Технології машинного навчання, лаб 1

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task1 code:

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

```
import tensorflow as tf
      (x_train,
                       y_train),
                                           (x_test,
                                                            y_test)
tf.keras.datasets.cifar100.load_data()
      print(f'x_train shape: {x_train.shape}')
      print(f'x_test shape: {x_test.shape}')
      print(f'y_train shape: {y_train.shape}')
      print(f'y_test shape: {y_test.shape}')
      plt.figure(figsize=(10, 10))
      for i in range(25):
         ax = plt.subplot(5, 5, i + 1)
         plt.imshow(x_train[i])
         plt.title(f'Label: {y_train[i]}')
         plt.axis('off')
      plt.show()
```

```
x_train shape: (50000, 32, 32, 3)
x_test shape: (10000, 32, 32, 3)
y_train shape: (50000, 1)
y_test shape: (10000, 1)
     Label: [19]
                             Label: [29]
                                                     Label: [0]
                                                                           Label: [11]
                                                                                                    Label: [1]
     Label: [86]
                             Label: [90]
                                                    Label: [28]
                                                                           Label: [23]
                                                                                                   Label: [31]
                             Label: [96]
                                                    Label: [82]
                                                                           Label: [17]
     Label: [39]
                                                                                                   Label: [71]
     Label: [39]
                             Label: [8]
                                                    Label: [97]
                                                                           Label: [80]
                                                                                                   Label: [71]
                                                    Label: [70]
                                                                           Label: [87]
                                                                                                   Label: [59]
      Label: [74]
                             Label: [59]
```

```
import keras
from keras import models, layers

def MyConv2D(x, filters = 32, kernel_size = (3, 3), strides = (1, 1),
padding = "same", activation = "relu"):
    x = layers.Conv2D(
        filters=filters,
        kernel_size=kernel_size,
        strides=strides,
        padding=padding,
    )(x)

    x = layers.BatchNormalization(axis=-1)(x)
    x = layers.Activation("relu")(x)
    return x
```

```
def Stem(input):
         x = MyConv2D(input, filters=32, kernel size=(3,3), strides=(2,2),
padding="valid", activation="relu")
         x = MyConv2D(x, filters=32, kernel size=(3,3), padding="valid",
activation="relu")
          x = MyConv2D(x, filters=64, kernel_size=(3,3), padding="same",
activation="relu")
       x = layers.MaxPooling2D((3, 3), strides=(2, 2), padding="valid")(x)
         x = MyConv2D(x, filters=80, kernel_size=(1,1), padding="valid",
activation="relu")
         x = MyConv2D(x, filters=192, kernel_size=(3,3), padding="valid",
activation="relu")
       x = layers.MaxPooling2D((3, 3), strides=(2, 2), padding="valid")(x)
       return x
     def InceptionA(input, filters=64):
        x1 = MyConv2D(input, filters=64, kernel size=(1,1), padding="same",
activation="relu")
         x1 = MyConv2D(x1, filters=96, kernel size=(3,3), padding="same",
activation="relu")
         x1 = MyConv2D(x1, filters=96, kernel size=(3,3), padding="same",
activation="relu")
        x2 = MyConv2D(input, filters=48, kernel_size=(1,1), padding="same",
activation="relu")
         x2 = MyConv2D(x2, filters=64, kernel_size=(3,3), padding="same",
activation="relu")
                = layers.AveragePooling2D((3, 3), strides=(1,
                                                                         1),
padding="same")(input)
        x3 = MyConv2D(x3, filters=filters, kernel_size=(1,1), padding="same",
activation="relu")
        x4 = MyConv2D(input, filters=64, kernel_size=(1,1), padding="same",
activation="relu")
       x = layers.Concatenate(axis=-1)([x1, x2, x3, x4])
       return x
     def ReductionA(input):
        x1 = MyConv2D(input, filters=64, kernel_size=(1,1), padding="same",
activation="relu")
```

```
x1 = MyConv2D(x1, filters=96, kernel_size=(3,3), padding="same",
activation="relu")
         x1 = MyConv2D(x1, filters=96, kernel_size=(3,3), strides=(2,2),
padding="valid", activation="relu")
        x2 = MyConv2D(input, filters=384, kernel_size=(3,3), strides=(2,2),
padding="valid", activation="relu")
              x3 = layers.MaxPooling2D((3, 3), strides=(2,
                                                                         2),
padding="valid")(input)
       x = layers.Concatenate(axis=-1)([x1, x2, x3])
       return x
     def InceptionB(input, filters=192):
                     MyConv2D(input, filters=filters, kernel_size=(1,1),
padding="same", activation="relu")
        x1 = MyConv2D(x1, filters=filters, kernel_size=(7,1), padding="same",
activation="relu")
        x1 = MyConv2D(x1, filters=filters, kernel_size=(1,7), padding="same",
activation="relu")
        x1 = MyConv2D(x1, filters=filters, kernel_size=(7,1), padding="same",
activation="relu")
         x1 = MyConv2D(x1, filters=192, kernel_size=(1,7), padding="same",
activation="relu")
            x2 = MyConv2D(input, filters=filters, kernel size=(1,1),
padding="same", activation="relu")
        x2 = MyConv2D(x2, filters=filters, kernel_size=(1,7), padding="same",
activation="relu")
         x2 = MyConv2D(x2, filters=192, kernel_size=(7,1), padding="same",
activation="relu")
            x3
                      layers.AveragePooling2D((3, 3), strides=(1,
                                                                         1),
padding="same")(input)
         x3 = MyConv2D(x3, filters=192, kernel_size=(1,1), padding="same",
activation="relu")
        x4 = MyConv2D(input, filters=192, kernel size=(1,1), padding="same",
activation="relu")
       x = layers.Concatenate(axis=-1)([x1, x2, x3, x4])
```

```
return x
      def ReductionB(input):
        x1 = MyConv2D(input, filters=192, kernel size=(1,1), padding="same",
activation="relu")
         x1 = MyConv2D(x1, filters=192, kernel_size=(1,7), padding="same",
activation="relu")
         x1 = MyConv2D(x1, filters=192, kernel_size=(7,1), padding="same",
activation="relu")
         x1 = MyConv2D(x1, filters=192, kernel_size=(3,3), strides=(2,2),
padding='valid', activation="relu")
        x2 = MyConv2D(input, filters=192, kernel_size=(1,1), padding="same",
activation="relu")
        x2 = MyConv2D(input, filters=320, kernel_size=(3,3), strides=(2,2),
padding='valid', activation="relu")
                         layers.MaxPooling2D((3, 3), strides=(2,
                                                                          2),
padding="valid")(input)
       x = layers.Concatenate(axis=-1)([x1, x2, x3])
       return x
      def InceptionC(input):
        x1 = MyConv2D(input, filters=448, kernel_size=(1,1), padding="same",
activation="relu")
         x1 = MyConv2D(x1, filters=384, kernel_size=(3,3), padding="same",
activation="relu")
        x1_1 = MyConv2D(x1, filters=384, kernel_size=(1,3), padding="same",
activation="relu")
        x1_2 = MyConv2D(x1, filters=384, kernel_size=(3,1), padding="same",
activation="relu")
        x2 = MyConv2D(input, filters=384, kernel_size=(1,1), padding="same",
activation="relu")
         x2_1 = MyConv2D(x2, filters=384, kernel_size=(1,3), padding="same",
activation="relu")
         x2_2 = MyConv2D(x2, filters=384, kernel_size=(3,1), padding="same",
activation="relu")
```

```
1),
             x3
                       layers.AveragePooling2D((3, 3), strides=(1,
padding="same")(input)
         x3 = MyConv2D(x3, filters=192, kernel_size=(1,1), padding="same",
activation="relu")
        x4 = MyConv2D(input, filters=320, kernel_size=(1,1), padding="same",
activation="relu")
        x = layers.Concatenate(axis=-1)([x1_1, x1_2, x2_1, x2_2, x3, x4])
       return x
      def Aux(input, n_classes = 1000):
                     layers.AveragePooling2D((5, 5), strides=(3,
                                                                          3),
padding="same")(input)
          x = MyConv2D(x, filters=128, kernel_size=(1,1), padding="same",
activation="relu")
       x = layers.Flatten()(x)
       x = layers.Dense(768, activation="relu")(x)
       x = layers.Dropout(rate = 0.2) (x)
       x = layers.Dense(n_classes, activation="softmax")(x)
       return x
     def InceptionV3(input shape=(299, 299, 3), n classes = 1000):
       input = layers.Input(input_shape)
       x = layers.Rescaling(1./255)(input)
       x = layers.Resizing(299, 299)(x)
       x = layers.RandomColorJitter(
         value_range=(0, 1),
         brightness_factor=0.2,
         contrast_factor=0.2,
         saturation factor=0.2,
         hue_factor=0.2
        )(x)
        x = layers.RandomFlip(
         mode="horizontal and vertical"
        )(x)
       x = Stem(x)
        x = InceptionA(x, 32)
        x = InceptionA(x)
        x = InceptionA(x)
```

```
x = ReductionA(x)

x = InceptionB(x, 128)
x = InceptionB(x, 160)
x = InceptionB(x, 160)
x = InceptionB(x, 192)

aux = Aux(x, n_classes)

x = ReductionB(x)

x = InceptionC(x)
x = InceptionC(x)

x = InceptionC(x)

x = layers.GlobalAveragePooling2D()(x)

x = layers.Dense(2048, activation="relu")(x)
x = layers.Dropout(rate = 0.2) (x)
output = layers.Dense(n_classes, activation="softmax")(x)

model = models.Model(inputs=input, outputs=output)
return model
```

```
from keras import optimizers

epochs = 50

initial_learning_rate = 1e-3
  final_learning_rate = 1e-8
  learning_rate_decay_factor = (final_learning_rate /
initial_learning_rate) ** (1 / epochs)
  batch_size = 32
  steps_per_epoch = len(x_train) * 8 // 10 // batch_size

print(f'Initial learning rate: {initial_learning_rate}')
  print(f'Final learning rate: {final_learning_rate}')
  print(f'Learning rate decay factor: {learning_rate_decay_factor}')
  print(f'Batch size: {batch_size}')
  print(f'Steps per epoch: {steps_per_epoch}')
```

```
learning_rate = optimizers.schedules.ExponentialDecay(
    initial_learning_rate=initial_learning_rate,
    decay_steps=steps_per_epoch,
    decay_rate=learning_rate_decay_factor
)

model = InceptionV3(input_shape=(32, 32, 3), n_classes=100)
    model.compile(optimizer=optimizers.Adam(learning_rate=learning_rate),
loss='sparse_categorical_crossentropy', metrics=['accuracy'])

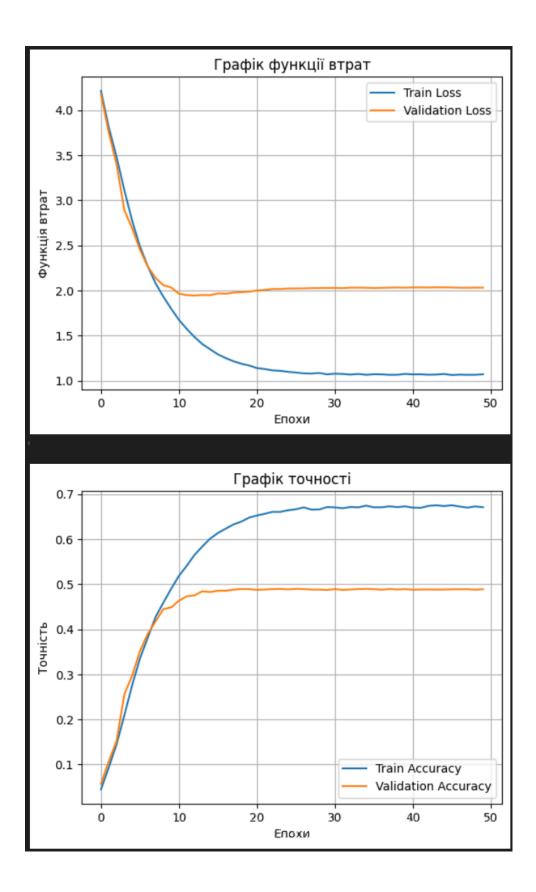
model.summary()
```

```
history = model.fit(x_train, y_train, epochs=epochs,
validation_split=0.2, steps_per_epoch=steps_per_epoch, batch_size=batch_size)
```

```
model.evaluate(x_test, y_test)
```

```
313/313 — 11s 34ms/step - accuracy: 0.5111 - loss: 1.9735 [2.0017733573913574, 0.5048999786376953]
```

```
plt.plot(history.history['loss'], label='Train Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.xlabel('Enoxu')
plt.ylabel('Функція втрат')
plt.title('Графік функції втрат')
plt.legend()
plt.grid()
plt.show()
plt.plot(history.history['accuracy'], label='Train Accuracy')
plt.plot(history.history['val accuracy'], label='Validation Accuracy')
plt.xlabel('Епохи')
plt.ylabel('Точність')
plt.title('Графік точності')
plt.legend()
plt.grid()
plt.show()
```



```
models.save_model(model, 'model.keras')
```

```
model = models.load_model('model.keras')
model.evaluate(x_test, y_test)
```

task2 code for .tflite file generation:

```
from keras import models
import tensorflow as tf

model = models.load_model('../task1/model.keras')
```

```
converter = tf.lite.TFLiteConverter.from_keras_model(model)

tflite_model = converter.convert()

with tf.io.gfile.GFile('./model/model.tflite', 'wb') as f:
    f.write(tflite_model)
```

task2 code for server with /predict endpoint:

```
from fastapi import FastAPI, File, UploadFile
     from fastapi.responses import JSONResponse
     from PIL import Image
     import io
     import numpy as np
     import tensorflow as tf
     app = FastAPI()
     cifar100 fine labels = [
          'apple', 'aquarium_fish', 'baby', 'bear', 'beaver', 'bed', 'bee',
beetle',
         'bicycle', 'bottle', 'bowl', 'boy', 'bridge', 'bus', 'butterfly',
             'camel', 'can', 'castle', 'caterpillar', 'cattle', 'chair',
chimpanzee',
         'clock', 'cloud', 'cockroach', 'couch', 'crab', 'crocodile', 'cup',
dinosaur',
             'dolphin', 'elephant', 'flatfish', 'forest', 'fox', 'girl',
hamster', 'house',
```

```
'kangaroo', 'keyboard', 'lamp', 'lawn_mower', 'leopard', 'lion',
lizard',
          'lobster', 'man', 'maple_tree', 'motorcycle', 'mountain', 'mouse',
mushroom',
            'oak_tree', 'orange', 'orchid', 'otter', 'palm_tree', 'pear',
pickup_truck',
            'pine_tree', 'plain', 'plate', 'poppy', 'porcupine', 'possum',
rabbit',
         'raccoon', 'ray', 'road', 'rocket', 'rose', 'sea', 'seal', 'shark',
shrew',
           'skunk', 'skyscraper', 'snail', 'snake', 'spider', 'squirrel',
streetcar',
              'sunflower', 'sweet_pepper', 'table', 'tank', 'telephone',
television', 'tiger',
         'tractor', 'train', 'trout', 'tulip', 'turtle', 'wardrobe', 'whale',
willow_tree',
         'wolf', 'woman', 'worm'
     ]
     print(len(cifar100_fine_labels))
    model_content = open("./model/model.tflite", "rb").read()
     interpreter = tf.lite.Interpreter(model content=model content)
    def infer(img):
       interpreter.allocate_tensors()
       input_details = interpreter.get_input_details()
       output_details = interpreter.get_output_details()
       input_data = np.array([np.array(img)], dtype=np.float32)
       interpreter.set_tensor(input_details[0]['index'], input_data)
       interpreter.invoke()
       output data = interpreter.get tensor(output details[0]['index'])
       return output_data
     @app.get("/ping")
     async def ping():
       return "pong"
```

task2 Dockerfile:

```
FROM tensorflow/tensorflow

WORKDIR /app

COPY . .

RUN pip install --upgrade pip

RUN pip install fastapi pillow uvicorn numpy python-multipart

tensorflow-cpu

EXPOSE 3000

CMD ["uvicorn", "--host", "0.0.0.0", "--port", "3000", "server:app"]
```

task2 process of running and querying the api:

1. build docker image

```
PYS E:\Maricrp\semestr2\machine-learning-technologies\Cd .\Ll\task2\
PYS E:\Maricrp\semestr2\machine-learning-technologies\L1\task2> docker build -t l1task2:1.0 .

[+] Building 56.5s (10/10) FINISHED

> [internal] load build definition from Dockerfile

> > transferring dockerfile: 2838

> [internal] load metadata for docker.io/tensorflow/tensorflow:latest

> [internal] load .dockerignore

> > transferring context: 28

> [1/5] FROM docker.io/tensorflow/tensorflow:latest@sha256:f24e8494d43a4e613edd7fbd3782594368d52d79afle216051139ed2d7830682

> [internal] load build context

> > transferring context: 626.51kB

> CACHED [2/5] WORKDIR /app

> [3/5] COPY .

> [4/5] RUN pip install --upgrade pip

> [5/5] RUN pip install fastapi pillow uvicorn numpy python-multipart tensorflow-cpu

> exporting to image

> > exporting to image

> > writing image sha256:4db0774acf8a5768be68eb57fba9e938c07c6e2562e20297199f5cd5c13b0f81

> > naming to docker.io/library/lltask2:1.0

View build details: docker-desktop://dashboard/build/desktop-linux/desktop-linux/7ge7sia8un2obuusijqamhxy0

What's next:

View a summary of image vulnerabilities and recommendations → docker scout quickview

PS E:\Maricrp\semestr2\machine-learning-technologies\L1\task2>
```

2. run docker container & expose port

```
PS E:\Marictp\semestr2\machine-learning-technologies\L1\task2> docker run -p 3000:3000 l1task2:1.0
2025-05-16 03:17:13.971795: I tensorflow/core/util/port.cc:153] oneDNN custom operations are on. You
. To turn them off, set the environment variable `TF_ENABLE_ONEDNN_OPTS=0`.
2025-05-16 03:17:16.002878: I tensorflow/core/platform/cpu_feature_guard.cc:210] This TensorFlow bi
To enable the following instructions: AVX2 AVX_VNNI FMA, in other operations, rebuild TensorFlow wi
2025-05-16 03:17:34.548278: I tensorflow/core/util/port.cc:153] oneDNN custom operations are on. You
. To turn them off, set the environment variable `TF_ENABLE_ONEDNN_OPTS=0`.
/usr/local/lib/python3.11/dist-packages/tensorflow/lite/python/interpreter.py:457: UserWarning:
    TF 2.20. Please use the LiteRT interpreter from the ai_edge_litert package.
    See the [migration guide](https://ai.google.dev/edge/litert/migration)
    for details.
 warnings.warn(_INTERPRETER_DELETION_WARNING)
          Started server process [1]
INFO:
         Waiting for application startup.
INFO:
          Application startup complete.
         Uvicorn running on http://0.0.0.0:3000 (Press CTRL+C to quit)
INFO:
```

3. Enter url & attach image file in postman





4. Make request POST /predict with image attached

Result is correct!