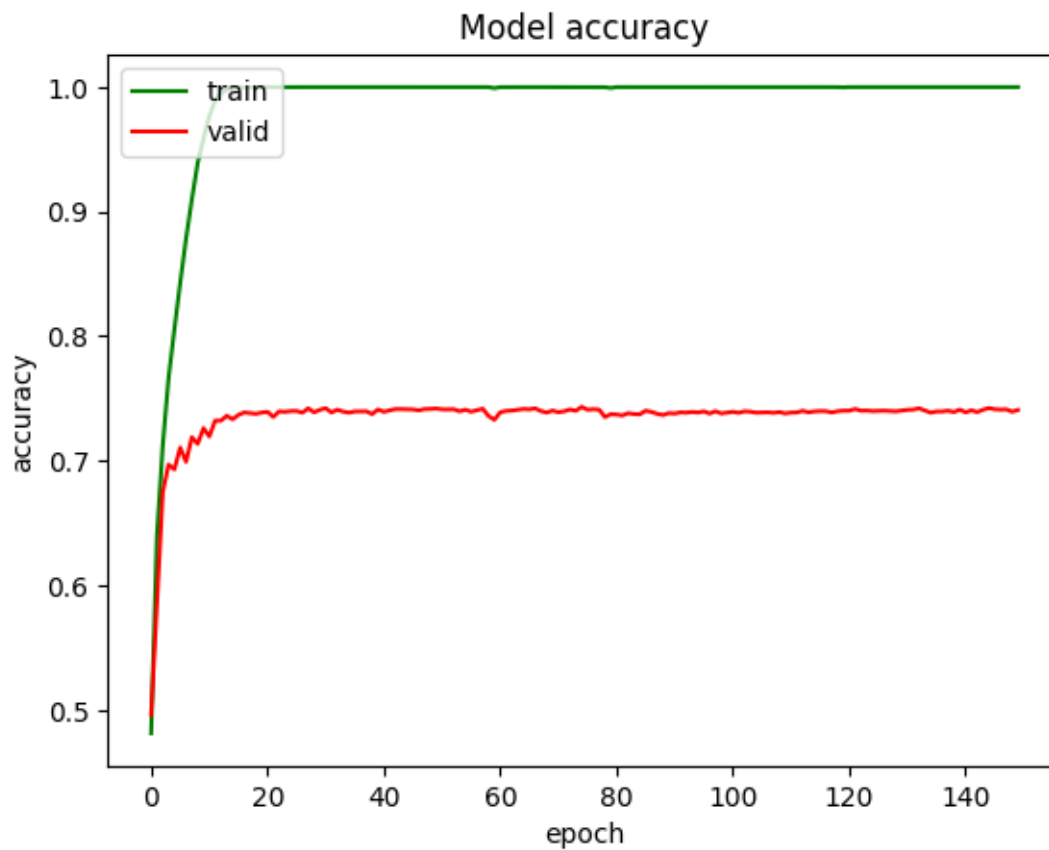
```

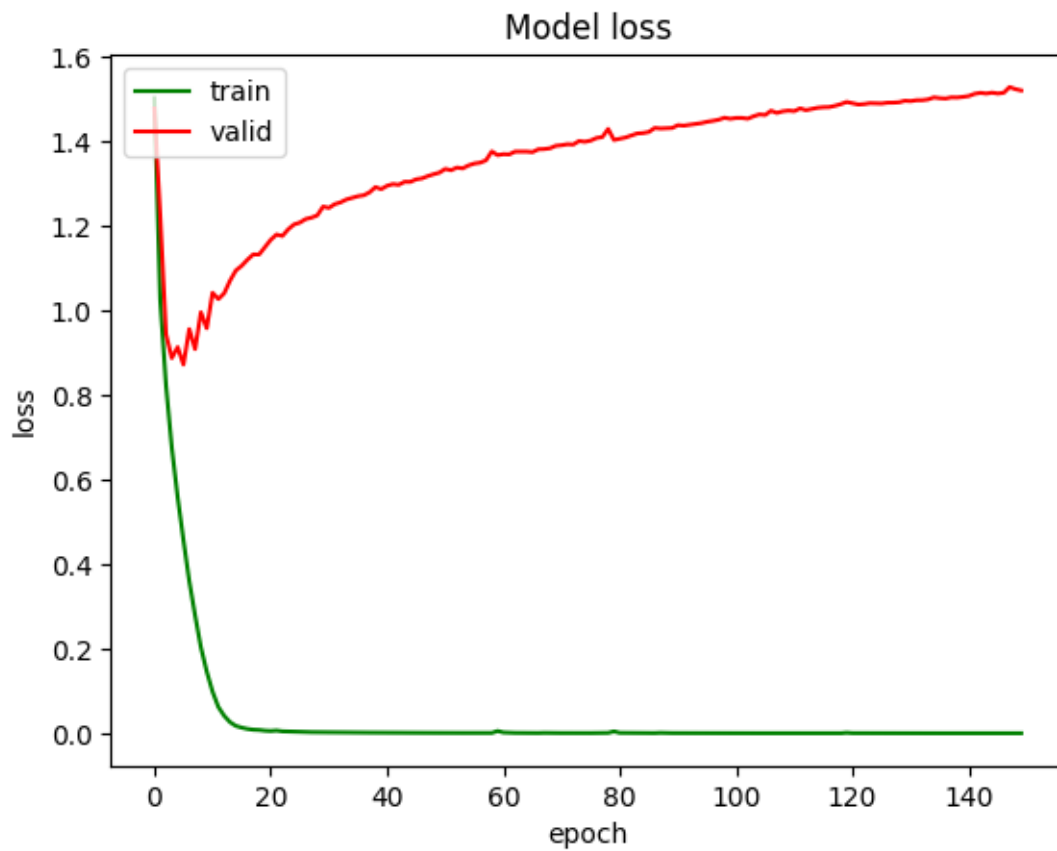
Tym razem niewiele zwiększyliśmy rozmiar modelu (szczególnie pod względem liczby parametrów, szczególnie pod względem liczby parametrów do uczenia), tymczasem czas trawania uczenia wzrósł około o tyle samo co poprzednim razem.

Dalej napotykamy ten sam problem - overfitting. Nasz model praktycznie doskonale zapamiętał zbiór treningowy, lecz nie daje aż takiej dobrej rady ze zbiorem walidacyjnym.

Model daje sobie radę nieznacznie lepiej niż poprzednia wersja bez normalizacji batchy.

```
[ ]: print_history(hisotry.history)
```





2 DROPOUT

```
[ ]: model = tf.keras.Sequential(
    block(channels=20, activation='relu', input_shape=train_images[0].shape,
    ↪ batch_normalisation=True, dropout=0.1) +
    block(channels=40, activation='relu', batch_normalisation=True, dropout=0.
    ↪ 2) +
    block(channels=80, activation='relu', batch_normalisation=True, dropout=0.
    ↪ 3) +
    block(channels=160, activation='relu', batch_normalisation=True, dropout=0.
    ↪ 4) + [
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dense(units=10, activation='softmax')
])
```

```
[ ]: model.summary()
```

Model: "sequential_19"

Layer (type)	Output Shape	Param #
conv2d_113 (Conv2D)	(None, 32, 32, 20)	560
batch_normalization_38 (Batch Normalization)	(None, 32, 32, 20)	80
conv2d_114 (Conv2D)	(None, 32, 32, 20)	3620
batch_normalization_39 (Batch Normalization)	(None, 32, 32, 20)	80
max_pooling2d_56 (MaxPooling2D)	(None, 16, 16, 20)	0
dropout_4 (Dropout)	(None, 16, 16, 20)	0
conv2d_115 (Conv2D)	(None, 16, 16, 40)	7240
batch_normalization_40 (Batch Normalization)	(None, 16, 16, 40)	160
conv2d_116 (Conv2D)	(None, 16, 16, 40)	14440
batch_normalization_41 (Batch Normalization)	(None, 16, 16, 40)	160
max_pooling2d_57 (MaxPooling2D)	(None, 8, 8, 40)	0
dropout_5 (Dropout)	(None, 8, 8, 40)	0
conv2d_117 (Conv2D)	(None, 8, 8, 80)	28880
batch_normalization_42 (Batch Normalization)	(None, 8, 8, 80)	320
conv2d_118 (Conv2D)	(None, 8, 8, 80)	57680
batch_normalization_43 (Batch Normalization)	(None, 8, 8, 80)	320
max_pooling2d_58 (MaxPooling2D)	(None, 4, 4, 80)	0
dropout_6 (Dropout)	(None, 4, 4, 80)	0

conv2d_119 (Conv2D)	(None, 4, 4, 160)	115360
batch_normalization_44 (Batch Normalization)	(None, 4, 4, 160)	640
conv2d_120 (Conv2D)	(None, 4, 4, 160)	230560
batch_normalization_45 (Batch Normalization)	(None, 4, 4, 160)	640
max_pooling2d_59 (MaxPooling2D)	(None, 2, 2, 160)	0
dropout_7 (Dropout)	(None, 2, 2, 160)	0
flatten_19 (Flatten)	(None, 640)	0
dense_19 (Dense)	(None, 10)	6410

```

=====
Total params: 467,150
Trainable params: 465,950
Non-trainable params: 1,200
-----

```

```
[ ]: model.compile(
    optimizer=tf.keras.optimizers.SGD(learning_rate=0.001, momentum=0.9),
    loss=tf.keras.losses.CategoricalCrossentropy(),
    metrics=['accuracy']
)
```

```
[ ]: start_time = time.time()
history = model.fit(
    x=train_images,
    y=train_labels,
    batch_size=64,
    epochs=150,
    verbose=1,
    validation_data=(val_images, val_labels)
)
print(f"time: {time.time()-start_time}")
```

Epoch 1/150

```

2022-10-27 22:44:48.449655: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

```

782/782 [=====] - ETA: 0s - loss: 1.8990 - accuracy:

0.3744

2022-10-27 22:45:04.592333: I

tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]

Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - 19s 22ms/step - loss: 1.8990 -
accuracy: 0.3744 - val_loss: 1.4651 - val_accuracy: 0.4627

Epoch 2/150

782/782 [=====] - 16s 21ms/step - loss: 1.4158 -
accuracy: 0.4985 - val_loss: 1.3939 - val_accuracy: 0.5429

Epoch 3/150

782/782 [=====] - 16s 21ms/step - loss: 1.2437 -
accuracy: 0.5579 - val_loss: 1.1554 - val_accuracy: 0.5926

Epoch 4/150

782/782 [=====] - 16s 21ms/step - loss: 1.1185 -
accuracy: 0.6037 - val_loss: 0.9758 - val_accuracy: 0.6532

Epoch 5/150

782/782 [=====] - 16s 21ms/step - loss: 1.0304 -
accuracy: 0.6367 - val_loss: 0.9476 - val_accuracy: 0.6678

Epoch 6/150

782/782 [=====] - 16s 21ms/step - loss: 0.9492 -
accuracy: 0.6636 - val_loss: 0.8802 - val_accuracy: 0.6922

Epoch 7/150

782/782 [=====] - 16s 21ms/step - loss: 0.8946 -
accuracy: 0.6835 - val_loss: 0.8773 - val_accuracy: 0.6903

Epoch 8/150

782/782 [=====] - 16s 21ms/step - loss: 0.8413 -
accuracy: 0.7043 - val_loss: 0.8147 - val_accuracy: 0.7205

Epoch 9/150

782/782 [=====] - 16s 21ms/step - loss: 0.7983 -
accuracy: 0.7185 - val_loss: 0.7859 - val_accuracy: 0.7239

Epoch 10/150

782/782 [=====] - 16s 21ms/step - loss: 0.7600 -
accuracy: 0.7330 - val_loss: 0.7492 - val_accuracy: 0.7422

Epoch 11/150

782/782 [=====] - 16s 21ms/step - loss: 0.7313 -
accuracy: 0.7431 - val_loss: 0.7449 - val_accuracy: 0.7389

Epoch 12/150

782/782 [=====] - 16s 21ms/step - loss: 0.7017 -
accuracy: 0.7533 - val_loss: 0.6893 - val_accuracy: 0.7587

Epoch 13/150

782/782 [=====] - 16s 21ms/step - loss: 0.6701 -
accuracy: 0.7643 - val_loss: 0.7089 - val_accuracy: 0.7530

Epoch 14/150

782/782 [=====] - 16s 21ms/step - loss: 0.6522 -
accuracy: 0.7700 - val_loss: 0.6696 - val_accuracy: 0.7634

Epoch 15/150

782/782 [=====] - 16s 21ms/step - loss: 0.6301 -

accuracy: 0.7817 - val_loss: 0.6656 - val_accuracy: 0.7686
 Epoch 16/150
 782/782 [=====] - 16s 21ms/step - loss: 0.6112 -
 accuracy: 0.7865 - val_loss: 0.6751 - val_accuracy: 0.7702
 Epoch 17/150
 782/782 [=====] - 16s 21ms/step - loss: 0.5937 -
 accuracy: 0.7915 - val_loss: 0.6802 - val_accuracy: 0.7691
 Epoch 18/150
 782/782 [=====] - 16s 21ms/step - loss: 0.5798 -
 accuracy: 0.7948 - val_loss: 0.6958 - val_accuracy: 0.7619
 Epoch 19/150
 782/782 [=====] - 16s 21ms/step - loss: 0.5654 -
 accuracy: 0.7999 - val_loss: 0.6216 - val_accuracy: 0.7871
 Epoch 20/150
 782/782 [=====] - 17s 22ms/step - loss: 0.5503 -
 accuracy: 0.8048 - val_loss: 0.6079 - val_accuracy: 0.7878
 Epoch 21/150
 782/782 [=====] - 17s 21ms/step - loss: 0.5357 -
 accuracy: 0.8121 - val_loss: 0.6279 - val_accuracy: 0.7867
 Epoch 22/150
 782/782 [=====] - 16s 21ms/step - loss: 0.5220 -
 accuracy: 0.8160 - val_loss: 0.6684 - val_accuracy: 0.7692
 Epoch 23/150
 782/782 [=====] - 16s 21ms/step - loss: 0.5098 -
 accuracy: 0.8205 - val_loss: 0.6143 - val_accuracy: 0.7863
 Epoch 24/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4974 -
 accuracy: 0.8243 - val_loss: 0.7947 - val_accuracy: 0.7339
 Epoch 25/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4877 -
 accuracy: 0.8272 - val_loss: 0.6189 - val_accuracy: 0.7846
 Epoch 26/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4775 -
 accuracy: 0.8304 - val_loss: 0.6054 - val_accuracy: 0.7933
 Epoch 27/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4627 -
 accuracy: 0.8359 - val_loss: 0.5788 - val_accuracy: 0.8009
 Epoch 28/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4491 -
 accuracy: 0.8423 - val_loss: 0.5798 - val_accuracy: 0.8038
 Epoch 29/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4445 -
 accuracy: 0.8417 - val_loss: 0.5763 - val_accuracy: 0.8054
 Epoch 30/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4350 -
 accuracy: 0.8466 - val_loss: 0.6083 - val_accuracy: 0.7979
 Epoch 31/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4235 -

accuracy: 0.8503 - val_loss: 0.6137 - val_accuracy: 0.7935
 Epoch 32/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4133 -
 accuracy: 0.8540 - val_loss: 0.5828 - val_accuracy: 0.8021
 Epoch 33/150
 782/782 [=====] - 16s 21ms/step - loss: 0.4028 -
 accuracy: 0.8558 - val_loss: 0.5622 - val_accuracy: 0.8127
 Epoch 34/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3984 -
 accuracy: 0.8578 - val_loss: 0.5812 - val_accuracy: 0.8096
 Epoch 35/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3906 -
 accuracy: 0.8604 - val_loss: 0.5554 - val_accuracy: 0.8127
 Epoch 36/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3825 -
 accuracy: 0.8647 - val_loss: 0.6068 - val_accuracy: 0.8013
 Epoch 37/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3684 -
 accuracy: 0.8685 - val_loss: 0.5846 - val_accuracy: 0.8096
 Epoch 38/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3693 -
 accuracy: 0.8690 - val_loss: 0.5911 - val_accuracy: 0.8055
 Epoch 39/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3596 -
 accuracy: 0.8723 - val_loss: 0.5603 - val_accuracy: 0.8161
 Epoch 40/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3527 -
 accuracy: 0.8748 - val_loss: 0.6650 - val_accuracy: 0.7881
 Epoch 41/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3483 -
 accuracy: 0.8764 - val_loss: 0.5559 - val_accuracy: 0.8189
 Epoch 42/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3413 -
 accuracy: 0.8773 - val_loss: 0.5676 - val_accuracy: 0.8158
 Epoch 43/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3306 -
 accuracy: 0.8810 - val_loss: 0.6433 - val_accuracy: 0.7963
 Epoch 44/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3281 -
 accuracy: 0.8821 - val_loss: 0.5580 - val_accuracy: 0.8201
 Epoch 45/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3234 -
 accuracy: 0.8854 - val_loss: 0.5735 - val_accuracy: 0.8143
 Epoch 46/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3132 -
 accuracy: 0.8883 - val_loss: 0.5818 - val_accuracy: 0.8123
 Epoch 47/150
 782/782 [=====] - 16s 21ms/step - loss: 0.3056 -

accuracy: 0.8930 - val_loss: 0.5786 - val_accuracy: 0.8158
Epoch 48/150
782/782 [=====] - 16s 21ms/step - loss: 0.3036 -
accuracy: 0.8911 - val_loss: 0.5877 - val_accuracy: 0.8143
Epoch 49/150
782/782 [=====] - 16s 21ms/step - loss: 0.3003 -
accuracy: 0.8917 - val_loss: 0.5655 - val_accuracy: 0.8198
Epoch 50/150
782/782 [=====] - 16s 21ms/step - loss: 0.2929 -
accuracy: 0.8936 - val_loss: 0.5762 - val_accuracy: 0.8207
Epoch 51/150
782/782 [=====] - 16s 21ms/step - loss: 0.2836 -
accuracy: 0.8989 - val_loss: 0.5850 - val_accuracy: 0.8171
Epoch 52/150
782/782 [=====] - 17s 21ms/step - loss: 0.2851 -
accuracy: 0.8985 - val_loss: 0.5692 - val_accuracy: 0.8229
Epoch 53/150
782/782 [=====] - 16s 21ms/step - loss: 0.2750 -
accuracy: 0.9010 - val_loss: 0.5629 - val_accuracy: 0.8237
Epoch 54/150
782/782 [=====] - 16s 21ms/step - loss: 0.2677 -
accuracy: 0.9041 - val_loss: 0.5564 - val_accuracy: 0.8249
Epoch 55/150
782/782 [=====] - 16s 21ms/step - loss: 0.2640 -
accuracy: 0.9063 - val_loss: 0.5680 - val_accuracy: 0.8238
Epoch 56/150
782/782 [=====] - 16s 21ms/step - loss: 0.2577 -
accuracy: 0.9084 - val_loss: 0.5784 - val_accuracy: 0.8212
Epoch 57/150
782/782 [=====] - 16s 21ms/step - loss: 0.2577 -
accuracy: 0.9061 - val_loss: 0.6005 - val_accuracy: 0.8227
Epoch 58/150
782/782 [=====] - 16s 21ms/step - loss: 0.2536 -
accuracy: 0.9082 - val_loss: 0.5981 - val_accuracy: 0.8183
Epoch 59/150
782/782 [=====] - 16s 21ms/step - loss: 0.2458 -
accuracy: 0.9111 - val_loss: 0.5740 - val_accuracy: 0.8280
Epoch 60/150
782/782 [=====] - 16s 21ms/step - loss: 0.2474 -
accuracy: 0.9116 - val_loss: 0.5990 - val_accuracy: 0.8216
Epoch 61/150
782/782 [=====] - 16s 21ms/step - loss: 0.2409 -
accuracy: 0.9141 - val_loss: 0.5788 - val_accuracy: 0.8247
Epoch 62/150
782/782 [=====] - 17s 21ms/step - loss: 0.2339 -
accuracy: 0.9158 - val_loss: 0.5881 - val_accuracy: 0.8240
Epoch 63/150
782/782 [=====] - 16s 21ms/step - loss: 0.2321 -

accuracy: 0.9173 - val_loss: 0.5932 - val_accuracy: 0.8217
 Epoch 64/150
 782/782 [=====] - 17s 21ms/step - loss: 0.2226 -
 accuracy: 0.9200 - val_loss: 0.5915 - val_accuracy: 0.8244
 Epoch 65/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2222 -
 accuracy: 0.9203 - val_loss: 0.6315 - val_accuracy: 0.8182
 Epoch 66/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2216 -
 accuracy: 0.9203 - val_loss: 0.5915 - val_accuracy: 0.8246
 Epoch 67/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2155 -
 accuracy: 0.9219 - val_loss: 0.5932 - val_accuracy: 0.8288
 Epoch 68/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2131 -
 accuracy: 0.9235 - val_loss: 0.6035 - val_accuracy: 0.8220
 Epoch 69/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2082 -
 accuracy: 0.9256 - val_loss: 0.6041 - val_accuracy: 0.8256
 Epoch 70/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2047 -
 accuracy: 0.9263 - val_loss: 0.5720 - val_accuracy: 0.8341
 Epoch 71/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2012 -
 accuracy: 0.9264 - val_loss: 0.5912 - val_accuracy: 0.8285
 Epoch 72/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1988 -
 accuracy: 0.9281 - val_loss: 0.5984 - val_accuracy: 0.8266
 Epoch 73/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1962 -
 accuracy: 0.9296 - val_loss: 0.6007 - val_accuracy: 0.8249
 Epoch 74/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1904 -
 accuracy: 0.9304 - val_loss: 0.6084 - val_accuracy: 0.8259
 Epoch 75/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1877 -
 accuracy: 0.9327 - val_loss: 0.5965 - val_accuracy: 0.8270
 Epoch 76/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1907 -
 accuracy: 0.9311 - val_loss: 0.5955 - val_accuracy: 0.8300
 Epoch 77/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1914 -
 accuracy: 0.9303 - val_loss: 0.5966 - val_accuracy: 0.8305
 Epoch 78/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1823 -
 accuracy: 0.9353 - val_loss: 0.5967 - val_accuracy: 0.8308
 Epoch 79/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1816 -

accuracy: 0.9341 - val_loss: 0.6119 - val_accuracy: 0.8281
 Epoch 80/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1724 -
 accuracy: 0.9380 - val_loss: 0.5993 - val_accuracy: 0.8337
 Epoch 81/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1734 -
 accuracy: 0.9384 - val_loss: 0.5998 - val_accuracy: 0.8337
 Epoch 82/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1714 -
 accuracy: 0.9378 - val_loss: 0.6205 - val_accuracy: 0.8294
 Epoch 83/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1691 -
 accuracy: 0.9396 - val_loss: 0.6179 - val_accuracy: 0.8314
 Epoch 84/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1663 -
 accuracy: 0.9405 - val_loss: 0.6243 - val_accuracy: 0.8268
 Epoch 85/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1657 -
 accuracy: 0.9386 - val_loss: 0.6269 - val_accuracy: 0.8279
 Epoch 86/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1656 -
 accuracy: 0.9409 - val_loss: 0.6287 - val_accuracy: 0.8309
 Epoch 87/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1595 -
 accuracy: 0.9434 - val_loss: 0.6230 - val_accuracy: 0.8321
 Epoch 88/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1606 -
 accuracy: 0.9418 - val_loss: 0.6326 - val_accuracy: 0.8291
 Epoch 89/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1570 -
 accuracy: 0.9419 - val_loss: 0.6486 - val_accuracy: 0.8284
 Epoch 90/150
 782/782 [=====] - 17s 22ms/step - loss: 0.1552 -
 accuracy: 0.9444 - val_loss: 0.6262 - val_accuracy: 0.8346
 Epoch 91/150
 782/782 [=====] - 18s 23ms/step - loss: 0.1495 -
 accuracy: 0.9469 - val_loss: 0.6055 - val_accuracy: 0.8359
 Epoch 92/150
 782/782 [=====] - 18s 22ms/step - loss: 0.1550 -
 accuracy: 0.9442 - val_loss: 0.6244 - val_accuracy: 0.8331
 Epoch 93/150
 782/782 [=====] - 18s 23ms/step - loss: 0.1500 -
 accuracy: 0.9462 - val_loss: 0.6430 - val_accuracy: 0.8324
 Epoch 94/150
 782/782 [=====] - 17s 22ms/step - loss: 0.1449 -
 accuracy: 0.9477 - val_loss: 0.6439 - val_accuracy: 0.8305
 Epoch 95/150
 782/782 [=====] - 17s 22ms/step - loss: 0.1446 -

accuracy: 0.9473 - val_loss: 0.6247 - val_accuracy: 0.8373
 Epoch 96/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1398 -
 accuracy: 0.9493 - val_loss: 0.6412 - val_accuracy: 0.8335
 Epoch 97/150
 782/782 [=====] - 17s 22ms/step - loss: 0.1382 -
 accuracy: 0.9497 - val_loss: 0.6383 - val_accuracy: 0.8340
 Epoch 98/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1401 -
 accuracy: 0.9498 - val_loss: 0.6261 - val_accuracy: 0.8355
 Epoch 99/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1347 -
 accuracy: 0.9511 - val_loss: 0.6511 - val_accuracy: 0.8323
 Epoch 100/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1369 -
 accuracy: 0.9504 - val_loss: 0.6400 - val_accuracy: 0.8361
 Epoch 101/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1367 -
 accuracy: 0.9500 - val_loss: 0.6255 - val_accuracy: 0.8355
 Epoch 102/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1309 -
 accuracy: 0.9528 - val_loss: 0.6410 - val_accuracy: 0.8322
 Epoch 103/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1295 -
 accuracy: 0.9531 - val_loss: 0.6709 - val_accuracy: 0.8290
 Epoch 104/150
 782/782 [=====] - 17s 22ms/step - loss: 0.1309 -
 accuracy: 0.9532 - val_loss: 0.6499 - val_accuracy: 0.8347
 Epoch 105/150
 782/782 [=====] - 18s 23ms/step - loss: 0.1297 -
 accuracy: 0.9523 - val_loss: 0.6432 - val_accuracy: 0.8347
 Epoch 106/150
 782/782 [=====] - 17s 22ms/step - loss: 0.1243 -
 accuracy: 0.9554 - val_loss: 0.6627 - val_accuracy: 0.8310
 Epoch 107/150
 782/782 [=====] - 17s 22ms/step - loss: 0.1251 -
 accuracy: 0.9554 - val_loss: 0.6698 - val_accuracy: 0.8321
 Epoch 108/150
 782/782 [=====] - 18s 23ms/step - loss: 0.1251 -
 accuracy: 0.9543 - val_loss: 0.6586 - val_accuracy: 0.8369
 Epoch 109/150
 782/782 [=====] - 17s 22ms/step - loss: 0.1251 -
 accuracy: 0.9549 - val_loss: 0.6619 - val_accuracy: 0.8355
 Epoch 110/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1218 -
 accuracy: 0.9565 - val_loss: 0.6598 - val_accuracy: 0.8362
 Epoch 111/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1206 -

accuracy: 0.9564 - val_loss: 0.6526 - val_accuracy: 0.8399
 Epoch 112/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1203 -
 accuracy: 0.9573 - val_loss: 0.6737 - val_accuracy: 0.8294
 Epoch 113/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1170 -
 accuracy: 0.9585 - val_loss: 0.6630 - val_accuracy: 0.8339
 Epoch 114/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1161 -
 accuracy: 0.9580 - val_loss: 0.6645 - val_accuracy: 0.8329
 Epoch 115/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1174 -
 accuracy: 0.9585 - val_loss: 0.6449 - val_accuracy: 0.8379
 Epoch 116/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1151 -
 accuracy: 0.9593 - val_loss: 0.6810 - val_accuracy: 0.8344
 Epoch 117/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1131 -
 accuracy: 0.9605 - val_loss: 0.6488 - val_accuracy: 0.8409
 Epoch 118/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1157 -
 accuracy: 0.9575 - val_loss: 0.6716 - val_accuracy: 0.8336
 Epoch 119/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1065 -
 accuracy: 0.9617 - val_loss: 0.6900 - val_accuracy: 0.8329
 Epoch 120/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1131 -
 accuracy: 0.9596 - val_loss: 0.6579 - val_accuracy: 0.8392
 Epoch 121/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1125 -
 accuracy: 0.9594 - val_loss: 0.6431 - val_accuracy: 0.8380
 Epoch 122/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1054 -
 accuracy: 0.9625 - val_loss: 0.6719 - val_accuracy: 0.8354
 Epoch 123/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1032 -
 accuracy: 0.9635 - val_loss: 0.6594 - val_accuracy: 0.8398
 Epoch 124/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1012 -
 accuracy: 0.9633 - val_loss: 0.7021 - val_accuracy: 0.8309
 Epoch 125/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1053 -
 accuracy: 0.9627 - val_loss: 0.6799 - val_accuracy: 0.8365
 Epoch 126/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1030 -
 accuracy: 0.9624 - val_loss: 0.6852 - val_accuracy: 0.8364
 Epoch 127/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1047 -

```

accuracy: 0.9620 - val_loss: 0.6837 - val_accuracy: 0.8352
Epoch 128/150
782/782 [=====] - 17s 21ms/step - loss: 0.1006 -
accuracy: 0.9639 - val_loss: 0.7198 - val_accuracy: 0.8313
Epoch 129/150
782/782 [=====] - 16s 21ms/step - loss: 0.1012 -
accuracy: 0.9645 - val_loss: 0.6773 - val_accuracy: 0.8365
Epoch 130/150
782/782 [=====] - 17s 21ms/step - loss: 0.0986 -
accuracy: 0.9644 - val_loss: 0.6756 - val_accuracy: 0.8366
Epoch 131/150
782/782 [=====] - 16s 21ms/step - loss: 0.0956 -
accuracy: 0.9662 - val_loss: 0.6828 - val_accuracy: 0.8353
Epoch 132/150
782/782 [=====] - 17s 21ms/step - loss: 0.0966 -
accuracy: 0.9655 - val_loss: 0.6830 - val_accuracy: 0.8383
Epoch 133/150
782/782 [=====] - 16s 21ms/step - loss: 0.0928 -
accuracy: 0.9675 - val_loss: 0.6769 - val_accuracy: 0.8392
Epoch 134/150
782/782 [=====] - 16s 21ms/step - loss: 0.0962 -
accuracy: 0.9661 - val_loss: 0.7006 - val_accuracy: 0.8335
Epoch 135/150
782/782 [=====] - 16s 21ms/step - loss: 0.0979 -
accuracy: 0.9642 - val_loss: 0.6854 - val_accuracy: 0.8393
Epoch 136/150
782/782 [=====] - 16s 21ms/step - loss: 0.0940 -
accuracy: 0.9660 - val_loss: 0.6919 - val_accuracy: 0.8375
Epoch 137/150
782/782 [=====] - 16s 21ms/step - loss: 0.0930 -
accuracy: 0.9671 - val_loss: 0.6875 - val_accuracy: 0.8371
Epoch 138/150
782/782 [=====] - 16s 21ms/step - loss: 0.0927 -
accuracy: 0.9668 - val_loss: 0.6994 - val_accuracy: 0.8343
Epoch 139/150
782/782 [=====] - 17s 21ms/step - loss: 0.0873 -
accuracy: 0.9689 - val_loss: 0.6925 - val_accuracy: 0.8374
Epoch 140/150
782/782 [=====] - 16s 21ms/step - loss: 0.0931 -
accuracy: 0.9670 - val_loss: 0.7042 - val_accuracy: 0.8342
Epoch 141/150
782/782 [=====] - 17s 21ms/step - loss: 0.0902 -
accuracy: 0.9681 - val_loss: 0.7459 - val_accuracy: 0.8270
Epoch 142/150
782/782 [=====] - 16s 21ms/step - loss: 0.0921 -
accuracy: 0.9671 - val_loss: 0.7143 - val_accuracy: 0.8340
Epoch 143/150
782/782 [=====] - 16s 21ms/step - loss: 0.0899 -

```

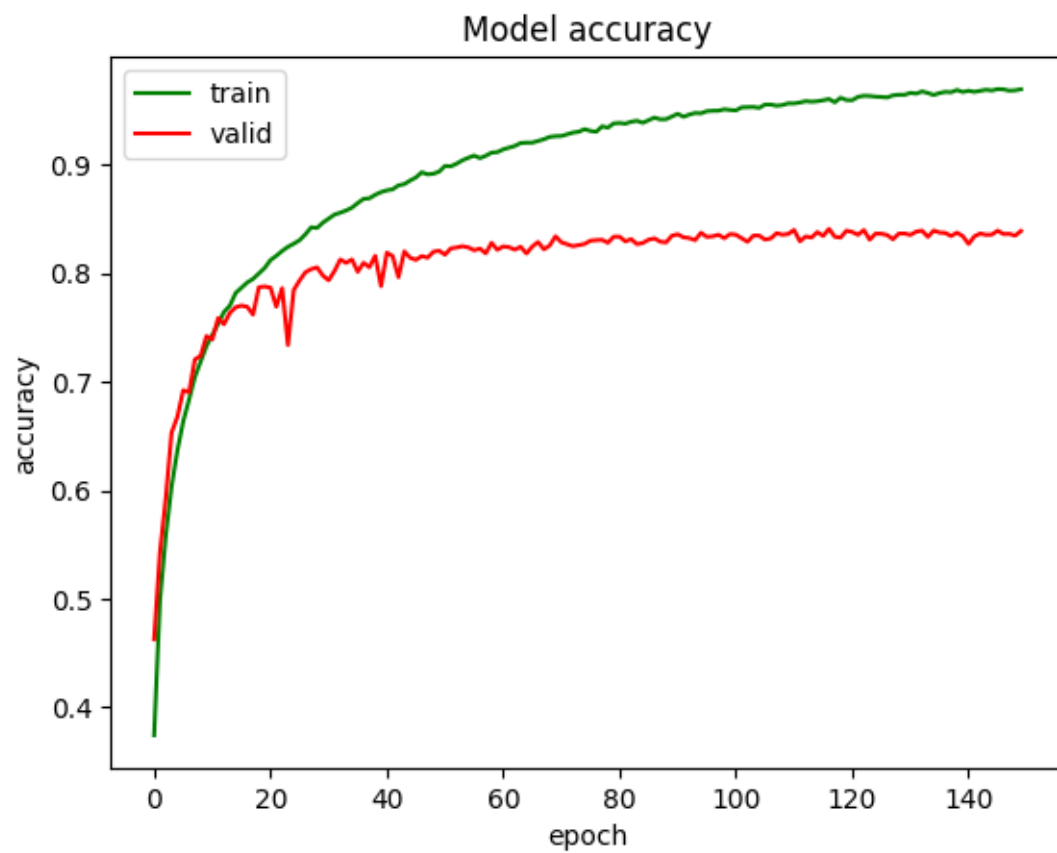
```
accuracy: 0.9680 - val_loss: 0.6914 - val_accuracy: 0.8366
Epoch 144/150
782/782 [=====] - 17s 21ms/step - loss: 0.0875 -
accuracy: 0.9691 - val_loss: 0.6996 - val_accuracy: 0.8354
Epoch 145/150
782/782 [=====] - 16s 21ms/step - loss: 0.0880 -
accuracy: 0.9684 - val_loss: 0.7176 - val_accuracy: 0.8356
Epoch 146/150
782/782 [=====] - 16s 21ms/step - loss: 0.0852 -
accuracy: 0.9697 - val_loss: 0.7114 - val_accuracy: 0.8392
Epoch 147/150
782/782 [=====] - 16s 21ms/step - loss: 0.0872 -
accuracy: 0.9694 - val_loss: 0.7167 - val_accuracy: 0.8361
Epoch 148/150
782/782 [=====] - 16s 21ms/step - loss: 0.0888 -
accuracy: 0.9682 - val_loss: 0.7049 - val_accuracy: 0.8363
Epoch 149/150
782/782 [=====] - 16s 21ms/step - loss: 0.0874 -
accuracy: 0.9685 - val_loss: 0.7218 - val_accuracy: 0.8347
Epoch 150/150
782/782 [=====] - 16s 21ms/step - loss: 0.0855 -
accuracy: 0.9695 - val_loss: 0.7038 - val_accuracy: 0.8389
time: 2476.677229642868
```

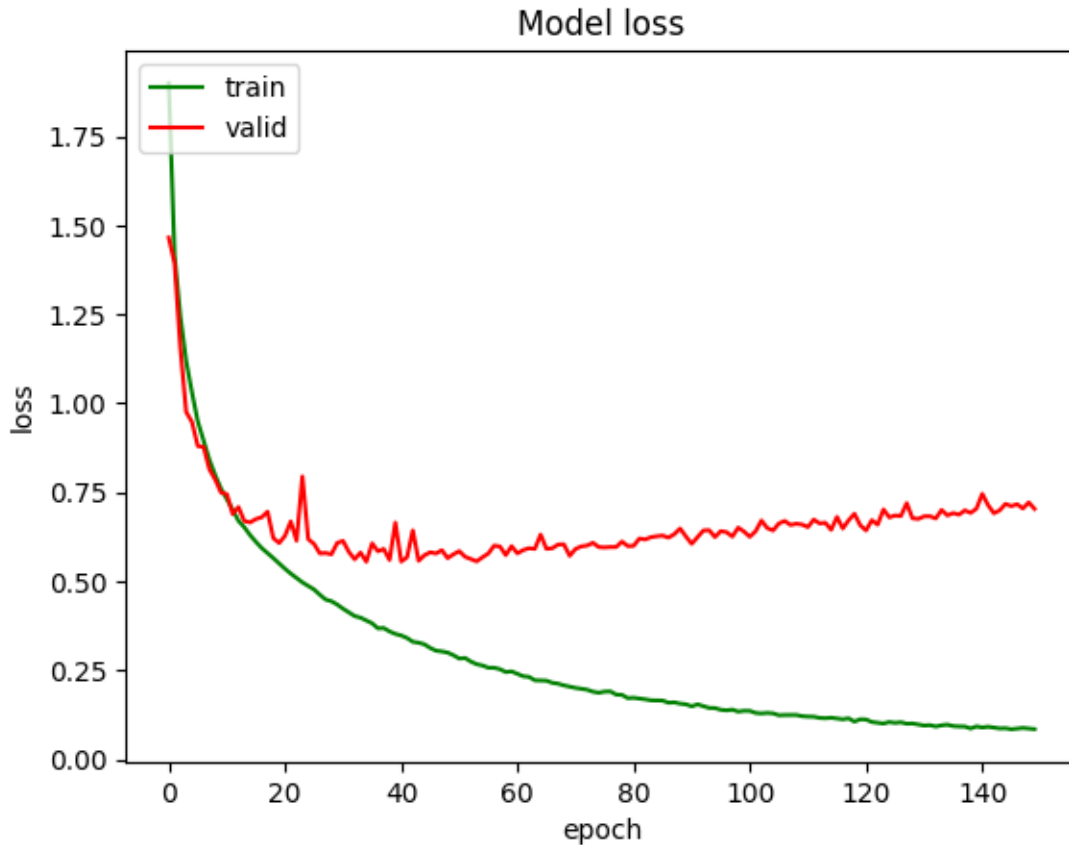
Tym razem, liczyło się niewiele dłużej co poprzednim razem (batch normalization).

Model uczył się około 40 min.

Widzimy, że tym razem model nie jest zbyt przeluczony i daje rozsądne wyniki.

```
[ ]: print_history(history.history)
```





3 Mind the gap

```
[ ]: model = tf.keras.Sequential(
    block(channels=20, activation='relu', input_shape=(None, None, 3),
    ↪batch_normalisation=True, dropout=0.1) +
    block(channels=40, activation='relu', batch_normalisation=True, dropout=0.
    ↪2) +
    block(channels=80, activation='relu', batch_normalisation=True, dropout=0.
    ↪3) +
    block(channels=160, activation='relu', batch_normalisation=True, dropout=0.
    ↪4, gap=True) + [
    tf.keras.layers.Dense(units=10, activation='softmax')
])
```

Ze względu na global average pooling, któremu ustawiłem keepdims na false, nie potrzebuję warstwy spłaszczającej, ponieważ warstwa GAP już to robi, zwłaszcza z parametrem keepdims=False. Dodatkowo ustawiłem input_shape na (None, None, 3), tak żeby sieć przyjmowała obrazy różnego rozmiaru na wejście.

```
[ ]: model.summary()
```

Model: "sequential_29"

Layer (type)	Output Shape	Param #
conv2d_193 (Conv2D)	(None, None, None, 20)	560
batch_normalization_118 (Batch Normalization)	(None, None, None, 20)	80
conv2d_194 (Conv2D)	(None, None, None, 20)	3620
batch_normalization_119 (Batch Normalization)	(None, None, None, 20)	80
max_pooling2d_87 (MaxPooling2D)	(None, None, None, 20)	0
dropout_44 (Dropout)	(None, None, None, 20)	0
conv2d_195 (Conv2D)	(None, None, None, 40)	7240
batch_normalization_120 (Batch Normalization)	(None, None, None, 40)	160
conv2d_196 (Conv2D)	(None, None, None, 40)	14440
batch_normalization_121 (Batch Normalization)	(None, None, None, 40)	160
max_pooling2d_88 (MaxPooling2D)	(None, None, None, 40)	0
dropout_45 (Dropout)	(None, None, None, 40)	0
conv2d_197 (Conv2D)	(None, None, None, 80)	28880
batch_normalization_122 (Batch Normalization)	(None, None, None, 80)	320
conv2d_198 (Conv2D)	(None, None, None, 80)	57680
batch_normalization_123 (Batch Normalization)	(None, None, None, 80)	320
max_pooling2d_89 (MaxPooling2D)	(None, None, None, 80)	0

dropout_46 (Dropout)	(None, None, None, 80)	0
conv2d_199 (Conv2D)	(None, None, None, 160)	115360
batch_normalization_124 (Batch Normalization)	(None, None, None, 160)	640
conv2d_200 (Conv2D)	(None, None, None, 160)	230560
batch_normalization_125 (Batch Normalization)	(None, None, None, 160)	640
global_average_pooling2d_9 (GlobalAveragePooling2D)	(None, 160)	0
dropout_47 (Dropout)	(None, 160)	0
dense_29 (Dense)	(None, 10)	1610

```

=====
Total params: 462,350
Trainable params: 461,150
Non-trainable params: 1,200
-----

```

```
[ ]: model.compile(
    optimizer=tf.keras.optimizers.SGD(learning_rate=0.001, momentum=0.9),
    loss=tf.keras.losses.CategoricalCrossentropy(),
    metrics=['accuracy']
)
```

```
[ ]: start_time = time.time()
history = model.fit(
    x=train_images,
    y=train_labels,
    batch_size=64,
    epochs=150,
    verbose=1,
    validation_data=(val_images, val_labels)
)
print(f"time: {time.time()-start_time}")
```

Epoch 1/150

```

2022-10-28 01:01:37.073854: I
tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]
Plugin optimizer for device_type GPU is enabled.

```


782/782 [=====] - ETA: 0s - loss: 1.6360 - accuracy: 0.3979

2022-10-28 01:01:53.844152: I

tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114]

Plugin optimizer for device_type GPU is enabled.

782/782 [=====] - 19s 23ms/step - loss: 1.6360 - accuracy: 0.3979 - val_loss: 1.6287 - val_accuracy: 0.4209

Epoch 2/150

782/782 [=====] - 17s 22ms/step - loss: 1.2431 - accuracy: 0.5536 - val_loss: 1.1602 - val_accuracy: 0.5889

Epoch 3/150

782/782 [=====] - 17s 22ms/step - loss: 1.1011 - accuracy: 0.6058 - val_loss: 1.1346 - val_accuracy: 0.6009

Epoch 4/150

782/782 [=====] - 17s 22ms/step - loss: 1.0060 - accuracy: 0.6435 - val_loss: 0.9870 - val_accuracy: 0.6543

Epoch 5/150

782/782 [=====] - 17s 22ms/step - loss: 0.9500 - accuracy: 0.6629 - val_loss: 0.9658 - val_accuracy: 0.6568

Epoch 6/150

782/782 [=====] - 17s 22ms/step - loss: 0.8977 - accuracy: 0.6827 - val_loss: 0.8889 - val_accuracy: 0.6836

Epoch 7/150

782/782 [=====] - 17s 22ms/step - loss: 0.8560 - accuracy: 0.6965 - val_loss: 0.8075 - val_accuracy: 0.7161

Epoch 8/150

782/782 [=====] - 17s 22ms/step - loss: 0.8168 - accuracy: 0.7106 - val_loss: 0.8749 - val_accuracy: 0.6956

Epoch 9/150

782/782 [=====] - 17s 21ms/step - loss: 0.7829 - accuracy: 0.7256 - val_loss: 0.7739 - val_accuracy: 0.7298

Epoch 10/150

782/782 [=====] - 17s 21ms/step - loss: 0.7465 - accuracy: 0.7382 - val_loss: 0.7497 - val_accuracy: 0.7417

Epoch 11/150

782/782 [=====] - 16s 21ms/step - loss: 0.7190 - accuracy: 0.7479 - val_loss: 0.7074 - val_accuracy: 0.7540

Epoch 12/150

782/782 [=====] - 17s 21ms/step - loss: 0.6890 - accuracy: 0.7594 - val_loss: 0.6909 - val_accuracy: 0.7583

Epoch 13/150

782/782 [=====] - 17s 22ms/step - loss: 0.6736 - accuracy: 0.7625 - val_loss: 0.7068 - val_accuracy: 0.7590

Epoch 14/150

782/782 [=====] - 17s 22ms/step - loss: 0.6486 - accuracy: 0.7740 - val_loss: 0.6648 - val_accuracy: 0.7730

Epoch 15/150

782/782 [=====] - 17s 22ms/step - loss: 0.6325 - accuracy: 0.7786 - val_loss: 0.6496 - val_accuracy: 0.7748
Epoch 16/150
782/782 [=====] - 17s 21ms/step - loss: 0.6167 - accuracy: 0.7831 - val_loss: 0.6844 - val_accuracy: 0.7616
Epoch 17/150
782/782 [=====] - 17s 21ms/step - loss: 0.5986 - accuracy: 0.7907 - val_loss: 0.6528 - val_accuracy: 0.7745
Epoch 18/150
782/782 [=====] - 16s 21ms/step - loss: 0.5853 - accuracy: 0.7950 - val_loss: 0.6369 - val_accuracy: 0.7833
Epoch 19/150
782/782 [=====] - 17s 21ms/step - loss: 0.5681 - accuracy: 0.8018 - val_loss: 0.6644 - val_accuracy: 0.7740
Epoch 20/150
782/782 [=====] - 17s 21ms/step - loss: 0.5523 - accuracy: 0.8057 - val_loss: 0.6154 - val_accuracy: 0.7903
Epoch 21/150
782/782 [=====] - 17s 21ms/step - loss: 0.5418 - accuracy: 0.8105 - val_loss: 0.6065 - val_accuracy: 0.7931
Epoch 22/150
782/782 [=====] - 16s 21ms/step - loss: 0.5378 - accuracy: 0.8122 - val_loss: 0.6071 - val_accuracy: 0.7916
Epoch 23/150
782/782 [=====] - 17s 21ms/step - loss: 0.5238 - accuracy: 0.8171 - val_loss: 0.5964 - val_accuracy: 0.7957
Epoch 24/150
782/782 [=====] - 17s 21ms/step - loss: 0.5111 - accuracy: 0.8207 - val_loss: 0.5653 - val_accuracy: 0.8075
Epoch 25/150
782/782 [=====] - 16s 21ms/step - loss: 0.4971 - accuracy: 0.8277 - val_loss: 0.6111 - val_accuracy: 0.7894
Epoch 26/150
782/782 [=====] - 16s 21ms/step - loss: 0.4917 - accuracy: 0.8278 - val_loss: 0.6027 - val_accuracy: 0.7995
Epoch 27/150
782/782 [=====] - 16s 21ms/step - loss: 0.4805 - accuracy: 0.8302 - val_loss: 0.6175 - val_accuracy: 0.7908
Epoch 28/150
782/782 [=====] - 17s 21ms/step - loss: 0.4712 - accuracy: 0.8357 - val_loss: 0.5918 - val_accuracy: 0.8026
Epoch 29/150
782/782 [=====] - 17s 21ms/step - loss: 0.4614 - accuracy: 0.8378 - val_loss: 0.5339 - val_accuracy: 0.8182
Epoch 30/150
782/782 [=====] - 17s 21ms/step - loss: 0.4546 - accuracy: 0.8409 - val_loss: 0.5538 - val_accuracy: 0.8168
Epoch 31/150

782/782 [=====] - 17s 21ms/step - loss: 0.4472 - accuracy: 0.8439 - val_loss: 0.5470 - val_accuracy: 0.8165
Epoch 32/150
782/782 [=====] - 17s 21ms/step - loss: 0.4373 - accuracy: 0.8464 - val_loss: 0.5443 - val_accuracy: 0.8180
Epoch 33/150
782/782 [=====] - 17s 21ms/step - loss: 0.4281 - accuracy: 0.8502 - val_loss: 0.5620 - val_accuracy: 0.8153
Epoch 34/150
782/782 [=====] - 17s 22ms/step - loss: 0.4207 - accuracy: 0.8528 - val_loss: 0.5528 - val_accuracy: 0.8192
Epoch 35/150
782/782 [=====] - 17s 22ms/step - loss: 0.4178 - accuracy: 0.8519 - val_loss: 0.5850 - val_accuracy: 0.8092
Epoch 36/150
782/782 [=====] - 16s 21ms/step - loss: 0.4095 - accuracy: 0.8559 - val_loss: 0.5426 - val_accuracy: 0.8225
Epoch 37/150
782/782 [=====] - 16s 21ms/step - loss: 0.4022 - accuracy: 0.8591 - val_loss: 0.5453 - val_accuracy: 0.8206
Epoch 38/150
782/782 [=====] - 17s 21ms/step - loss: 0.3957 - accuracy: 0.8618 - val_loss: 0.5312 - val_accuracy: 0.8296
Epoch 39/150
782/782 [=====] - 17s 21ms/step - loss: 0.3881 - accuracy: 0.8653 - val_loss: 0.5315 - val_accuracy: 0.8240
Epoch 40/150
782/782 [=====] - 17s 21ms/step - loss: 0.3801 - accuracy: 0.8654 - val_loss: 0.5552 - val_accuracy: 0.8222
Epoch 41/150
782/782 [=====] - 16s 21ms/step - loss: 0.3758 - accuracy: 0.8671 - val_loss: 0.5308 - val_accuracy: 0.8279
Epoch 42/150
782/782 [=====] - 17s 21ms/step - loss: 0.3691 - accuracy: 0.8699 - val_loss: 0.5285 - val_accuracy: 0.8314
Epoch 43/150
782/782 [=====] - 17s 21ms/step - loss: 0.3617 - accuracy: 0.8732 - val_loss: 0.5153 - val_accuracy: 0.8314
Epoch 44/150
782/782 [=====] - 16s 21ms/step - loss: 0.3554 - accuracy: 0.8744 - val_loss: 0.5437 - val_accuracy: 0.8274
Epoch 45/150
782/782 [=====] - 16s 21ms/step - loss: 0.3583 - accuracy: 0.8732 - val_loss: 0.5300 - val_accuracy: 0.8317
Epoch 46/150
782/782 [=====] - 16s 21ms/step - loss: 0.3497 - accuracy: 0.8768 - val_loss: 0.5227 - val_accuracy: 0.8351
Epoch 47/150

782/782 [=====] - 16s 21ms/step - loss: 0.3421 -
accuracy: 0.8798 - val_loss: 0.5574 - val_accuracy: 0.8200
Epoch 48/150
782/782 [=====] - 16s 21ms/step - loss: 0.3340 -
accuracy: 0.8828 - val_loss: 0.5308 - val_accuracy: 0.8339
Epoch 49/150
782/782 [=====] - 17s 21ms/step - loss: 0.3285 -
accuracy: 0.8842 - val_loss: 0.5264 - val_accuracy: 0.8332
Epoch 50/150
782/782 [=====] - 16s 21ms/step - loss: 0.3266 -
accuracy: 0.8857 - val_loss: 0.5471 - val_accuracy: 0.8260
Epoch 51/150
782/782 [=====] - 16s 21ms/step - loss: 0.3197 -
accuracy: 0.8877 - val_loss: 0.5497 - val_accuracy: 0.8275
Epoch 52/150
782/782 [=====] - 17s 21ms/step - loss: 0.3198 -
accuracy: 0.8889 - val_loss: 0.5288 - val_accuracy: 0.8349
Epoch 53/150
782/782 [=====] - 17s 21ms/step - loss: 0.3128 -
accuracy: 0.8887 - val_loss: 0.5424 - val_accuracy: 0.8313
Epoch 54/150
782/782 [=====] - 16s 21ms/step - loss: 0.3064 -
accuracy: 0.8919 - val_loss: 0.5307 - val_accuracy: 0.8343
Epoch 55/150
782/782 [=====] - 17s 21ms/step - loss: 0.3040 -
accuracy: 0.8913 - val_loss: 0.5266 - val_accuracy: 0.8311
Epoch 56/150
782/782 [=====] - 16s 21ms/step - loss: 0.2965 -
accuracy: 0.8953 - val_loss: 0.5319 - val_accuracy: 0.8358
Epoch 57/150
782/782 [=====] - 17s 21ms/step - loss: 0.2948 -
accuracy: 0.8969 - val_loss: 0.5658 - val_accuracy: 0.8247
Epoch 58/150
782/782 [=====] - 17s 21ms/step - loss: 0.2932 -
accuracy: 0.8972 - val_loss: 0.5480 - val_accuracy: 0.8309
Epoch 59/150
782/782 [=====] - 17s 21ms/step - loss: 0.2870 -
accuracy: 0.8990 - val_loss: 0.5303 - val_accuracy: 0.8370
Epoch 60/150
782/782 [=====] - 16s 21ms/step - loss: 0.2858 -
accuracy: 0.8985 - val_loss: 0.5403 - val_accuracy: 0.8353
Epoch 61/150
782/782 [=====] - 17s 22ms/step - loss: 0.2789 -
accuracy: 0.9021 - val_loss: 0.5430 - val_accuracy: 0.8363
Epoch 62/150
782/782 [=====] - 17s 21ms/step - loss: 0.2748 -
accuracy: 0.9037 - val_loss: 0.5333 - val_accuracy: 0.8357
Epoch 63/150

782/782 [=====] - 17s 21ms/step - loss: 0.2664 -
accuracy: 0.9057 - val_loss: 0.5307 - val_accuracy: 0.8400
Epoch 64/150
782/782 [=====] - 17s 21ms/step - loss: 0.2664 -
accuracy: 0.9041 - val_loss: 0.5251 - val_accuracy: 0.8377
Epoch 65/150
782/782 [=====] - 17s 21ms/step - loss: 0.2638 -
accuracy: 0.9063 - val_loss: 0.5503 - val_accuracy: 0.8326
Epoch 66/150
782/782 [=====] - 17s 22ms/step - loss: 0.2600 -
accuracy: 0.9086 - val_loss: 0.5404 - val_accuracy: 0.8372
Epoch 67/150
782/782 [=====] - 16s 21ms/step - loss: 0.2601 -
accuracy: 0.9077 - val_loss: 0.5377 - val_accuracy: 0.8375
Epoch 68/150
782/782 [=====] - 17s 21ms/step - loss: 0.2549 -
accuracy: 0.9093 - val_loss: 0.5219 - val_accuracy: 0.8400
Epoch 69/150
782/782 [=====] - 17s 22ms/step - loss: 0.2532 -
accuracy: 0.9094 - val_loss: 0.5393 - val_accuracy: 0.8405
Epoch 70/150
782/782 [=====] - 17s 21ms/step - loss: 0.2489 -
accuracy: 0.9107 - val_loss: 0.5549 - val_accuracy: 0.8356
Epoch 71/150
782/782 [=====] - 16s 21ms/step - loss: 0.2432 -
accuracy: 0.9130 - val_loss: 0.5507 - val_accuracy: 0.8395
Epoch 72/150
782/782 [=====] - 16s 21ms/step - loss: 0.2429 -
accuracy: 0.9130 - val_loss: 0.5575 - val_accuracy: 0.8381
Epoch 73/150
782/782 [=====] - 16s 21ms/step - loss: 0.2397 -
accuracy: 0.9141 - val_loss: 0.5400 - val_accuracy: 0.8384
Epoch 74/150
782/782 [=====] - 16s 21ms/step - loss: 0.2316 -
accuracy: 0.9176 - val_loss: 0.5743 - val_accuracy: 0.8340
Epoch 75/150
782/782 [=====] - 16s 21ms/step - loss: 0.2285 -
accuracy: 0.9200 - val_loss: 0.5536 - val_accuracy: 0.8391
Epoch 76/150
782/782 [=====] - 16s 21ms/step - loss: 0.2266 -
accuracy: 0.9202 - val_loss: 0.5438 - val_accuracy: 0.8442
Epoch 77/150
782/782 [=====] - 17s 21ms/step - loss: 0.2255 -
accuracy: 0.9192 - val_loss: 0.5518 - val_accuracy: 0.8391
Epoch 78/150
782/782 [=====] - 17s 21ms/step - loss: 0.2204 -
accuracy: 0.9216 - val_loss: 0.5340 - val_accuracy: 0.8426
Epoch 79/150

782/782 [=====] - 16s 21ms/step - loss: 0.2264 -
 accuracy: 0.9188 - val_loss: 0.5561 - val_accuracy: 0.8404
 Epoch 80/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2174 -
 accuracy: 0.9230 - val_loss: 0.5911 - val_accuracy: 0.8333
 Epoch 81/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2123 -
 accuracy: 0.9248 - val_loss: 0.5444 - val_accuracy: 0.8424
 Epoch 82/150
 782/782 [=====] - 17s 21ms/step - loss: 0.2118 -
 accuracy: 0.9237 - val_loss: 0.5514 - val_accuracy: 0.8394
 Epoch 83/150
 782/782 [=====] - 17s 21ms/step - loss: 0.2126 -
 accuracy: 0.9238 - val_loss: 0.5496 - val_accuracy: 0.8457
 Epoch 84/150
 782/782 [=====] - 17s 21ms/step - loss: 0.2098 -
 accuracy: 0.9248 - val_loss: 0.5594 - val_accuracy: 0.8402
 Epoch 85/150
 782/782 [=====] - 17s 21ms/step - loss: 0.2014 -
 accuracy: 0.9283 - val_loss: 0.5717 - val_accuracy: 0.8324
 Epoch 86/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2060 -
 accuracy: 0.9261 - val_loss: 0.5608 - val_accuracy: 0.8439
 Epoch 87/150
 782/782 [=====] - 16s 21ms/step - loss: 0.2024 -
 accuracy: 0.9285 - val_loss: 0.6205 - val_accuracy: 0.8258
 Epoch 88/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1998 -
 accuracy: 0.9282 - val_loss: 0.5558 - val_accuracy: 0.8422
 Epoch 89/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1955 -
 accuracy: 0.9296 - val_loss: 0.5532 - val_accuracy: 0.8415
 Epoch 90/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1963 -
 accuracy: 0.9314 - val_loss: 0.5883 - val_accuracy: 0.8342
 Epoch 91/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1924 -
 accuracy: 0.9309 - val_loss: 0.5840 - val_accuracy: 0.8394
 Epoch 92/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1913 -
 accuracy: 0.9311 - val_loss: 0.5638 - val_accuracy: 0.8406
 Epoch 93/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1891 -
 accuracy: 0.9328 - val_loss: 0.5686 - val_accuracy: 0.8390
 Epoch 94/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1850 -
 accuracy: 0.9336 - val_loss: 0.5603 - val_accuracy: 0.8458
 Epoch 95/150

782/782 [=====] - 17s 21ms/step - loss: 0.1872 -
accuracy: 0.9334 - val_loss: 0.5665 - val_accuracy: 0.8382
Epoch 96/150
782/782 [=====] - 17s 21ms/step - loss: 0.1838 -
accuracy: 0.9342 - val_loss: 0.5941 - val_accuracy: 0.8374
Epoch 97/150
782/782 [=====] - 17s 21ms/step - loss: 0.1807 -
accuracy: 0.9370 - val_loss: 0.5703 - val_accuracy: 0.8369
Epoch 98/150
782/782 [=====] - 17s 21ms/step - loss: 0.1803 -
accuracy: 0.9357 - val_loss: 0.5803 - val_accuracy: 0.8391
Epoch 99/150
782/782 [=====] - 17s 21ms/step - loss: 0.1763 -
accuracy: 0.9384 - val_loss: 0.5730 - val_accuracy: 0.8400
Epoch 100/150
782/782 [=====] - 17s 21ms/step - loss: 0.1764 -
accuracy: 0.9362 - val_loss: 0.5759 - val_accuracy: 0.8445
Epoch 101/150
782/782 [=====] - 17s 21ms/step - loss: 0.1696 -
accuracy: 0.9393 - val_loss: 0.5872 - val_accuracy: 0.8429
Epoch 102/150
782/782 [=====] - 17s 21ms/step - loss: 0.1710 -
accuracy: 0.9397 - val_loss: 0.5714 - val_accuracy: 0.8453
Epoch 103/150
782/782 [=====] - 17s 22ms/step - loss: 0.1696 -
accuracy: 0.9391 - val_loss: 0.5769 - val_accuracy: 0.8413
Epoch 104/150
782/782 [=====] - 17s 21ms/step - loss: 0.1681 -
accuracy: 0.9395 - val_loss: 0.5676 - val_accuracy: 0.8460
Epoch 105/150
782/782 [=====] - 17s 21ms/step - loss: 0.1701 -
accuracy: 0.9393 - val_loss: 0.5661 - val_accuracy: 0.8456
Epoch 106/150
782/782 [=====] - 17s 21ms/step - loss: 0.1672 -
accuracy: 0.9398 - val_loss: 0.5761 - val_accuracy: 0.8429
Epoch 107/150
782/782 [=====] - 17s 21ms/step - loss: 0.1641 -
accuracy: 0.9416 - val_loss: 0.5753 - val_accuracy: 0.8419
Epoch 108/150
782/782 [=====] - 17s 21ms/step - loss: 0.1606 -
accuracy: 0.9429 - val_loss: 0.5686 - val_accuracy: 0.8438
Epoch 109/150
782/782 [=====] - 17s 21ms/step - loss: 0.1638 -
accuracy: 0.9431 - val_loss: 0.5738 - val_accuracy: 0.8439
Epoch 110/150
782/782 [=====] - 17s 21ms/step - loss: 0.1624 -
accuracy: 0.9425 - val_loss: 0.5911 - val_accuracy: 0.8422
Epoch 111/150

782/782 [=====] - 17s 21ms/step - loss: 0.1562 - accuracy: 0.9442 - val_loss: 0.5879 - val_accuracy: 0.8417
Epoch 112/150
782/782 [=====] - 17s 21ms/step - loss: 0.1549 - accuracy: 0.9453 - val_loss: 0.5779 - val_accuracy: 0.8461
Epoch 113/150
782/782 [=====] - 16s 21ms/step - loss: 0.1563 - accuracy: 0.9439 - val_loss: 0.5853 - val_accuracy: 0.8421
Epoch 114/150
782/782 [=====] - 17s 21ms/step - loss: 0.1532 - accuracy: 0.9439 - val_loss: 0.5874 - val_accuracy: 0.8445
Epoch 115/150
782/782 [=====] - 17s 21ms/step - loss: 0.1515 - accuracy: 0.9465 - val_loss: 0.5855 - val_accuracy: 0.8420
Epoch 116/150
782/782 [=====] - 17s 21ms/step - loss: 0.1474 - accuracy: 0.9468 - val_loss: 0.6026 - val_accuracy: 0.8430
Epoch 117/150
782/782 [=====] - 17s 21ms/step - loss: 0.1532 - accuracy: 0.9459 - val_loss: 0.5878 - val_accuracy: 0.8459
Epoch 118/150
782/782 [=====] - 17s 21ms/step - loss: 0.1444 - accuracy: 0.9486 - val_loss: 0.5816 - val_accuracy: 0.8453
Epoch 119/150
782/782 [=====] - 17s 21ms/step - loss: 0.1478 - accuracy: 0.9466 - val_loss: 0.6020 - val_accuracy: 0.8421
Epoch 120/150
782/782 [=====] - 17s 22ms/step - loss: 0.1460 - accuracy: 0.9486 - val_loss: 0.6025 - val_accuracy: 0.8424
Epoch 121/150
782/782 [=====] - 17s 21ms/step - loss: 0.1412 - accuracy: 0.9500 - val_loss: 0.5915 - val_accuracy: 0.8422
Epoch 122/150
782/782 [=====] - 17s 21ms/step - loss: 0.1428 - accuracy: 0.9491 - val_loss: 0.5933 - val_accuracy: 0.8468
Epoch 123/150
782/782 [=====] - 17s 21ms/step - loss: 0.1471 - accuracy: 0.9467 - val_loss: 0.5936 - val_accuracy: 0.8434
Epoch 124/150
782/782 [=====] - 17s 21ms/step - loss: 0.1404 - accuracy: 0.9503 - val_loss: 0.6090 - val_accuracy: 0.8441
Epoch 125/150
782/782 [=====] - 17s 21ms/step - loss: 0.1414 - accuracy: 0.9508 - val_loss: 0.6195 - val_accuracy: 0.8402
Epoch 126/150
782/782 [=====] - 17s 21ms/step - loss: 0.1371 - accuracy: 0.9514 - val_loss: 0.6461 - val_accuracy: 0.8368
Epoch 127/150

782/782 [=====] - 17s 21ms/step - loss: 0.1393 -
 accuracy: 0.9508 - val_loss: 0.6042 - val_accuracy: 0.8459
 Epoch 128/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1385 -
 accuracy: 0.9509 - val_loss: 0.6088 - val_accuracy: 0.8444
 Epoch 129/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1331 -
 accuracy: 0.9525 - val_loss: 0.5983 - val_accuracy: 0.8432
 Epoch 130/150
 782/782 [=====] - 17s 21ms/step - loss: 0.1356 -
 accuracy: 0.9518 - val_loss: 0.6221 - val_accuracy: 0.8461
 Epoch 131/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1357 -
 accuracy: 0.9519 - val_loss: 0.6176 - val_accuracy: 0.8447
 Epoch 132/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1314 -
 accuracy: 0.9533 - val_loss: 0.6216 - val_accuracy: 0.8419
 Epoch 133/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1325 -
 accuracy: 0.9518 - val_loss: 0.6193 - val_accuracy: 0.8426
 Epoch 134/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1302 -
 accuracy: 0.9537 - val_loss: 0.6057 - val_accuracy: 0.8487
 Epoch 135/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1296 -
 accuracy: 0.9544 - val_loss: 0.6098 - val_accuracy: 0.8462
 Epoch 136/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1322 -
 accuracy: 0.9534 - val_loss: 0.6199 - val_accuracy: 0.8434
 Epoch 137/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1244 -
 accuracy: 0.9551 - val_loss: 0.6169 - val_accuracy: 0.8427
 Epoch 138/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1239 -
 accuracy: 0.9567 - val_loss: 0.6213 - val_accuracy: 0.8458
 Epoch 139/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1244 -
 accuracy: 0.9559 - val_loss: 0.6264 - val_accuracy: 0.8443
 Epoch 140/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1220 -
 accuracy: 0.9574 - val_loss: 0.6185 - val_accuracy: 0.8420
 Epoch 141/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1211 -
 accuracy: 0.9564 - val_loss: 0.6250 - val_accuracy: 0.8435
 Epoch 142/150
 782/782 [=====] - 16s 21ms/step - loss: 0.1231 -
 accuracy: 0.9573 - val_loss: 0.6297 - val_accuracy: 0.8416
 Epoch 143/150

```

782/782 [=====] - 16s 21ms/step - loss: 0.1208 -
accuracy: 0.9573 - val_loss: 0.6173 - val_accuracy: 0.8467
Epoch 144/150
782/782 [=====] - 16s 21ms/step - loss: 0.1209 -
accuracy: 0.9569 - val_loss: 0.6059 - val_accuracy: 0.8449
Epoch 145/150
782/782 [=====] - 16s 21ms/step - loss: 0.1176 -
accuracy: 0.9580 - val_loss: 0.6260 - val_accuracy: 0.8448
Epoch 146/150
782/782 [=====] - 17s 22ms/step - loss: 0.1208 -
accuracy: 0.9567 - val_loss: 0.6284 - val_accuracy: 0.8439
Epoch 147/150
782/782 [=====] - 17s 21ms/step - loss: 0.1157 -
accuracy: 0.9604 - val_loss: 0.6290 - val_accuracy: 0.8443
Epoch 148/150
782/782 [=====] - 17s 21ms/step - loss: 0.1105 -
accuracy: 0.9611 - val_loss: 0.6216 - val_accuracy: 0.8484
Epoch 149/150
782/782 [=====] - 17s 21ms/step - loss: 0.1126 -
accuracy: 0.9605 - val_loss: 0.6159 - val_accuracy: 0.8485
Epoch 150/150
782/782 [=====] - 17s 21ms/step - loss: 0.1161 -
accuracy: 0.9584 - val_loss: 0.6253 - val_accuracy: 0.8470
time: 2493.4596331119537

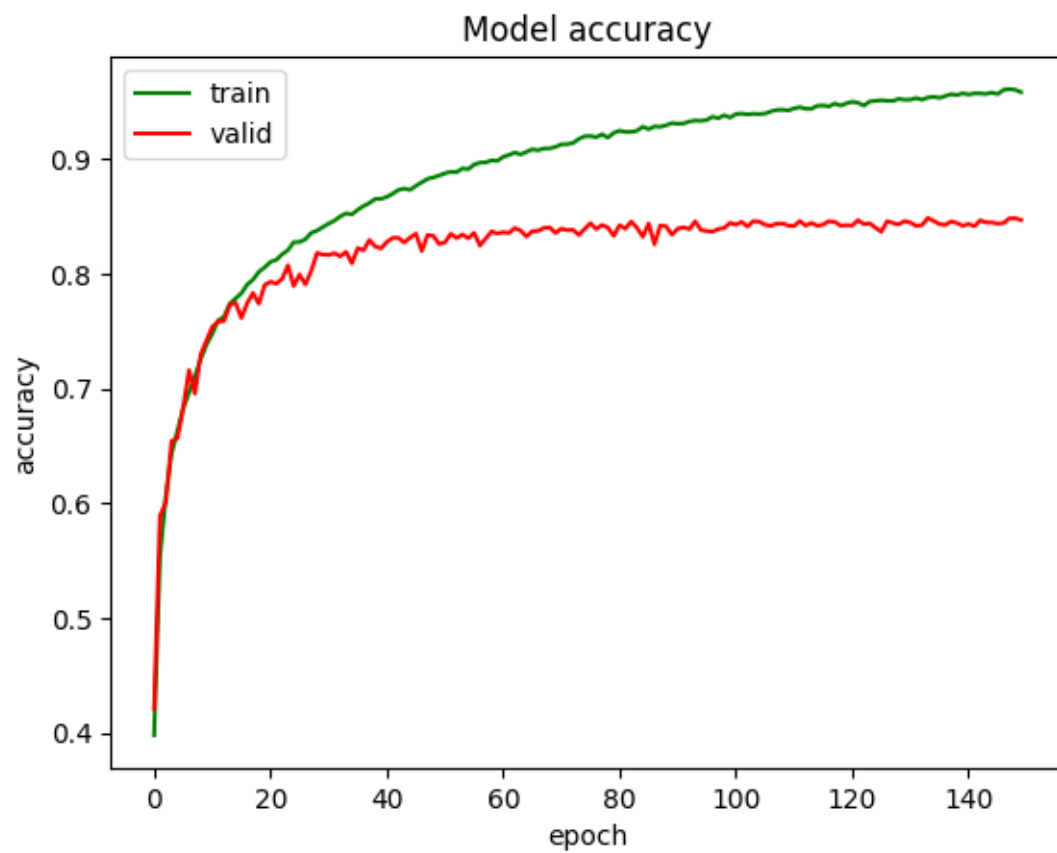
```

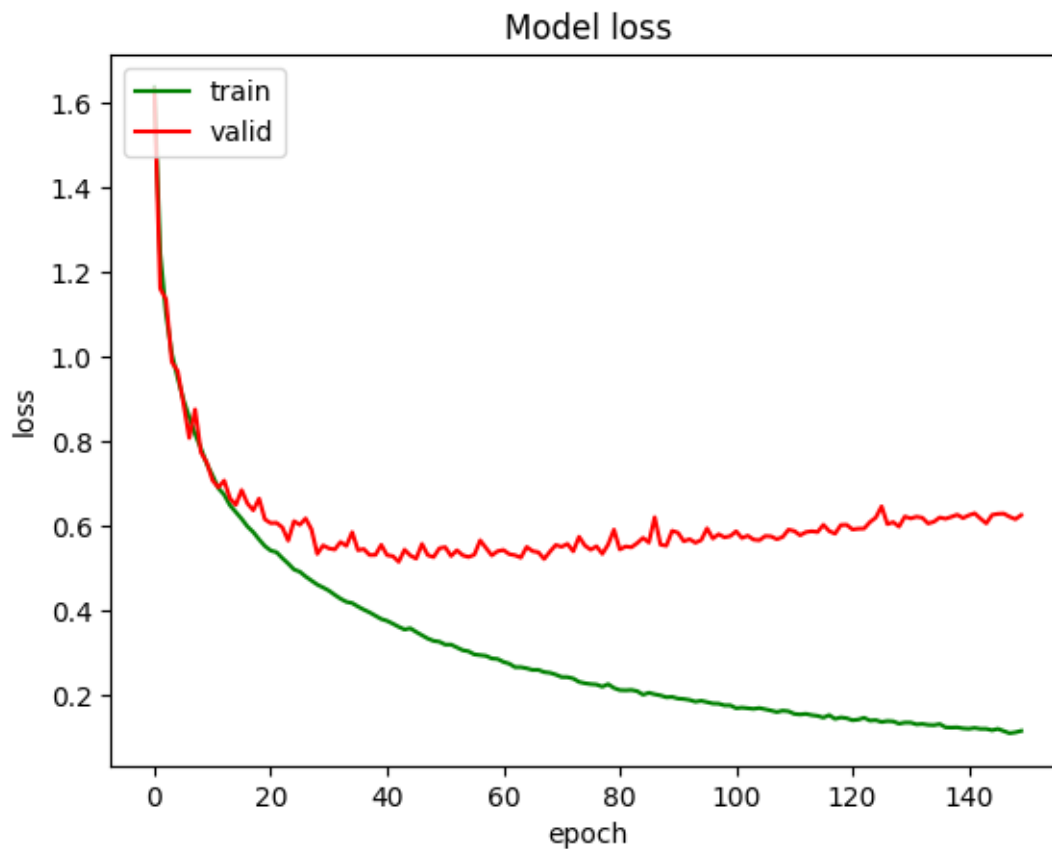
Model znowu uczy się trochę dłużej, około 40 minut, ale nie aż tak dużo, biorąc pod uwagę zaawansowanie modelu.

Nie zachodzi zjawisko overfittingu, a wyniki są zadawalające.

```
[ ]: model.save_weights('./checkpoints/CIFAR')
```

```
[ ]: print_history(history.history)
```





4 New Photo

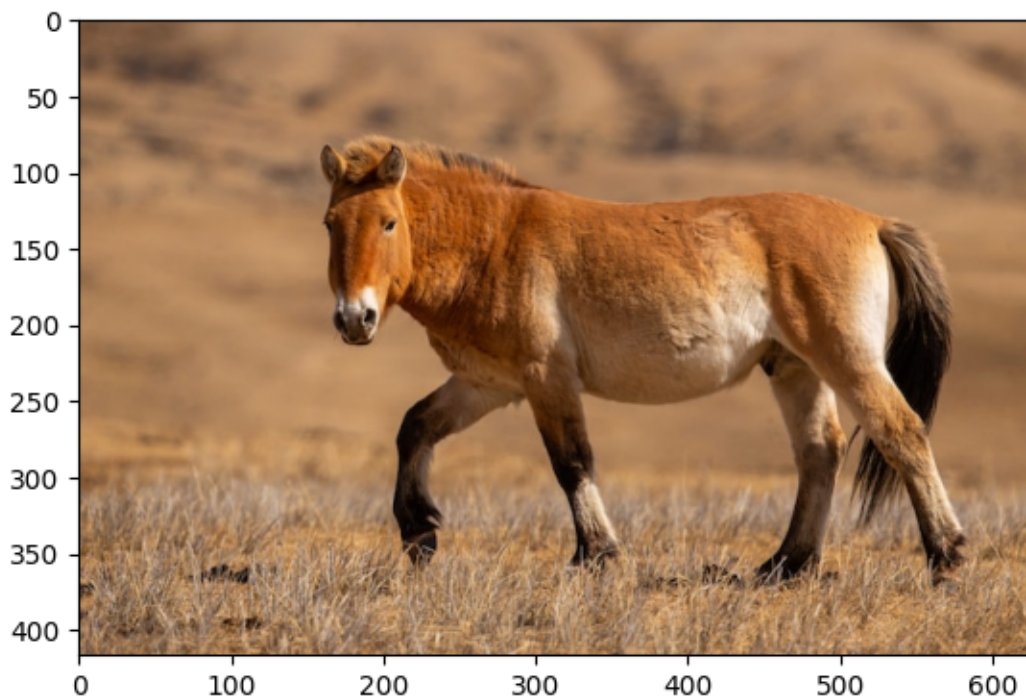
```
[ ]: classes
```

```
[ ]: ['airplane',  
      'automobile',  
      'bird',  
      'cat',  
      'deer',  
      'dog',  
      'frog',  
      'horse',  
      'ship',  
      'truck']
```

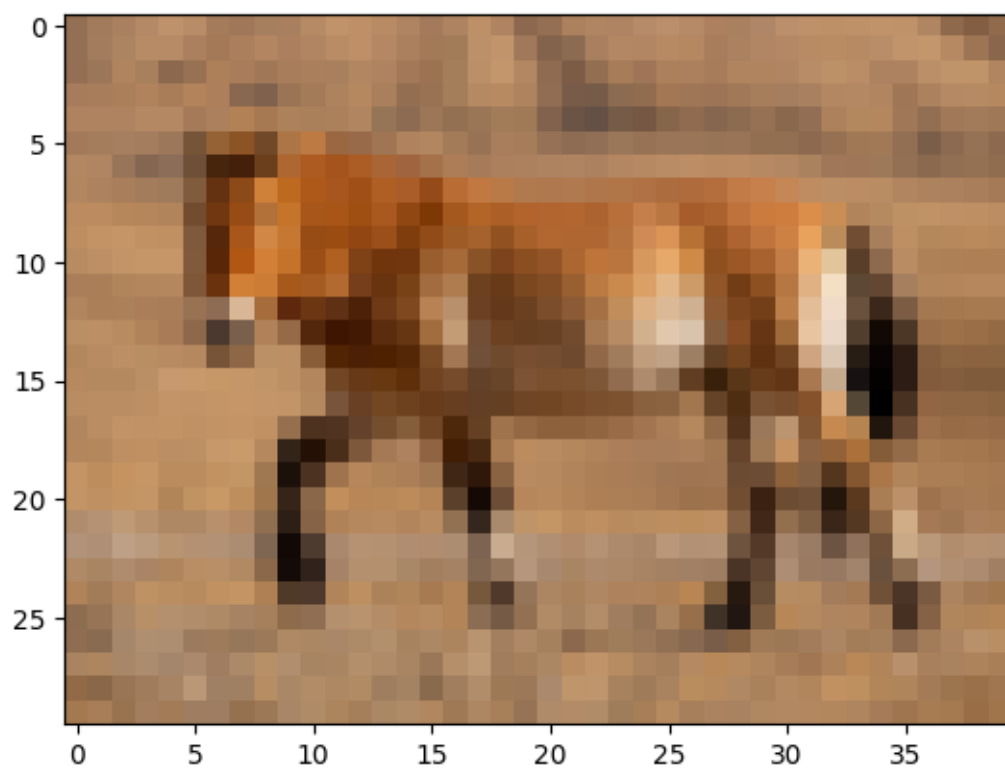
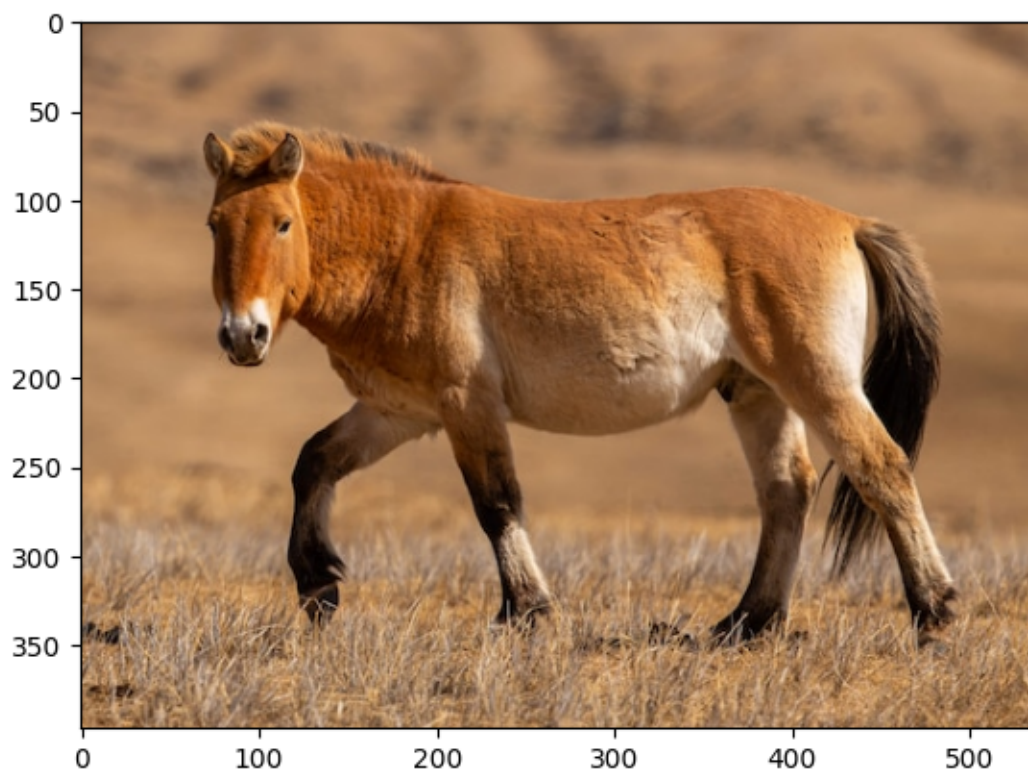
```
[ ]: url = 'https://img.freepik.com/free-photo/  
        ↪przewalskis-horse-portrait-magical-soft-light-during-winter-time-mongolia_475641-365.  
        ↪jpg'  
     im = Image.open(requests.get(url, stream=True).raw)
```

```
print(im.size)
plt.imshow(im)
plt.show()
img = im.crop((90,20,626,417))
print(img.size)
plt.imshow(img)
plt.show()
img.thumbnail((40,36))
plt.imshow(img)
plt.show()
print(img.size)
```

(626, 417)



(536, 397)



(40, 30)

Proporcja jest bardzo niesymetryczna, więc trochę ucinam zbędnych boków, ale dalej zachowując ciekawszy kształt niż kwadrat.

```
[ ]: img = tf.expand_dims(tf.keras.utils.img_to_array(img),axis=0)
img.shape
```

```
[ ]: TensorShape([1, 30, 40, 3])
```

```
[ ]: pred = np.ndarray.flatten(model.predict(img))
pred
```

```
1/1 [=====] - 0s 16ms/step
```

```
[ ]: array([0., 0., 0., 0., 0., 0., 0., 0., 0., 1.], dtype=float32)
```

```
[ ]: classes = tf.convert_to_tensor(classes)
classes
```

```
[ ]: <tf.Tensor: shape=(10,), dtype=string, numpy=
array([b'airplane', b'automobile', b'bird', b'cat', b'deer', b'dog',
       b'frog', b'horse', b'ship', b'truck'], dtype=object)>
```

```
[ ]: classes[pred==1]
```

```
[ ]: <tf.Tensor: shape=(1,), dtype=string, numpy=array([b'truck'], dtype=object)>
```

Jak widać, niestety sieć nie poradziła sobie z niekwadratowym zdjęciem konia.

Na podstawie tego zadania, zrozumiałem jak łatwo przetrenować sieć i jak można poprawić działanie sieci.

Niestety mimo dosyć skomplikowanej architektury i sporego czasu liczenia, nie udało się uzyskać bardzo dobrego wyniku accuracy dla danych nietreningowych.