

```
from google.colab import drive
drive.mount('/content/drive', force_remount=True)
```

Mounted at /content/drive

```
!pip install -q libtiff
!pip install -q tqdm
```

```
|████████████████████████████████████████| 133kB 17.6MB/s
Building wheel for libtiff (setup.py) ... done
```

```
# import the necessary packages
import matplotlib.pyplot as plt
import sys
import os
from pathlib import Path
from libtiff import TIFF
import numpy as np
from typing import List
from tqdm.notebook import tqdm
from time import sleep
from PIL import Image
import IPython.display
import tensorflow as tf
from sklearn.metrics import balanced_accuracy_score
from keras.optimizers import SGD, RMSprop, Adam
from keras import optimizers, Model, callbacks
from sklearn.metrics import mean_squared_error
from keras.utils import plot_model
from sklearn.preprocessing import StandardScaler, MinMaxScaler
```

Класс Metrics

Реализует метрики точности, используемые для оценивания модели:

1. точность,
2. сбалансированную точность.

```
class Metrics:

    @staticmethod
    def accuracy(gt: List[int], pred: List[int]):
        assert len(gt) == len(pred), 'gt and prediction should be of equal
        return sum(int(i[0] == i[1]) for i in zip(gt, pred)) / len(gt)
```

```

@staticmethod
def accuracy_balanced(gt: List[int], pred: List[int]):
    return balanced_accuracy_score(gt, pred)

@staticmethod
def print_all(gt: List[int], pred: List[int], info: str):
    print(f'metrics for {info}:')
    print('\t accuracy {:.4f}:'.format(Metrics.accuracy(gt, pred)))
    print('\t balanced accuracy {:.4f}:'.format(Metrics.accuracy_balanc

```

Класс Dataset

Предназначен для работы с наборами данных, хранящихся на Google Drive, обеспечивает чтение изображений и соответствующих меток, а также формирование пакетов (батчей).

```

class Dataset:
    def __init__(self, name, gdrive_dir):
        self.name = name
        self.is_loaded = False
        p = Path("/content/drive/MyDrive/" + gdrive_dir + name + '.npz')
        if p.exists():
            print(f'[INFO] Loading dataset {self.name} from npz...')
            np_obj = np.load(str(p))
            self.images = np_obj['data']
            self.labels = np_obj['labels']
            self.n_files = self.images.shape[0]
            self.is_loaded = True
            print(f'[INFO] Done. Dataset {name} consists of {self.n_files}')

    def image(self, i):
        # read i-th image in dataset and return it as numpy array
        if self.is_loaded:
            return self.images[i, :, :, :]

    def images_seq(self, n=None):
        # sequential access to images inside dataset (is needed for testing)
        for i in range(self.n_files if not n else n):
            yield self.image(i)

    def random_image_with_label(self):
        # get random image with label from dataset
        i = np.random.randint(self.n_files)
        return self.image(i), self.labels[i]

    def random_batch_with_labels(self, n):
        # create random batch of images with labels (is needed for training)

```

```

# Create random batch of images with labels (is needed for training
indices = np.random.choice(self.n_files, n)
imgs = []
for i in indices:
    imgs.append(self.image(i))
logits = np.array([self.labels[i] for i in indices])
return np.array(imgs), logits

def image_with_label(self, i: int):
    # return i-th image with label from dataset
    return self.image(i), self.labels[i]

```

Класс Model

Предназначен для работы с нейросетью. Обеспечивает загрузку, обучение, тестирование и сохранение нейронной сети, отображение дополнительной информации об алгоритме обучения.

```
class Model:
```

```

def __init__(self):
    chanDim = -1
    self.model = tf.keras.models.Sequential(
        [tf.keras.layers.Conv2D(8, (3, 3), padding="same", activation='r
        input_shape=(224, 224, 3)),
        tf.keras.layers.MaxPooling2D(pool_size=(2, 2)),
        tf.keras.layers.Conv2D(64, (3, 3), padding="same" , activation='r
        tf.keras.layers.MaxPooling2D(data_format='channels_last'),

        tf.keras.layers.Conv2D(128, (3, 3), padding="same" , activation='
        tf.keras.layers.MaxPooling2D(data_format='channels_last'),

        tf.keras.layers.Conv2D(256, (3, 3), padding="same", activation='r
        tf.keras.layers.MaxPooling2D(data_format='channels_last'),

        tf.keras.layers.Conv2D(512, (3, 3), padding="same", activation='r
        tf.keras.layers.MaxPooling2D(data_format='channels_last'),
        tf.keras.layers.Dropout(0.5),

        tf.keras.layers.Flatten(),

        tf.keras.layers.Dense(1000, kernel_regularizer=tf.keras.regulariz
        tf.keras.layers.Activation("relu"),
        tf.keras.layers.Dropout(0.3),

```

```

tf.keras.layers.Dense(512, kernel_regularizer=tf.keras.regularize
tf.keras.layers.Activation("relu"),
tf.keras.layers.Dropout(0.3),

tf.keras.layers.Dense(256, kernel_regularizer=tf.keras.regularize
tf.keras.layers.Activation("relu"),
tf.keras.layers.Dropout(0.3),

tf.keras.layers.Dense(9),
tf.keras.layers.Activation("softmax")
])

self.save_dir = "/content/drive/MyDrive/nn_tasks/"
self.INIT_LR = 6e-4
self.EPOCHS = 100
self.BS = 64

def save_model(self, name: str):
    # save model to SAVE_DIR folder on gdrive with name 'name'
    #self.save_dir += name
    self.model.save(self.save_dir + "model-res/" + name + "_model.h5")
    self.model.save_weights(self.save_dir + "model-res/" + name + "_mod
    plot_model(self.model, to_file=self.save_dir + "model-res/" + name

def load(self, name: str, is_checkpoint=False):
    p1 = Path(self.save_dir + "model-res/" + name + ".h5")
    p2 = Path(self.save_dir + "model-res/best_checkpoint/")

    self.model = None
    if is_checkpoint and p2.exists():
        self.model = tf.keras.models.load_model(p2)
    elif p1.exists():
        self.model = tf.keras.models.load_model(p1)

def train(self, dataset: Dataset, is_showing_history=True):

    print(f"[INFO] split and shuffle the data...")
    d_train_images_1, d_train_labels_1 = dataset.random_batch_with_labe
    n_train = int(dataset.n_files * 0.92)

    d_train_images = d_train_images_1[0:n_train]
    d_train_labels = d_train_labels_1[0:n_train]
    #LBL1
    d_val_images = d_train_images_1[n_train:dataset.n_files]

```

```

d_val_labels = d_train_labels_1[n_train:dataset.n_files]

#scaler = MinMaxScaler()
#X = scaler.fit_transform(d_train_images)
#X_val = scaler.fit_transform(d_val_images)

optA = Adam(learning_rate=self.INIT_LR)
self.model.compile(loss=tf.keras.losses.SparseCategoricalCrossentropy,
                    metrics=["accuracy"])

#LBL2
checkpoint = callbacks.ModelCheckpoint(self.save_dir + "model-res/best_val_acc.h5",
save_weights_only=False, save_freq='epoch')
callbacks_list = [checkpoint]
#LBL3
print(f"[INFO] training network...")
self.History = self.model.fit(d_train_images, d_train_labels, validation_data=(d_val_images, d_val_labels),
                              callbacks=callbacks_list, verbose=1)
print(f'[INFO] training done')
#LBL4
self.show_history(str(n_train))

def test_on_dataset(self, dataset: Dataset, limit=None):
    # you can upgrade this code if you want to speed up testing using batch processing
    predictions = []
    n = dataset.n_files if not limit else int(dataset.n_files * limit)
    for img in tqdm(dataset.images_seq(n), total=n):
        predictions.append(self.test_on_image(img))
    return predictions

def test_on_image(self, img):
    # todo: replace this code with a more efficient one
    prediction = self.model.predict(np.array([img]), batch_size=1)
    return np.argmax(prediction)

def evaluate_model(self, dataset: Dataset):
    print("[INFO] evaluating network...")
    result = self.model.evaluate(dataset.images, dataset.labels, batch_size=100)
    print('test loss, test acc:', result)

    predictions = self.model.predict(dataset.images)
    pred = []
    print("MSE (test_y, predictions)")
    for i in range(0, len(dataset.labels)):
        pred.append(predictions[i].max())

```

```

        pred.append(predictions[i].max())
    print(mean_squared_error(dataset.labels, pred, squared=False))

```

```

def show_history(self, size):
    N = np.arange(0, self.EPOCHS)
    plt.style.use("ggplot")
    plt.figure()
    plt.plot(N, self.History.history["loss"], label="train_loss")
    plt.plot(N, self.History.history["accuracy"], label="train_acc")
    plt.plot(N, self.History.history["val_loss"], label="test_loss")
    plt.plot(N, self.History.history["val_accuracy"], label="test_acc")
    plt.title("Training Los")
    plt.xlabel("Epoch #")
    plt.ylabel("Loss, accuracy")
    plt.legend()
    plt.savefig(self.save_dir + "model-res/history_training_on_" + size

```

```
PROJECT_DIR = "nn_tasks/"
```

```
EVALUATE_ONLY = False
```

```
TEST_ON_LARGE_DATASET = True
```

```
TISSUE_CLASSES = ('ADI', 'BACK', 'DEB', 'LYM', 'MUC', 'MUS', 'NORM', 'STR',
```

```
dataset_test = Dataset('test', PROJECT_DIR)
```

```
dataset_train = Dataset('train', PROJECT_DIR)
```

```

[INFO] Loading dataset test from npz...
[INFO] Done. Dataset test consists of 4500 images.
[INFO] Loading dataset train from npz...
[INFO] Done. Dataset train consists of 18000 images.

```

```
model = Model()
```

```
if not EVALUATE_ONLY:
```

```
    model.train(dataset_train, is_showing_history=True)
```

```
    model.save_model('best')
```

```
else:
```

```
    model.load('best')
```

```
[INFO] split and shuffle the data...
[INFO] training network...
Epoch 1/100
259/259 [=====] - 25s 94ms/step - loss: 25.2997 - accur

Epoch 00001: val_accuracy improved from -inf to 0.42361, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 2/100
259/259 [=====] - 25s 96ms/step - loss: 9.3074 - accur

Epoch 00002: val_accuracy improved from 0.42361 to 0.62153, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 3/100
259/259 [=====] - 24s 94ms/step - loss: 5.1818 - accur

Epoch 00003: val_accuracy did not improve from 0.62153
Epoch 4/100
259/259 [=====] - 24s 94ms/step - loss: 3.4037 - accur

Epoch 00004: val_accuracy improved from 0.62153 to 0.70139, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 5/100
259/259 [=====] - 24s 94ms/step - loss: 2.4769 - accur

Epoch 00005: val_accuracy improved from 0.70139 to 0.77431, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 6/100
259/259 [=====] - 24s 94ms/step - loss: 1.9708 - accur

Epoch 00006: val_accuracy improved from 0.77431 to 0.80625, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 7/100
259/259 [=====] - 24s 94ms/step - loss: 1.6131 - accur

Epoch 00007: val_accuracy improved from 0.80625 to 0.81250, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 8/100
259/259 [=====] - 24s 94ms/step - loss: 1.3739 - accur

Epoch 00008: val_accuracy improved from 0.81250 to 0.85556, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 9/100
259/259 [=====] - 25s 95ms/step - loss: 1.1241 - accur

Epoch 00009: val_accuracy did not improve from 0.85556
Epoch 10/100
259/259 [=====] - 24s 94ms/step - loss: 1.0445 - accur

Epoch 00010: val_accuracy improved from 0.85556 to 0.89444, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 11/100
259/259 [=====] - 24s 94ms/step - loss: 0.9527 - accur

Epoch 00011: val_accuracy improved from 0.89444 to 0.91736, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 12/100
259/259 [=====] - 25s 95ms/step - loss: 0.8349 - accur
```

```
Epoch 00012: val_accuracy did not improve from 0.91736
Epoch 13/100
259/259 [=====] - 24s 94ms/step - loss: 0.7818 - accuracy: 0.91736

Epoch 00013: val_accuracy did not improve from 0.91736
Epoch 14/100
259/259 [=====] - 24s 94ms/step - loss: 0.7302 - accuracy: 0.91736

Epoch 00014: val_accuracy improved from 0.91736 to 0.93681, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 15/100
259/259 [=====] - 24s 94ms/step - loss: 0.6678 - accuracy: 0.93681

Epoch 00015: val_accuracy did not improve from 0.93681
Epoch 16/100
259/259 [=====] - 24s 94ms/step - loss: 0.6698 - accuracy: 0.93681

Epoch 00016: val_accuracy did not improve from 0.93681
Epoch 17/100
259/259 [=====] - 24s 95ms/step - loss: 0.6749 - accuracy: 0.93681

Epoch 00017: val_accuracy did not improve from 0.93681
Epoch 18/100
259/259 [=====] - 24s 94ms/step - loss: 0.6094 - accuracy: 0.93681

Epoch 00018: val_accuracy did not improve from 0.93681
Epoch 19/100
259/259 [=====] - 24s 94ms/step - loss: 0.6396 - accuracy: 0.93681

Epoch 00019: val_accuracy improved from 0.93681 to 0.94097, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 20/100
259/259 [=====] - 24s 94ms/step - loss: 0.5821 - accuracy: 0.94097

Epoch 00020: val_accuracy did not improve from 0.94097
Epoch 21/100
259/259 [=====] - 24s 94ms/step - loss: 0.5542 - accuracy: 0.94097

Epoch 00021: val_accuracy improved from 0.94097 to 0.94306, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 22/100
259/259 [=====] - 24s 94ms/step - loss: 0.5652 - accuracy: 0.94306

Epoch 00022: val_accuracy improved from 0.94306 to 0.96111, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
Epoch 23/100
259/259 [=====] - 24s 94ms/step - loss: 0.5099 - accuracy: 0.96111

Epoch 00023: val_accuracy did not improve from 0.96111
Epoch 24/100
259/259 [=====] - 24s 94ms/step - loss: 0.5439 - accuracy: 0.96111

Epoch 00024: val_accuracy did not improve from 0.96111
Epoch 25/100
259/259 [=====] - 24s 94ms/step - loss: 0.5219 - accuracy: 0.96111

Epoch 00025: val_accuracy improved from 0.96111 to 0.96458, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
```



```
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/bes
Epoch 26/100
259/259 [=====] - 24s 94ms/step - loss: 0.4968 - accuracy: 0.96458

Epoch 00026: val_accuracy did not improve from 0.96458
Epoch 27/100
259/259 [=====] - 24s 94ms/step - loss: 0.5384 - accuracy: 0.96458

Epoch 00027: val_accuracy did not improve from 0.96458
Epoch 28/100
259/259 [=====] - 24s 94ms/step - loss: 0.5284 - accuracy: 0.96458

Epoch 00028: val_accuracy did not improve from 0.96458
Epoch 29/100
259/259 [=====] - 24s 94ms/step - loss: 0.4781 - accuracy: 0.96458

Epoch 00029: val_accuracy did not improve from 0.96458
Epoch 30/100
259/259 [=====] - 24s 94ms/step - loss: 0.4540 - accuracy: 0.96458

Epoch 00030: val_accuracy improved from 0.96458 to 0.96875, saving model to /content/drive/MyDrive/nn_tasks/model-res/bes
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/bes
Epoch 31/100
259/259 [=====] - 24s 94ms/step - loss: 0.4485 - accuracy: 0.96875

Epoch 00031: val_accuracy did not improve from 0.96875
Epoch 32/100
259/259 [=====] - 24s 94ms/step - loss: 0.4749 - accuracy: 0.96875

Epoch 00032: val_accuracy did not improve from 0.96875
Epoch 33/100
259/259 [=====] - 24s 94ms/step - loss: 0.5126 - accuracy: 0.96875

Epoch 00033: val_accuracy did not improve from 0.96875
Epoch 34/100
259/259 [=====] - 24s 93ms/step - loss: 0.4417 - accuracy: 0.96875

Epoch 00034: val_accuracy did not improve from 0.96875
Epoch 35/100
259/259 [=====] - 24s 94ms/step - loss: 0.4410 - accuracy: 0.96875

Epoch 00035: val_accuracy did not improve from 0.96875
Epoch 36/100
259/259 [=====] - 24s 94ms/step - loss: 0.4482 - accuracy: 0.96875

Epoch 00036: val_accuracy did not improve from 0.96875
Epoch 37/100
259/259 [=====] - 24s 94ms/step - loss: 0.4558 - accuracy: 0.96875

Epoch 00037: val_accuracy did not improve from 0.96875
Epoch 38/100
259/259 [=====] - 24s 94ms/step - loss: 0.4198 - accuracy: 0.96875

Epoch 00038: val_accuracy did not improve from 0.96875
Epoch 39/100
259/259 [=====] - 24s 93ms/step - loss: 0.4046 - accuracy: 0.96875

Epoch 00039: val accuracy did not improve from 0.96875
```

```
Epoch 40/100
259/259 [=====] - 24s 93ms/step - loss: 0.4267 - accuracy: 0.96875

Epoch 00040: val_accuracy did not improve from 0.96875
Epoch 41/100
259/259 [=====] - 24s 94ms/step - loss: 0.3782 - accuracy: 0.96875

Epoch 00041: val_accuracy did not improve from 0.96875
Epoch 42/100
259/259 [=====] - 24s 94ms/step - loss: 0.4626 - accuracy: 0.96875

Epoch 00042: val_accuracy did not improve from 0.96875
Epoch 43/100
259/259 [=====] - 24s 93ms/step - loss: 0.4053 - accuracy: 0.96875

Epoch 00043: val_accuracy improved from 0.96875 to 0.97500, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model/assets
Epoch 44/100
259/259 [=====] - 24s 94ms/step - loss: 0.4063 - accuracy: 0.97500

Epoch 00044: val_accuracy did not improve from 0.97500
Epoch 45/100
259/259 [=====] - 24s 93ms/step - loss: 0.3764 - accuracy: 0.97500

Epoch 00045: val_accuracy did not improve from 0.97500
Epoch 46/100
259/259 [=====] - 24s 93ms/step - loss: 0.3950 - accuracy: 0.97500

Epoch 00046: val_accuracy did not improve from 0.97500
Epoch 47/100
259/259 [=====] - 24s 94ms/step - loss: 0.3843 - accuracy: 0.97500

Epoch 00047: val_accuracy did not improve from 0.97500
Epoch 48/100
259/259 [=====] - 24s 94ms/step - loss: 0.3664 - accuracy: 0.97500

Epoch 00048: val_accuracy improved from 0.97500 to 0.97917, saving model to /content/drive/MyDrive/nn_tasks/model-res/best_model.h5
INFO:tensorflow:Assets written to: /content/drive/MyDrive/nn_tasks/model-res/best_model/assets
Epoch 49/100
259/259 [=====] - 24s 94ms/step - loss: 0.3900 - accuracy: 0.97917

Epoch 00049: val_accuracy did not improve from 0.97917
Epoch 50/100
259/259 [=====] - 24s 94ms/step - loss: 0.3849 - accuracy: 0.97917

Epoch 00050: val_accuracy did not improve from 0.97917
Epoch 51/100
259/259 [=====] - 24s 93ms/step - loss: 0.3485 - accuracy: 0.97917

Epoch 00051: val_accuracy did not improve from 0.97917
Epoch 52/100
259/259 [=====] - 24s 94ms/step - loss: 0.3859 - accuracy: 0.97917

Epoch 00052: val_accuracy did not improve from 0.97917
Epoch 53/100
259/259 [=====] - 24s 94ms/step - loss: 0.3480 - accuracy: 0.97917
```