

Практическая работа №2

Потоковая аналитика эмоциональных оттенков речи человека

Цель работы:

Выполнение практической работы направлено на изучение современных программных средств определения эмоционального фона речи человека в голосе и в тексте.

Порядок работы:

1. Определение эмоционального фона речи человека в текстовой форме с использованием модуля *Text* библиотеки *Aniemore* для *Python*. Проверка модели *rubert-tiny-emotion-russian-cedr-m7*. Программа определила эмоциональный фон как «нейтральный» с точностью 38%, как показано на рисунке 1.

```
✓ 1
CEK. # @title Imports and variables settings
from rich import print # for better view
import torch
from aniemore.recognizers.text import TextRecognizer
from aniemore.models import HuggingFaceModel

model = HuggingFaceModel.Text.Bert_Tiny
device = 'cuda' if torch.cuda.is_available() else 'cpu'

# tr - acronym for TextRecognizer from two first capital letters
tr = TextRecognizer(model=model, device=device)

✓ 0
CEK. [20] # @title Recognize: single text - single label
text = 'если ты не прекратишь, я тебя убью!'
result = tr.recognize(text, return_single_label=True), tr.recognize(text, return_single_label=False)
print(result)

(
  'neutral',
  {
    'anger': 0.33996066451072693,
    'disgust': 0.025116577744483948,
    'enthusiasm': 0.018978049978613853,
    'fear': 0.1447315365076065,
    'happiness': 0.012643322348594666,
    'neutral': 0.3827205002307892,
    'sadness': 0.2320241630077362
  }
)
```

Рисунок 1 - Проверка модели архитектуры *Bert_Tiny*

2. Проверка модели *rubert-tiny2-russian-emotion-detection*. Программа определила эмоциональный фон как «ГНЕВНЫЙ» с точностью 80%, как показано на рисунке 2.

```
Imports and variables settings

# @title Imports and variables settings
from rich import print # for better view
import torch
from aniemore.recognizers.text import TextRecognizer
from aniemore.models import HuggingFaceModel

model = HuggingFaceModel.Text.Bert_Tiny2
device = 'cuda' if torch.cuda.is_available() else 'cpu'

# tr - acronym for TextRecognizer from two first capital letters
tr = TextRecognizer(model=model, device=device)

Recognize: single text - single label

[14] # @title Recognize: single text - single label
text = 'если ты не прекратишь, я тебя убью!'
result = tr.recognize(text, return_single_label=True), tr.recognize(text, return_single_label=False)
print(result)

(
  'anger',
  {
    'neutral': 0.08041860908269882,
    'happiness': 0.02092890627682209,
    'sadness': 0.16193069517612457,
    'enthusiasm': 0.034999947994947433,
    'fear': 0.0488605834543705,
    'anger': 0.7970464825630188,
    'disgust': 0.03219384700059891
  }
)
```

Рисунок 2 - Проверка модели архитектуры *Bert_Tiny2*

3. Проверка модели *rubert-large-emotion-russian-cedr-m7*. Программа определила эмоциональный фон как «ГНЕВНЫЙ» с точностью 78%, как показано на рисунке 3.

```
Imports and variables settings

# @title Imports and variables settings
from rich import print # for better view
import torch
from aniemore.recognizers.text import TextRecognizer
from aniemore.models import HuggingFaceModel

model = HuggingFaceModel.Text.Bert_Large
device = 'cuda' if torch.cuda.is_available() else 'cpu'

# tr - acronym for TextRecognizer from two first capital letters
tr = TextRecognizer(model=model, device=device)

Recognize: single text - single label

[18] # @title Recognize: single text - single label
text = 'если ты не прекратишь, я тебя убью!'
result = tr.recognize(text, return_single_label=True), tr.recognize(text, return_single_label=False)
print(result)

(
  'anger',
  {
    'anger': 0.7758040428161621,
    'disgust': 0.03686130419373512,
    'enthusiasm': 0.03008371964097023,
    'fear': 0.10172183066606522,
    'happiness': 0.07250823080539703,
    'neutral': 0.09607995301485062,
    'sadness': 0.0768846720457077
  }
)
```

Рисунок 3 - Проверка модели архитектуры *Bert_Large*

4. Проверка модели *rubert-base-emotion-russian-cedr-m7*. Программа определила эмоциональный фон как «ГНЕВНЫЙ» с точностью 43%, как показано на рисунке 4.

```
Imports and variables settings

# @title Imports and variables settings
from rich import print # for better view
import torch
from aniemore.recognizers.text import TextRecognizer
from aniemore.models import HuggingFaceModel

model = HuggingFaceModel.Text.Bert_Base
device = 'cuda' if torch.cuda.is_available() else 'cpu'

# tr - acronym for TextRecognizer from two first capital letters
tr = TextRecognizer(model=model, device=device)

Recognize: single text - single label

[16] # @title Recognize: single text - single label
text = 'если ты не прекратишь, я тебя убью!'
result = tr.recognize(text, return_single_label=True), tr.recognize(text, return_single_label=False)
print(result)

(
  'anger',
  {
    'anger': 0.4286688268184662,
    'disgust': 0.01858402229845524,
    'enthusiasm': 0.028649266809225082,
    'fear': 0.06029261648654938,
    'happiness': 0.04082256928086281,
    'neutral': 0.3020907938480377,
    'sadness': 0.09338401257991791
  }
)
```

Рисунок 4 - Проверка модели архитектуры *Bert_Base*

5. Подготовка к определению эмоционального фона речи человека в голосовой форме с использованием модуля *Voice* библиотеки *Aniemore* для *Python*. Загрузка файла голосовой записи представлена на рисунке 5.

```
Load our test voice1 file

[21] # @title Load our test voice1 file
import wget
from pathlib import Path
voice_url = "https://github.com/aniemore/Aniemore/blob/master/tests/aniemore/recognizers/src/"
voices_path: Path = Path('my_voice.ogg')

wget.download(voice_url, str('my_voice.ogg'))

'my_voice.ogg'
```

Рисунок 5 - Загрузка звукового файла

6. Проверка модели *wavlm-emotion-russian-resd*. Программа определила эмоциональный фон как «отвращенный» с точностью 71%, как показано на рисунке 6.

✓ Recognize: single file - all labels

```
# @title Recognize: single file - all labels
from rich import print
import torch
from aniemore.recognizers.voice import VoiceRecognizer
from aniemore.models import HuggingFaceModel

model = HuggingFaceModel.Voice.WavLM
device = 'cuda' if torch.cuda.is_available() else 'cpu'

vr = VoiceRecognizer(model, device)
result = vr.recognize(voice_files[0]), vr.recognize(voice_files[0], return_single_label=True)
print(result)
```

Some weights of the model checkpoint at aniemore/wavlm-emotion-russian-resd were not used when
- This IS expected if you are initializing WavLMForSequenceClassification from the checkpoint o
- This IS NOT expected if you are initializing WavLMForSequenceClassification from the checkpoi
Some weights of WavLMForSequenceClassification were not initialized from the model checkpoint a
You should probably TRAIN this model on a down-stream task to be able to use it for predictions
Some weights of the model checkpoint at aniemore/wavlm-emotion-russian-resd were not used when
- This IS expected if you are initializing WavLMForSequenceClassification from the checkpoint o
- This IS NOT expected if you are initializing WavLMForSequenceClassification from the checkpoi
Some weights of WavLMForSequenceClassification were not initialized from the model checkpoint a
You should probably TRAIN this model on a down-stream task to be able to use it for predictions
(
 {
 'anger': 0.0002296616294188425,
 'disgust': 0.707155704498291,
 'enthusiasm': 0.0616530179977417,
 'fear': 9.370104089612141e-05,
 'happiness': 8.026974683161825e-05,
 'neutral': 0.2295886129140854,
 'sadness': 0.0011990233324468136
 },
 'disgust'
)

Рисунок 6 - Проверка модели архитектуры *WavLM*

7. Проверка модели *wav2vec2-xlsr-53-russian-emotion-recognition*.

Программа определила эмоциональный фон как «нейтральный» с точностью 99%, как показано на рисунке 7.



```
# @title Recognize: single file - all labels
from rich import print
import torch
from aniemore.recognizers.voice import VoiceRecognizer
from aniemore.models import HuggingFaceModel

model = HuggingFaceModel.Voice.Wav2Vec2
device = 'cuda' if torch.cuda.is_available() else 'cpu'

vr = VoiceRecognizer(model, device)
result = vr.recognize(voice_files[0]), vr.recognize(voice_files[0], return_single_label=True)
print(result)
```

Some weights of the model checkpoint at aniemore/wav2vec2-emotion-russian-resd were not used w
- This IS expected if you are initializing Wav2Vec2ForSequenceClassification from the checkpoi
- This IS NOT expected if you are initializing Wav2Vec2ForSequenceClassification from the chec
Some weights of Wav2Vec2ForSequenceClassification were not initialized from the model checkpoi
You should probably TRAIN this model on a down-stream task to be able to use it for prediction
Some weights of the model checkpoint at aniemore/wav2vec2-emotion-russian-resd were not used w
- This IS expected if you are initializing Wav2Vec2ForSequenceClassification from the checkpoi
- This IS NOT expected if you are initializing Wav2Vec2ForSequenceClassification from the chec
Some weights of Wav2Vec2ForSequenceClassification were not initialized from the model checkpoi
You should probably TRAIN this model on a down-stream task to be able to use it for prediction

```
(
  {
    'anger': 1.5641922800568864e-05,
    'disgust': 0.0001745589543133974,
    'enthusiasm': 0.0008272648556157947,
    'fear': 4.06472872782615e-06,
    'happiness': 1.4103216017247178e-05,
    'neutral': 0.9977772831916809,
    'sadness': 0.0011870539747178555
  },
  'neutral'
)
```

Рисунок 7 - Проверка модели архитектуры *Wav2Vec2*

8. Проверка модели *hubert-emotion-russian-resd*. Программа определила эмоциональный фон как «воодушевленный» с точностью 92%, как показано на рисунке 8.

```
✓ 14 сек. ▶ # @title Recognize: single file - all labels
from rich import print
import torch
from aniemore.recognizers.voice import VoiceRecognizer
from aniemore.models import HuggingFaceModel

model = HuggingFaceModel.Voice.Hubert
device = 'cuda' if torch.cuda.is_available() else 'cpu'

vr = VoiceRecognizer(model, device)
result = vr.recognize(voice_files[0]), vr.recognize(voice_files[0], return_single_label=True)
print(result)
```

⏏ Some weights of the model checkpoint at aniemore/hubert-emotion-russian-resd were not used when
- This IS expected if you are initializing HubertForSequenceClassification from the checkpoint o
- This IS NOT expected if you are initializing HubertForSequenceClassification from the checkpoi
Some weights of HubertForSequenceClassification were not initialized from the model checkpoint a
You should probably TRAIN this model on a down-stream task to be able to use it for predictions
Some weights of the model checkpoint at aniemore/hubert-emotion-russian-resd were not used when
- This IS expected if you are initializing HubertForSequenceClassification from the checkpoint o
- This IS NOT expected if you are initializing HubertForSequenceClassification from the checkpoi
Some weights of HubertForSequenceClassification were not initialized from the model checkpoint a
You should probably TRAIN this model on a down-stream task to be able to use it for predictions
(
 {
 'anger': 1.8258066120324656e-05,
 'disgust': 0.03384198993444443,
 'enthusiasm': 0.9230761528015137,
 'fear': 6.742441473761573e-05,
 'happiness': 7.468734111171216e-05,
 'neutral': 0.041868288069963455,
 'sadness': 0.0010530701838433743
 },
 'enthusiasm'
)

Рисунок 8 - Проверка модели архитектуры *Hubert*

9. Проверка модели *unispeech-sat-emotion-russian-resd*. Программа определила эмоциональный фон как «воодушевленный» с точностью 64%, как показано на рисунке 9.

```
# @title Recognize: single file - all labels
from rich import print
import torch
from aniemore.recognizers.voice import VoiceRecognizer
from aniemore.models import HuggingFaceModel

model = HuggingFaceModel.Voice.UniSpeech
device = 'cuda' if torch.cuda.is_available() else 'cpu'

vr = VoiceRecognizer(model, device)
result = vr.recognize(voice_files[0]), vr.recognize(voice_files[0], return_single_label=True)
print(result)
```

Some weights of the model checkpoint at aniemore/unispeech-sat-emotion-russian-resd were not used
- This IS expected if you are initializing UniSpeechSatForSequenceClassification from the checkpoint
- This IS NOT expected if you are initializing UniSpeechSatForSequenceClassification from the configuration file
Some weights of UniSpeechSatForSequenceClassification were not initialized from the model checkpoint at aniemore/unispeech-sat-emotion-russian-resd
You should probably TRAIN this model on a down-stream task to be able to use it for predictions

Some weights of the model checkpoint at aniemore/unispeech-sat-emotion-russian-resd were not used
- This IS expected if you are initializing UniSpeechSatForSequenceClassification from the checkpoint
- This IS NOT expected if you are initializing UniSpeechSatForSequenceClassification from the configuration file
Some weights of UniSpeechSatForSequenceClassification were not initialized from the model checkpoint at aniemore/unispeech-sat-emotion-russian-resd
You should probably TRAIN this model on a down-stream task to be able to use it for predictions

```
{
  'anger': 0.00022612408793065697,
  'disgust': 0.20026256144046783,
  'enthusiasm': 0.637061595916748,
  'fear': 0.00029982629348523915,
  'happiness': 0.00013136488269083202,
  'neutral': 0.14902694523334503,
  'sadness': 0.012991590425372124
},
'enthusiasm'
```

Рисунок 9 - Проверка модели архитектуры *UniSpeech*

10. Результаты исследований приведены в таблице 1.

Таблица 1 - Сравнение точности

Модель	Заявленная точность	Подтвержденная точность
Голосовые модели		
<u>wav2vec2-xlsr-53-russian-emotion-recognition</u>	73%	99%
<u>wav2vec2-emotion-russian-resd</u>	75%	99%
<u>wavlm-emotion-russian-resd</u>	82%	71%
<u>hubert-emotion-russian-resd</u>	75%	92%
<u>unispeech-sat-emotion-russian-resd Copied</u>	72%	64%
<u>wavlm-bert-base</u>	81%	71%
<u>wavlm-bert-fusion</u>	83%	71%
Текстовые модели		
<u>rubert-base-emotion-russian-cedr-m7</u>	74%	43%
<u>rubert-tiny2-russian-emotion-detection</u>	85%	80%
<u>rubert-large-emotion-russian-cedr-m7</u>	76%	78%
<u>rubert-tiny-emotion-russian-cedr-m7</u>	72%	38%

Вывод: в ходе работы ознакомились с возможностями современных программных средств определения эмоционального фона речи человека в голосе и в тексте, получили навыки работы с открытой библиотекой искусственного интеллекта *Aniemore*, в результате проверки моделей библиотеки определили, что в модуле *Text* наивысшую точность показали модели архитектуры *Bert_Tiny2* (80%) и *Bert_Large* (78%), в модуле *Voice* — *Wav2Vec2* (99%) и *Hubert* (92%).