## Лабораторная работа №1. Логистическая регрессия в качестве нейронной сети

```
import os
try:
  import wget
except:
  !pip install wget
  import wget
import tarfile
out dir = 'data/not mnist'
small arhive = f'{out dir}/notMNIST small.tar.gz'
large arhive = f'{out dir}/notMNIST large.tar.gz'
large url = 'https://commondatastorage.googleapis.com/books1000/notMNIST large.tar.gz'
small url = 'https://commondatastorage.googleapis.com/books1000/notMNIST small.tar.gz'
if not os.path.exists(out dir):
  os.makedirs(out dir)
if not os.path.exists(small arhive):
  print(f"Downloading {small arhive}.")
  wget.download(small url, small arhive)
  print()
else:
  print(f"Skipping {small arhive} download (already exists)")
if not os.path.exists(large arhive):
  print(f"Downloading {large arhive}.")
  wget.download(large url, large arhive)
  print()
else:
  print(f"Skipping {large arhive} download (already exists)")
```

```
Skipping data/not mnist/notMNIST small.tar.gz download (already exists)
     Skipping data/not mnist/notMNIST large.tar.gz download (already exists)
print(f"Extracting {small arhive}")
with tarfile.open(small arhive) as tar:
  tar.extractall(out dir)
print(f"Extracting {large arhive}")
with tarfile.open(large arhive) as tar:
  tar.extractall(out dir)
Extracting data/not mnist/notMNIST small.tar.gz
    Extracting data/not mnist/notMNIST large.tar.gz
import numpy as np
from pathlib import Path
from PIL import Image
def remove duplicates(img train, labels train, img test):
    img new, labels new = [], []
    test set = {e.tostring() for e in img test}
    for i, (x, y) in enumerate(zip(img train, labels train)):
        if x.tostring() not in test set:
            img new.append(x)
            labels new.append(y)
    print(f'Removed {img train.shape[0] - len(img new)} duplicated images')
    return np.array(img new), np.array(labels new)
def load images(path, n):
    labels = ['I', 'G', 'A', 'F', 'H', 'J', 'C', 'D', 'E', 'B']
    x, y = [], []
    for i, 1 in enumerate(labels):
        d = Path(path) / 1
        print(f'Loading {str(d)} ', end='')
```

```
for j, f in zip(range(n), d.iterdir()):
            try:
                with Image.open(f) as img:
                    x.append(np.array(img))
                    y.append(i)
            except OSError:
                pass
            if j % 1000 == 0:
                print('.', end='', flush=True)
        print(flush=True)
    return np.array(labels), np.array(x), np.array(y)
def load not mnist data(path='data/not mnist/', use cache=True):
   train folder = Path(path) / 'notMNIST large'
   test folder = Path(path) / 'notMNIST small'
   train cache file = Path(path) / 'train.npz'
   test cache file = Path(path) / 'test.npz'
   if train cache file.exists() and test cache file.exists() and use cache:
        f = np.load(train cache file)
        labels, img train, labels train = [v for k, v in f.items()]
        f = np.load(test cache file)
        labels, img test, labels test = [v for k, v in f.items()]
        print('Loaded cached arrays')
    else:
        labels, img train, labels train = load images(train folder, 10000000)
        labels, img test, labels_test = load_images(test_folder, 10000000)
        np.savez(train cache file, labels, img train, labels train)
        np.savez(test cache file, labels, img test, labels test)
   return labels, img train, labels train, img test, labels test
load not mnist data()
```

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```
Loaded cached arrays
(array(['I', 'G', 'A', 'F', 'H', 'J', 'C', 'D', 'E', 'B'], dtype='<U1'),
array([[[180, 185, 214, ..., 217, 189, 184],
        [ 56, 58, 70, ..., 76, 66, 64],
             0, 0, ..., 0, 0,
                                      01,
        [0, 0, 0, \dots, 0, 0,
                                      01,
        [ 58, 60, 70, ..., 73, 64, 62],
        [177, 182, 211, \ldots, 209, 180, 174]
       [[255, 255, 255, \ldots, 54, 48, 48],
        [255, 255, 255, ..., 54, 48,
                                     481,
        [255, 255, 255, ..., 54, 48,
                                     481,
        . . . ,
        [255, 255, 255, ..., 54, 48,
                                     481,
        [255, 255, 255, ..., 54, 48,
                                     481,
        [255, 255, 255, ..., 54, 48,
                                     48]],
                    0, ..., 0, 0,
       [[ 0,
                                      0],
        [ 0,
               0, 0, ..., 0, 0,
                                      0],
        [ 0,
              0, 0, ...,
                            0, 0,
                                      0],
        . . . ,
        [0, 0, 0, \dots, 0, 1,
                                      0],
        [0, 0, 0, \dots, 0, 0,
                                     0],
        [ 0,
             0, 0, ..., 0, 0,
                                      0]],
       . . . ,
       [[0, 0, 0, \dots, 1, 0,
                                      0],
          0, 0, 0, ..., 0, 1,
                                      0],
          0,
             0, 0, ...,
                                      01,
         0, 0, 0, ...,
                                      0],
        [ 0,
               0, 0, ...,
                             0, 0,
                                      0],
        [ 0,
                             0,
                    0, ...,
                                  0,
                                      0]],
       [[164, 175, 185, \ldots,
                             0, 0,
                                      0],
        [255, 255, 255, ...,
                                      0],
        [243, 253, 252, ...,
                             0,
                                      0],
        . . . ,
        [241, 255, 254, ...,
                                 1,
                                      0],
        [241, 255, 254, ...,
                             1,
                                      0],
```

```
[239, 250, 245, ..., 0, 0,
                                        0]],
                    0, \ldots, 27, 0,
      [[ 0,
                                        2],
       [ 0,
                    0, ..., 234, 68,
                                        01,
                    0, ..., 255, 232,
               0,
       . . . ,
       [ 30, 91, 89, ...,
                              2,
                                        0],
       [128, 255, 253, \ldots, 0,
                                        0],
       [176, 255, 249, ..., 0,
                                   0.
                                        0]]], dtype=uint8),
array([0, 0, 0, ..., 9, 9, 9]),
array([[[ 81, 96, 138, ..., 124, 60, 37],
       [131, 153, 215, \ldots, 191, 94, 59],
       [122, 145, 210, ..., 188, 90, 55],
       [124, 147, 212, \ldots, 190, 92, 56],
       [125, 148, 213, \ldots, 190, 92, 56],
       [120, 143, 206, \ldots, 184, 89, 54]],
               0, 0, ..., 134, 150, 110],
      [[ 0,
              0, 0, ..., 255, 145, 401,
             0, 0, ..., 132, 0,
       . . . ,
       [100, 255, 133, \ldots, 0,
                                        0],
       [ 11, 172, 238, ...,
                             0, 0,
                                        0],
                            0,
       [ 0, 0, 98, ...,
                                  0,
                                        011,
      [[8, 16, 23, ..., 107, 116, 119],
       [176, 170, 166, \ldots, 244, 251, 255],
       [211, 236, 255, ..., 255, 254, 253],
       [202, 233, 255, ..., 255, 255, 255],
       [212, 233, 247, ..., 227, 237, 244],
       [110, 98, 89, \ldots, 71, 62, 48]],
       . . . ,
      [[13, 0, 6, \ldots, 0, 0,
                                        0],
                              0, 1,
       [236, 189, 216, ...,
                                        0],
       [190, 255, 255, ...,
                                        0],
                              0,
       [159, 249, 255, ...,
                                   2,
                                        0],
       [125, 161, 239, ...,
                                        0],
```

```
[0, 17, 114, \ldots, 0, 0, 0],
           [[0, 0, 0, \dots, 0, 2, 0],
           [0, 0, 0, \ldots, 136, 2,
                                        0],
           [ 0, 0, 0, ..., 255, 143,
           [0, 0, 3, \ldots, 2, 0,
                                        01,
           [0, 1, 2, \ldots, 0, 0, 0],
           [137, 210, 217, ..., 0, 0, 0]],
          [[217, 224, 246, ..., 0, 0,
                                        01,
           [4, 13, 42, \ldots, 3, 0,
                                        01,
           [0, 1, 0, \ldots, 0, 2,
                                        01,
           [0, 1, 0, \ldots, 80, 0, 3],
           [6, 20, 55, \ldots, 0, 3, 0],
           [219, 229, 252, ..., 3, 0, 0]]], dtype=uint8),
     array([0, 0, 0, ..., 9, 9, 9]))
labels, img train, labels train, img test, labels test = load not mnist data()

    Loaded cached arrays
```

## Задание 1. Загрузите данные и отобразите на экране несколько из изображений с помощью языка Python;

```
import matplotlib.pyplot as plt

plt.rcParams['font.size'] = 14
plt.rcParams['font.family'] = 'consolas'

rows = 2
cols = 3
fig = plt.figure(figsize=(16, 6.5))
for i in range(1, cols * rows + 1):
    ax = fig.add_subplot(rows, cols, i)
    ax.set_xticks([])
    ax.set_yticks([])
```

```
j = np.random.randint(0, labels test.shape[0] - 1)
   plt.imshow(img_test[j], cmap='gray')
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```

Задание 2. Проверьте, что классы являются сбалансированными, т.е. количество изображений, принадлежащих каждому из классов, примерно одинаково (В данной задаче 10 классов).

```
from collections import defaultdict

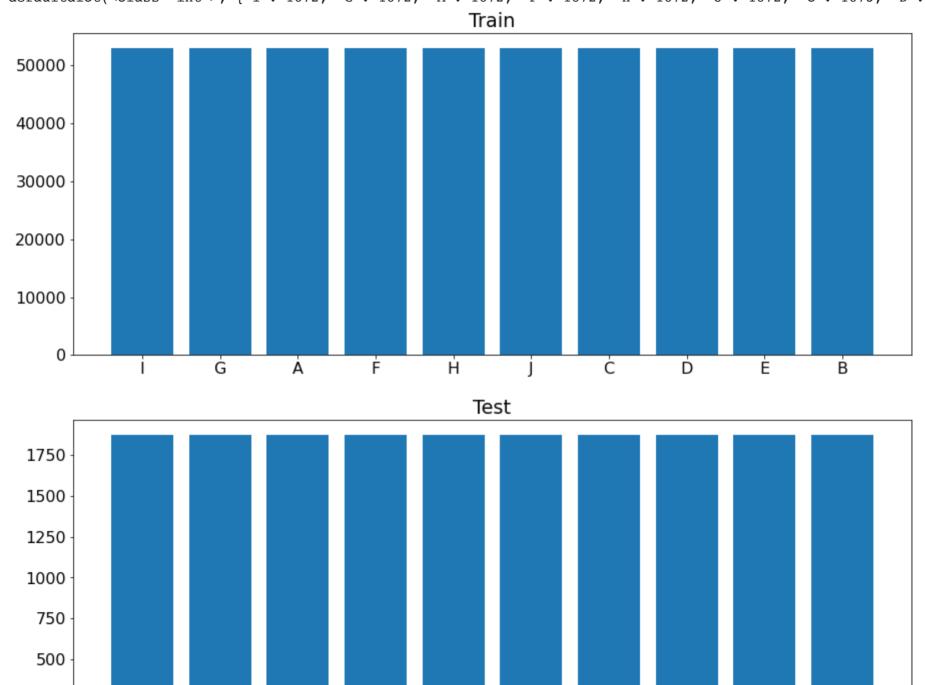
plt.rcParams['font.size'] = 16

def show_balance(title, ax, a, labels):
    counts = defaultdict(int)
    for e in a:
        counts[labels[e]] += 1
```

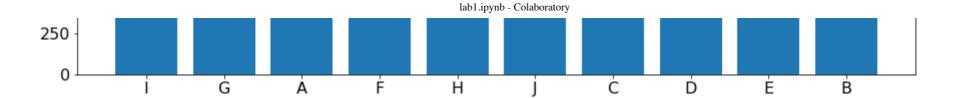
```
ax.bar(counts.keys(), counts.values())
ax.set_title(title)

fig = plt.figure(figsize=(15, 13), )
show_balance("Train", fig.add_subplot(2, 1, 1), labels_train, labels)
show_balance("Test", fig.add_subplot(2, 1, 2), labels_test, labels)
```

defaultdict(<class 'int'>, {'I': 52912, 'G': 52912, 'A': 52909, 'F': 52912, 'H': 52912, 'J': 52911, 'C': 52912, 'D': ! defaultdict(<class 'int'>, {'I': 1872, 'G': 1872, 'A': 1872, 'F': 1872, 'H': 1872, 'J': 1872, 'C': 1873, 'D': 1873, 'I







## Задание 3,4 Разделение на обучающую, валидационную и контрольную выборки. Удаление дубликатов

```
from sklearn.linear_model import SGDClassifier, LogisticRegression
from sklearn.model_selection import train_test_split

def flatten(a):
    return a.reshape(a.shape[0], a.shape[1] * a.shape[2])

def load_data():
    labels, img_train, labels_train, img_test, labels_test = load_not_mnist_data()
    img_train, labels_train = remove_duplicates(img_train, labels_train, img_test)
    return labels, flatten(img_train), labels_train, flatten(img_test), labels_test

_, x_train, y_train, x_test, y_test = load_data()
x_train, x_val, y_train, y_val = train_test_split(x_train, y_train, test_size=0.1)

_} Loaded cached arrays
    Removed 12213 duplicated images
```

Задание 5. Постройте простейший классификатор (например, с помощью логистической регрессии). Постройте график зависимости точности классификатора от размера обучающей выборки (50, 100, 1000, 50000)

```
results = {}
results.setdefault('val_acc', {})
ns = [50, 100, 1000, 50000]
```

```
print('Training Logistic Regression model:')
for n in ns:
  indices = np.arange(len(x train))
  if n < len(x train):</pre>
    indices = np.random.choice(indices, n, replace=False)
 # model = SGDClassifier(loss='log', tol=1e-4, early stopping=True)
 model = LogisticRegression(solver='lbfgs', max iter = 1000000)
 model.fit(x train[indices], y train[indices])
  y pred = model.predict(x val)
  results["val acc"][str(n)] = acc = np.mean(np.equal(y pred, y val).astype(np.int))
 print(f"n = {n}, accuracy = {acc:.5f}, iterations = {model.n iter }")
 Training Logistic Regression model:
    n = 50, accuracy = 0.56613, iterations = [60]
    n = 100, accuracy = 0.69766, iterations = [97]
    n = 1000, accuracy = 0.75181, iterations = [486]
    n = 50000, accuracy = 0.81211, iterations = [4805]
print (results["val acc"].keys())
print (results["val acc"].values())
□ dict keys(['50', '100', '1000', '50000'])
     dict values([0.566133369445358, 0.6976649706912228, 0.7518136619527578, 0.8121142945580468])
# The Data
x = list(results["val acc"].keys())
y = list(results["val acc"].values())
# Create the figure and axes objects
fig, ax = plt.subplots(1, figsize=(8, 6))
fig.suptitle('Logistic Regression')
# Plot the data
ax.plot(x,y)
```

# Snow the grid lines as dark grey lines
plt.grid(b=True, which='major', color='#666666', linestyle='-')
plt.show()

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## Logistic Regression

