### Задание

Берем отызывы за лето (из архива с материалами или предыдущего занятия)

- 1. Учим conv сеть для классификации
- 2. Рассмотреть 2-а варианта сеточек
- 3. 1 Инициализировать tf.keras.layers.Embedding предобученными векторами взять к примеру с <a href="https://rusvectores.org/ru/">https://rusvectores.org/ru/</a>)
- 4. 2 Инициализировать слой tf.keras.layers.Embedding по умолчанию (ну то есть вам ничего не делать с весами)

Сравнить две архитектуры с предобученными весами и когда tf.keras.layers.Embedding обучается сразу со всей сеточкой, что получилось лучше

# In [4]:

```
max_words = 200
max_len = 40
num_classes = 1

# Training
epochs = 20
batch_size = 128
print_batch_n = 100
```

## In [5]:

```
import pandas as pd

df_train = pd.read_csv("data/train.csv")

df_test = pd.read_csv("data/test.csv")

df_val = pd.read_csv("data/val.csv")
```

## In [6]:

```
df_train.head()
```

### Out[6]:

|   | id | text                                           | class |  |  |
|---|----|------------------------------------------------|-------|--|--|
| 0 | 0  | @alisachachka не уезжаааааааай. :(♥ я тоже не  | 0     |  |  |
| 1 | 1  | RT @GalyginVadim: Ребята и девчата!\nВсе в кин |       |  |  |
| 2 | 2  | RT @ARTEM_KLYUSHIN: Кто ненавидит пробки ретви |       |  |  |
| 3 | 3  | RT @epupybobv: Хочется котлету по-киевски. Зап |       |  |  |
| 4 | 4  | @KarineKurganova @Yess Вoss босапопа есбоса н  | 1     |  |  |

### In [7]:

```
import re
def funk_del(input_txt):
    pattern = "@[\w]*"
    if re.findall(pattern, input_txt):
        return re.sub(pattern, ' ', input_txt)
    else:
        return re.sub(pattern, ' ', input_txt)
```

### In [8]:

```
# df_train['text'] = df_train['text'].apply(funk_del)
# df_test['text'] = df_test['text'].apply(funk_del)
# df_val['text'] = df_val['text'].apply(funk_del)
```

# In [9]:

```
# pattern = r'[^\w\s]'
# df_train['text'] = df_train['text'].apply(lambda x: re.sub(pattern, ' ', x))
# df_test['text'] = df_test['text'].apply(lambda x: re.sub(pattern, ' ', x))
# df_val['text'] = df_val['text'].apply(lambda x: re.sub(pattern, ' ', x))
```

### In [10]:

```
# pattern = r'[^a-яА-ЯО-9]'
# df_train['text'] = df_train['text'].apply(lambda x: re.sub(pattern, ' ', x))
# df_test['text'] = df_test['text'].apply(lambda x: re.sub(pattern, ' ', x))
# df_val['text'] = df_val['text'].apply(lambda x: re.sub(pattern, ' ', x))
```

## In [11]:

```
df_train.head()
```

### Out[11]:

|   | id | text                                           | class |
|---|----|------------------------------------------------|-------|
| 0 | 0  | @alisachachka не уезжаааааааай. :(♥ я тоже не  | 0     |
| 1 | 1  | RT @GalyginVadim: Ребята и девчата!\nВсе в кин | 1     |
| 2 | 2  | RT @ARTEM_KLYUSHIN: Кто ненавидит пробки ретви | 0     |
| 3 | 3  | RT @epupybobv: Хочется котлету по-киевски. Зап | 1     |
| 4 | 4  | @KarineKurganova @YessВoss босапопа есбоса н   | 1     |

### In [12]:

```
from string import punctuation
from stop_words import get_stop_words
from pymorphy2 import MorphAnalyzer
```

# In [13]:

```
sw = set(get_stop_words("ru"))
exclude = set(punctuation)
morpher = MorphAnalyzer()

def preprocess_text(txt):
    txt = str(txt)
    txt = re.sub(r'[^a-яА-Я]', ' ', txt) # Заменим спец. символы на пробелы
    txt = funk_del(txt) # Удалим @word из всех мвитов

# txt = "".join(c for c in txt if c not in exclude)
    txt = txt.lower()
    txt = re.sub("\she", "He", txt)
    txt = [morpher.parse(word)[0].normal_form for word in txt.split() if word not in sw]
    return " ".join(txt)

df_train['text'] = df_train['text'].apply(preprocess_text)
df_val['text'] = df_val['text'].apply(preprocess_text)
df_test['text'] = df_test['text'].apply(preprocess_text)
```

### In [14]:

```
df_train.head()
```

### Out[14]:

|   | id | text                                           | class |
|---|----|------------------------------------------------|-------|
| 0 | 0  | уезжаааааааать тожена хотеть уезжать           | 0     |
| 1 | 1  | ребята девчата кино любовь завтра вотэтолюбовь | 1     |
| 2 | 2  | ктоненавидеть пробка ретвит                    | 0     |
| 3 | 3  | хотеться котлета киевск запретный плод         | 1     |
| 4 | 4  | босапоп есбосан бояться мороз                  | 1     |

## In [15]:

```
train_corpus = " ".join(df_train["text"])
```

```
In [16]:
```

```
import nltk
from nltk.tokenize import word_tokenize
nltk.download("punkt")
```

## Out[16]:

True

## In [17]:

```
tokens = word_tokenize(train_corpus)
```

### Отфильтруем данные

и соберём в корпус N наиболее частых токенов

## In [18]:

```
tokens_filtered = [word for word in tokens if word.isalnum()]
```

# In [19]:

```
len(set(tokens_filtered))
```

## Out[19]:

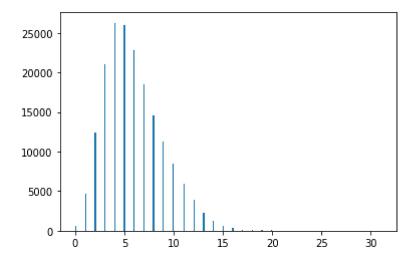
96559

## In [20]:

```
%matplotlib inline
import matplotlib.pyplot as plt
_, _, hist = plt.hist(df_train.text.apply(lambda text: len(text.split())), bins='auto')
hist
```

## Out[20]:

<BarContainer object of 220 artists>



## In [21]:

```
num_tok = 0
for i, row in df_train.iterrows():
    line = len(row['text'].split())
    if num_tok < line:
        num_tok = line
num_tok</pre>
```

## Out[21]:

31

## In [22]:

```
max_words = 12000
max_len = 31
```

# In [23]:

```
from nltk.probability import FreqDist
dist = FreqDist(tokens_filtered)
tokens_filtered_top = [pair[0] for pair in dist.most_common(max_words-1)]
```

до удаления числовых символов

```
In [71]:
```

```
tokens_filtered_top[:10]
Out[71]:
['хотеть', '3', '1', '2', '4', 'знать', '5', '9', 'любить', 'ян']
после удаления, слова больше смысла имеют
In [24]:
tokens_filtered_top[:20]
Out[24]:
['хотеть',
 'знать',
 'любить',
 'ян',
 'мочь',
 'завтра',
 'новый',
 'вс',
 'мой',
 'хороший',
 'делать',
 'смотреть',
 'блин',
 'думать',
 'вообще',
 'день',
 'спать',
 'идти',
 'дом',
 'самый']
In [25]:
vocabulary = {v: k for k, v in enumerate(tokens filtered top, 1)}
```

```
In [26]:
```

```
vocabulary
Out[26]:
{'хотеть': 1,
 'знать': 2,
 'любить': 3,
 'ян': 4,
 'мочь': 5,
 'завтра': 6,
 'новый': 7,
 'вс': 8,
 'мой': 9,
 'хороший': 10,
 'делать': 11,
 'смотреть': 12,
 'блин': 13,
 'думать': 14,
 'вообще': 15,
 'день': 16,
 'спать': 17,
 'идти': 18.
In [27]:
import numpy as np
def text_to_sequence(text, maxlen):
    result = []
    tokens = word_tokenize(text.lower())
    tokens_filtered = [word for word in tokens if word.isalnum()]
    for word in tokens_filtered:
        if word in vocabulary:
            result.append(vocabulary[word])
    padding = (maxlen-len(result)) * [0]
    return result[-maxlen:] + padding
In [28]:
x_train = np.asarray([text_to_sequence(text, max_len) for text in df_train["text"]], dtype=
x_test = np.asarray([text_to_sequence(text, max_len) for text in df_test["text"]], dtype=np
x_val = np.asarray([text_to_sequence(text, max_len) for text in df_val["text"]], dtype=np.i
In [29]:
x train.shape
Out[29]:
(181467, 31)
```

```
In [30]:
```

```
x_train[1]
```

# Out[30]:

```
array([ 191, 6285, 432,
                           149,
                                    6,
                                          0,
                                                 0,
                                                        0,
                                                              0,
                                                                    0,
                                                                           0,
                             0,
                                          0,
          0,
                 0,
                       0,
                                    0,
                                                 0,
                                                        0,
                                                              0,
                                                                    0,
                                                                           0,
                                    0,
          0,
                 0,
                       0,
                              0,
                                          0,
                                                 0,
                                                        0,
                                                              01)
```

# Keras model

```
In [31]:
```

```
import numpy as np
# import keras
from tensorflow import keras
from keras.models import Sequential, Model
from keras.layers import Dense, Dropout, Activation, Input, Embedding, Conv1D, GlobalMaxPoo
from keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
from keras.callbacks import TensorBoard
from keras import losses
# from keras.objectives import categorical_crossentropy
from keras.callbacks import EarlyStopping
```

### In [32]:

```
# keras.losses.categorical_crossentropy(y_true, y_pred, from_logits=False, label_smoothing=
```

# In [33]:

```
num_classes = 2
y_train = keras.utils.to_categorical(df_train["class"], num_classes)
y_val = keras.utils.to_categorical(df_val["class"], num_classes)
```

### In [34]:

```
# ?GlobalMaxPool1D
```

# In [35]:

```
import keras
print('keras: %s' % keras.__version__)
```

keras: 2.9.0

## In [36]:

```
num_classes = 2

# Training
epochs = 20
batch_size = 512
print_batch_n = 100
```

# In [55]:

```
model = Sequential()
model.add(Embedding(input_dim=max_words, output_dim=128, input_length=max_len))
model.add(Conv1D(128, 3))
model.add(Activation("relu"))
model.add(GlobalMaxPool1D())
model.add(Dense(10))
model.add(Activation("relu"))
model.add(Dense(num_classes))
model.add(Activation('softmax'))
model.summary()
```

Model: "sequential\_10"

| Layer (type)                                            | Output Shape    | Param # |
|---------------------------------------------------------|-----------------|---------|
| embedding_10 (Embedding)                                | (None, 31, 128) | 1536000 |
| conv1d_20 (Conv1D)                                      | (None, 29, 128) | 49280   |
| activation_36 (Activation)                              | (None, 29, 128) | 0       |
| <pre>global_max_pooling1d_7 (Glo balMaxPooling1D)</pre> | (None, 128)     | 0       |
| dense_16 (Dense)                                        | (None, 10)      | 1290    |
| activation_37 (Activation)                              | (None, 10)      | 0       |
| dense_17 (Dense)                                        | (None, 2)       | 22      |
| activation_38 (Activation)                              | (None, 2)       | 0       |
|                                                         |                 | ======= |

Total params: 1,586,592

Trainable params: 1,586,592 Non-trainable params: 0

### In [43]:

```
model1 = Sequential()
model1.add(Embedding(input_dim=max_words, output_dim=512, input_length=max_len))
model1.add(Conv1D(512, 3, padding = 'same'))
model1.add(Activation("tanh"))
model1.add(MaxPool1D(pool_size=3, strides=3))
# model.add(GlobalMaxPool1D())
model1.add(Conv1D(256, 3, padding = 'same'))
model1.add(Activation("tanh"))
model1.add(MaxPool1D(pool_size=3, strides=3))
# model.add(GLobalMaxPool1D())
model1.add(Conv1D(128, 2, padding = 'same'))
model1.add(Activation("tanh"))
# model1.add(MaxPool1D(pool_size=2, strides=1))
model1.add(GlobalMaxPool1D())
model1.add(Dense(10))
model1.add(Activation("relu"))
model1.add(Dense(num_classes))
model1.add(Activation('softmax'))
model1.summary()
```

Model: "sequential\_6"

| Layer (type)                                            | Output Shape    | Param #            |
|---------------------------------------------------------|-----------------|--------------------|
| embedding_6 (Embedding)                                 |                 | =======<br>6144000 |
| conv1d_14 (Conv1D)                                      | (None, 31, 512) | 786944             |
| activation_22 (Activation)                              | (None, 31, 512) | 0                  |
| <pre>max_pooling1d_7 (MaxPooling 1D)</pre>              | (None, 10, 512) | 0                  |
| conv1d_15 (Conv1D)                                      | (None, 10, 256) | 393472             |
| activation_23 (Activation)                              | (None, 10, 256) | 0                  |
| <pre>max_pooling1d_8 (MaxPooling 1D)</pre>              | (None, 3, 256)  | 0                  |
| conv1d_16 (Conv1D)                                      | (None, 3, 128)  | 65664              |
| activation_24 (Activation)                              | (None, 3, 128)  | 0                  |
| <pre>global_max_pooling1d_3 (Glo balMaxPooling1D)</pre> | (None, 128)     | 0                  |
| dense_8 (Dense)                                         | (None, 10)      | 1290               |
| activation_25 (Activation)                              | (None, 10)      | 0                  |
| dense_9 (Dense)                                         | (None, 2)       | 22                 |
| activation_26 (Activation)                              | •               | 0                  |

Total params: 7,391,392 Trainable params: 7,391,392 Non-trainable params: 0

### In [56]:

### In [45]:

### In [58]:

модель где три conv1d слоя обучалась гораздо дольше, но результат на валидации немного лучше, а ассигасу больше на меньше количестве слоев, думаю вторая легче переобучается, за такой надо внимательнее следить

```
In [ ]:
```

```
In [37]:
```

```
token_dict = {}
```

читаю веса токенов из скачаной модели

```
In [38]:
```

```
p = r'[a-zA-Za-яA-Я]'
with open('model.txt','r',buffering=100000, encoding='utf-8') as f:
    for line in f:
        line = line.split('_')
        if len(line)==1:
            pass
        else:
            k = line[0]
            v = re.sub(p, '', line[1]).split()
#
              print(v)
            v = [re.sub('-', '', i) for i in v]
              print(v)
#
            v = [float(i) for i in v]
            token_dict[k] = v
              token\_dict[line.split('_')[0]] = re.sub(p, '', line.split('_')[1]).split()
#
              token_dict = {i:j for i,j in line.split('_')}
#
          print (line, '\n', '-'*30, '\n')
#
```

### In [39]:

```
len(token_dict)
```

## Out[39]:

205707

### In [52]:

```
a = [1,2,3,4,5]
np.mean(a)
```

### Out[52]:

3.0

беру среднее каждого вектора токена, взять все 300 для обучения у меня не получилось, пк не справился

```
In [58]:
```

### In [56]:

```
from tqdm import tqdm_notebook
```

### In [59]:

```
x_train = np.asarray([text_to_sequence(text, maxlen) for text in tqdm_notebook(df_train["te
x_test = np.asarray([text_to_sequence(text, maxlen) for text in tqdm_notebook(df_test["text
x_val = np.asarray([text_to_sequence(text, maxlen) for text in tqdm_notebook(df_val["text"]
```

```
c:\program files\python37\lib\site-packages\ipykernel_launcher.py:1: TqdmDep
recationWarning: This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
   """Entry point for launching an IPython kernel.
```

100%

181467/181467 [01:26<00:00, 3639.03it/s]

c:\program files\python37\lib\site-packages\ipykernel\_launcher.py:2: TqdmDep
recationWarning: This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm\_notebook`

100%

22684/22684 [00:25<00:00, 2228.24it/s]

c:\program files\python37\lib\site-packages\ipykernel\_launcher.py:3: TqdmDep
recationWarning: This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm\_notebook`

This is separate from the ipykernel package so we can avoid doing imports until

100%

22683/22683 [00:07<00:00, 3843.92it/s]

### In [44]:

31\*300

### Out[44]:

9300

```
In [60]:
```

```
x_train.shape
```

## Out[60]:

(181467, 31)

## In [61]:

```
len(x_train[1])
```

## Out[61]:

31

## In [62]:

```
model3 = Sequential()
model3.add(Embedding(input_dim=x_train.shape[0], output_dim=batch_size, input_length=x_trai
model3.add(Conv1D(128, 3))
model3.add(Activation("relu"))
model3.add(GlobalMaxPool1D())
model3.add(Dense(10))
model3.add(Activation("relu"))
model3.add(Dense(num_classes))
model3.add(Activation('softmax'))
model3.summary()
```

Model: "sequential\_1"

| Layer (type)                                            | Output Shape                            | Param #  |
|---------------------------------------------------------|-----------------------------------------|----------|
| embedding_1 (Embedding)                                 | (None, 31, 512)                         | 92911104 |
| conv1d_1 (Conv1D)                                       | (None, 29, 128)                         | 196736   |
| activation_3 (Activation)                               | (None, 29, 128)                         | 0        |
| <pre>global_max_pooling1d_1 (Glo balMaxPooling1D)</pre> | (None, 128)                             | 0        |
| dense_2 (Dense)                                         | (None, 10)                              | 1290     |
| activation_4 (Activation)                               | (None, 10)                              | 0        |
| dense_3 (Dense)                                         | (None, 2)                               | 22       |
| activation_5 (Activation)                               | (None, 2)                               | 0        |
| =======================================                 | ======================================= | ======== |

Total params: 93,109,152 Trainable params: 93,109,152 Non-trainable params: 0

### In [63]:

### In [64]:

точность стоит на месте, может стоит попробывать добавить слои конвулиционные, но у меня уже компьютер на пределе работает, из трех моделей лучше всего я считаю показала себя с тремя conv1d слоями обученная на самостоятельно построенных эмбедингах, ее стоит обучить побольше

| ın [ ]: |  |  |  |  |
|---------|--|--|--|--|
|         |  |  |  |  |
|         |  |  |  |  |