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Aspect-Based Sentiment Analysis

NLP Homework 2

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Overview

Models (denoted with just letters are transformer-based):

- Given: **Aspect-Based Sentiment Analysis task**

- Aspect Extraction
- Aspect Sentiment
- Aspect Extraction + Aspect Sentiment
- Category Extraction
- Category Evaluation
- Category Extraction + Category Sentiment

- Goal: **Obtain the best-performing model (in terms of macro-F1)**

- **Naive**

- a

- b

- c

- d

- a -> b

- a + b

- c -> d

- c + d

Naive (just word embeddings)

Pre-processing



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Pre-processing

1. First input sentence is tokenized and each token maps to ground truth labels as follows:



1 if token \in gt term
0 otherwise

Before:

'The **selection** changes frequently but
the **basic dishes** are always
available.'

After 1:

```
['The: 0', 'selection: 1', 'changes:  
0', 'frequently: 0', 'but: 0', 'the:  
0', 'basic: 1', 'dishes: 1', 'are:  
0', 'always: 0', 'available: 0']
```

2. Then **indexed vocabulary** (with <UNK> and <PAD> tokens indexed as 1 and 0) and **indexed label vocabulary** (with 3 elements: 2 labels and <PAD> token indexed as 2) are created:

```
Vocab: ['A:2', 'hearty:1', 'two:3', ...,  
'subwoofer:3553', 'scary:3552']
```

```
Label vocab: ['<pad>:2', '0:0', '1:1']
```

After 2:

```
[('The', 36), ('selection', 76), ('changes',  
1), ('frequently', 77), ('but', 78), ('the',  
9), ('basic', 79), ('dishes', 80), ('are',  
81), ('always', 82), ('available', 83),  
('None', 0), ('None', 0), ..., ('None', 0)]
```

```
[0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 2, 2, ..., 2]
```

Embeddings

Two ways of creating indexed vocabulary are applied:

- Based on the dataset;
- Downloading **GloVe 100d** embeddings, so they could be further applied as pre-trained in the network layer.

Naive

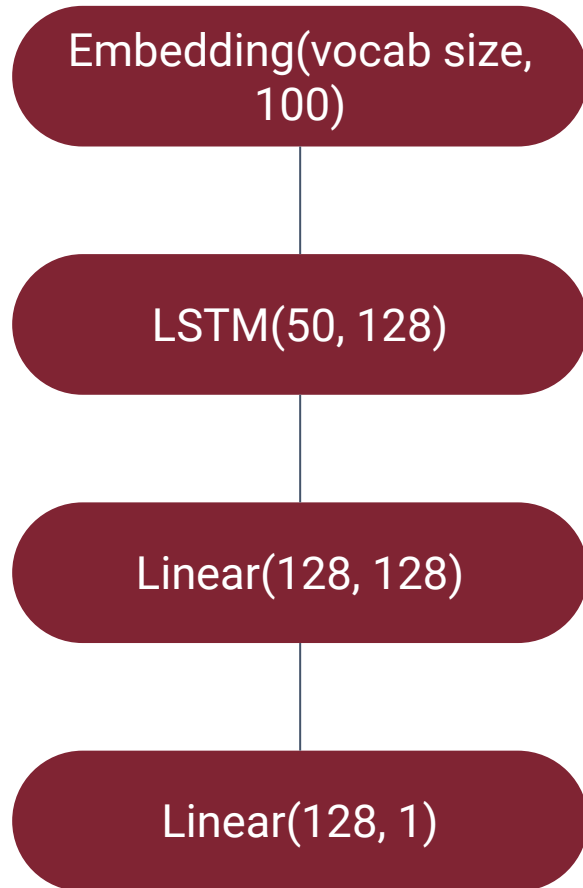
Model



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Model

Architecture



Hyper-parameters

| | |
|-----------------------|--------|
| Epochs | 100 |
| Batch size | 128 |
| Embedding dim | 100 |
| Window Size | 100 |
| Window Shift | 50 |
| N hidden units | 128 |
| N LSTM cells | 1 |
| Optimizer | Adam |
| Learning Rate | 0.0001 |
| Dropout Rate | 0.0 |

Naive

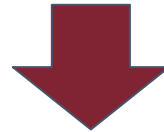
Post-processing and Performance



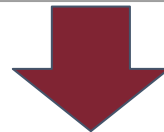
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Post-processing: collect predicted tokens to the multi-token terms

```
{'targets': [[[11, 29], 'wines by the glass', 'negative']],  
  'text': 'Not enough wines by the glass either.'}
```



| Ground truth tokens | Input tokens | Gold labels | Predicted labels | Predicted tokens |
|---------------------|--------------|-------------|------------------|------------------|
| | Not | 0 | 0 | |
| | enough | 0 | 0 | |
| wines | wines | 1 | 1 | wines |
| by | by | 1 | 1 | by |
| the | the | 1 | 0 | |
| glass | glass | 1 | 1 | glass |
| | either | 0 | 0 | |



```
{'wines by', 'glass'}
```

• Best F1: **0.67**

BERT-models

Pre-processing



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- Interpret as a **sentence-pair classification** task:

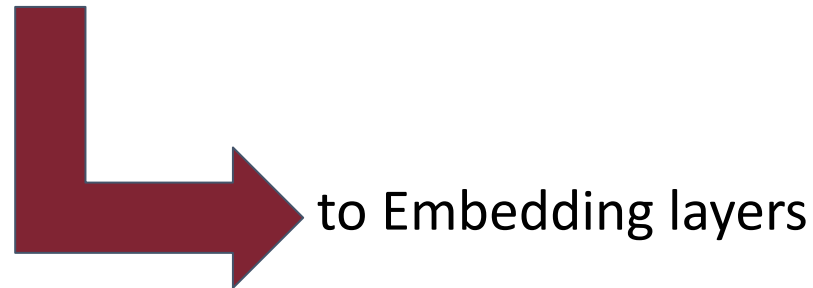

<CLS> Sentence 1 <SEP> Sentence 2 <SEP> Label

- Apply BERT tokenizer from pre-trained for the corresponding BERT model:

[ind(<CLS>),
indices(tokenized(sentence 1)), ind(<SEP>),
indices(tokenized(sentence 2)), ind(<SEP>),
indices(padding)]

+ **attention mask**

[1,
1,...,1, 1,
1,...,1, 1,
0, ...,0]



Given

```
{'categories': [['food', 'neutral']],  
  'targets': [[[4, 13], 'selection', 'neutral'],  
               [[41, 53], 'basic dishes', 'neutral']],  
  'text': 'The selection changes frequently but the basic dishes are always  
available.'}
```

Aspect Extraction

Pre-processing

Binary classification

| Sentence (Sentence 1) | Aspect (Sentence 2) | Label |
|---|---------------------|----------------|
| The selection changes frequently but the basic... | The | unrelated |
| The selection changes frequently but the basic... | selection | related |
| The selection changes frequently but the basic... | changes | unrelated |
| The selection changes frequently but the basic... | frequently | unrelated |
| The selection changes frequently but the basic... | but | unrelated |
| The selection changes frequently but the basic... | the | unrelated |
| The selection changes frequently but the basic... | basic | related |
| The selection changes frequently but the basic... | dishes | related |
| The selection changes frequently but the basic... | are | unrelated |
| The selection changes frequently but the basic... | always | unrelated |
| The selection changes frequently but the basic... | available | unrelated |
| The selection changes frequently but the basic... | . | unrelated |

Given

```
{'categories': [['food', 'neutral']],  
  'targets': [[[4, 13], 'selection', 'neutral'],  
               [[41, 53], 'basic dishes', 'neutral']],  
  'text': 'The selection changes frequently but the basic dishes are always  
available.'}
```

4-class classification
{positive, negative, neutral, conflict}

Aspect Sentiment

| Sentence (Sentence 1) | Aspect (Sentence 2) | Label |
|---|---------------------|---------|
| The selection changes frequently but the basic... | selection | neutral |
| The selection changes frequently but the basic... | basic dishes | neutral |



ground truth terms

Given

```
{'categories': [['food', 'neutral']],  
 'targets': [[[4, 13], 'selection', 'neutral'],  
              [[41, 53], 'basic dishes', 'neutral']],  
 'text': 'The selection changes frequently but the basic dishes are always  
available.'}
```


Category Extraction

| Sentence (Sentence 1) | Category (Sentence 2) | Label |
|---|-------------------------|----------------|
| The selection changes frequently but the basic... | food | related |
| The selection changes frequently but the basic... | service | unrelated |
| The selection changes frequently but the basic... | price | unrelated |
| The selection changes frequently but the basic... | ambience | unrelated |
| The selection changes frequently but the basic... | anecdotes/miscellaneous | unrelated |



fixed 5 for each Sentence 1

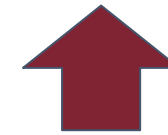
Given

```
{'categories': [['food', 'neutral']],  
 'targets': [[[4, 13], 'selection', 'neutral'],  
              [[41, 53], 'basic dishes', 'neutral']],  
 'text': 'The selection changes frequently but the basic dishes are always  
available.'}
```

4-class classification
{positive, negative, neutral, conflict}

Category Sentiment

| Sentence (Sentence 1) | Category (Sentence 2) | Label |
|---|-----------------------|----------------|
| The selection changes frequently but the basic... | food | neutral |



ground truth categories

Given

```
{ 'categories': [['food', 'neutral']],  
  'targets': [[[4, 13], 'selection', 'neutral'],  
               [[41, 53], 'basic dishes', 'neutral']],  
  'text': 'The selection changes frequently but the basic dishes are always  
available.' }
```

Aspect Extraction + Aspect Sentiment

Pre-processing

5-class classification
{none, positive, negative, neutral, conflict}

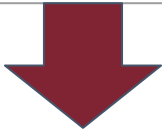
| Sentence (Sentence 1) | Aspect (Sentence 2) | Label |
|---|---------------------|----------------|
| The selection changes frequently but the basic... | The | none |
| The selection changes frequently but the basic... | selection | neutral |
| The selection changes frequently but the basic... | changes | none |
| The selection changes frequently but the basic... | frequently | none |
| The selection changes frequently but the basic... | but | none |
| The selection changes frequently but the basic... | the | none |
| The selection changes frequently but the basic... | basic | neutral |
| The selection changes frequently but the basic... | dishes | neutral |
| The selection changes frequently but the basic... | are | none |
| ... | ... | ... |

Aspect Extraction -> Aspect Sentiment

Pre-processing

Binary classification

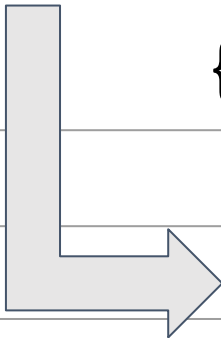
| Sentence (Sentence 1) | Aspect (Sentence 2) | Label |
|---|---------------------|----------------|
| The selection changes frequently but the basic... | The | unrelated |
| The selection changes frequently but the basic... | selection | related |
| ... | ... | ... |



predictions

4-class classification {positive, negative, neutral, conflict}

| Sentence (Sentence 1) | Aspect (Sentence 2) | Label |
|---|---------------------|---------|
| The selection changes frequently but the basic... | predicted terms | neutral |
| ... | ... | ... |



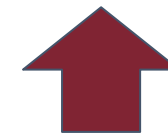
Given

```
{'categories': [['food', 'neutral']],  
  'targets': [[[4, 13], 'selection', 'neutral'],  
               [[41, 53], 'basic dishes', 'neutral']],  
  'text': 'The selection changes frequently but the basic dishes are always  
available.'}
```

Category Extraction + Category Sentiment

5-class classification
{none, positive, negative, neutral, conflict}

| Sentence (Sentence 1) | Category (Sentence 2) | Label |
|---|-------------------------|----------------|
| The selection changes frequently but the basic... | food | neutral |
| The selection changes frequently but the basic... | service | none |
| The selection changes frequently but the basic... | price | none |
| The selection changes frequently but the basic... | ambience | none |
| The selection changes frequently but the basic... | anecdotes/miscellaneous | none |



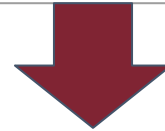
fixed 5 for each Sentence 1

Category Extraction -> Category Sentiment

Pre-processing

Binary classification

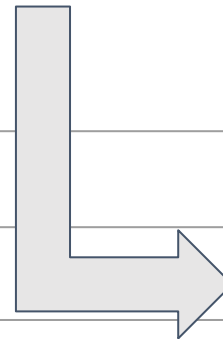
| Sentence (Sentence 1) | Aspect (Sentence 2) | Label |
|---|---------------------|----------------|
| The selection changes frequently but the basic... | food | unrelated |
| The selection changes frequently but the basic... | service | related |
| ... | ... | ... |



predictions

4-class classification
{positive, negative, neutral, conflict}

| Sentence (Sentence 1) | Aspect (Sentence 2) | Label |
|---|----------------------|---------|
| The selection changes frequently but the basic... | predicted categories | neutral |
| ... | ... | ... |



BERT-models

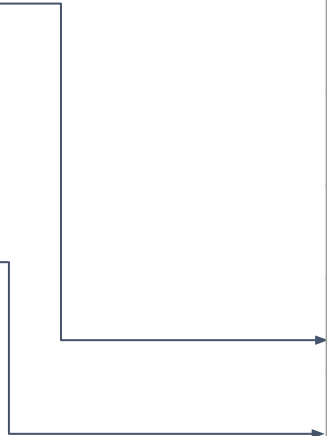
Models



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Models

a
b
a+b
c
d
c+d



| Model | Epochs | Batch Size | Learning Rate |
|-------------|--------|------------|---------------|
| Small BERT | 10 | 24 | 0.00002 |
| BERT | 4 | 16 | 0.00002 |
| Distil BERT | 4 | 24 | 0.00002 |
| Large BERT | 4 | 24 | 0.00002 |

BERT-models

Performances

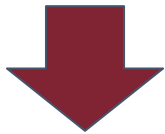


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Performances: Naive vs BERT for Aspect Extraction and Choose the best BERT

- Task: **Aspect Extraction**
- Dataset: **Restaurants**

| Model | Best F1 |
|-------------|-----------|
| Naive | 67 |
| Distil BERT | 83 |



Transformer based architecture is preferable

- Task: **Aspect Extraction**
- Dataset: **Restaurants**

| Model | Best F1 |
|-------------|-----------|
| Small BERT | 79 |
| BERT | 81 |
| Distil BERT | 83 |
| Large BERT | 83 |



Distil BERT is preferable

- Dataset: **Restaurants**

| Task | Distil BERT | Large BERT |
|-------|-------------|------------|
| b | 55 | 60 |
| a + b | 44 | 48 |
| c + d | 48 | 56 |



Large BERT is preferable

Performances: Choose best model for joined and separate tasks, merge datasets

- Dataset: **Restaurants**

| Task | -> | + |
|------|-----------|-----------|
| ab | 48 | 40 |
| cd | 50 | 56 |

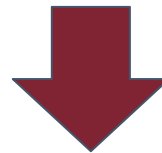


For Aspect Extraction + Aspect Sentiment a -> b performs better than a + b

For Category Extraction + Category Sentiment c + d performs better than c -> d

- Dataset: **Restaurants**

| Task | As is | From Combined |
|------|-----------|---------------|
| a | 83 | 82 |
| b | 60 | 50 |
| c | 82 | 82 |
| d | 64 | 55 |



Separately implemented models perform better than their estimations from a + b and c + d

- Task: **Aspect Sentiment**

| Dataset | Best F1 |
|------------|---------|
| Restaurant | 60 |
| Laptops | 54 |
| Mixed | 59 |

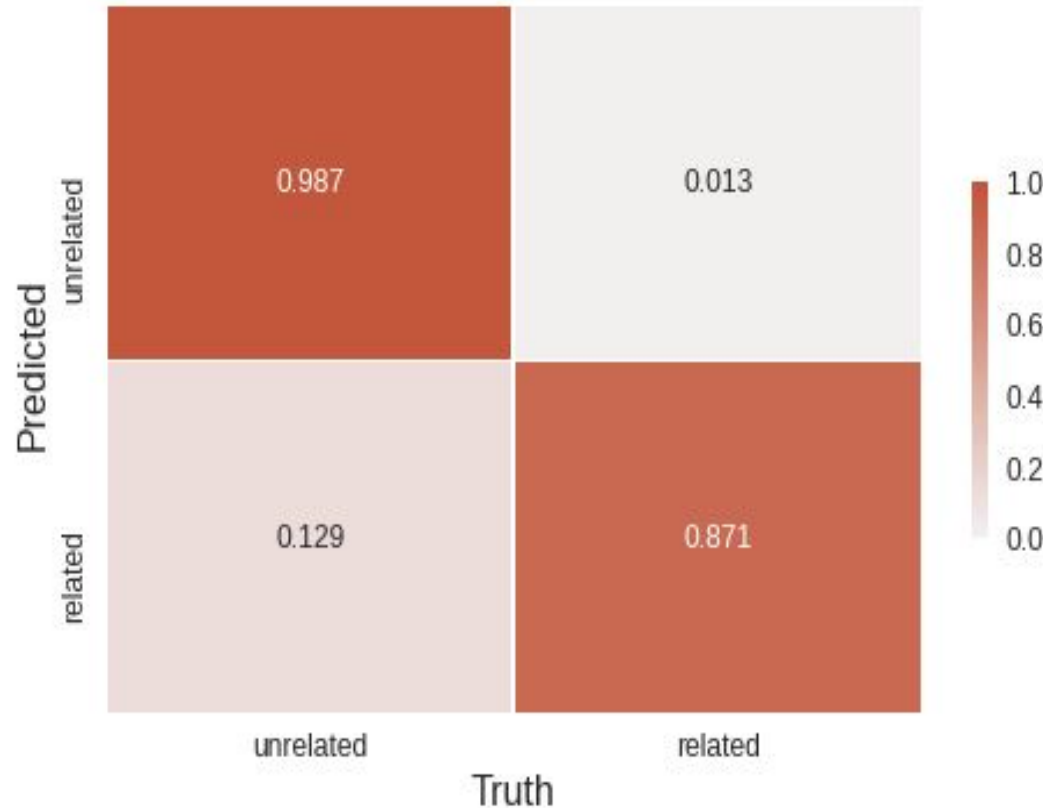


Mixed dataset can be approximated by Restaurants
Laptops are a bit harder than Restaurants

Performances: Confusion matrices for Extractions

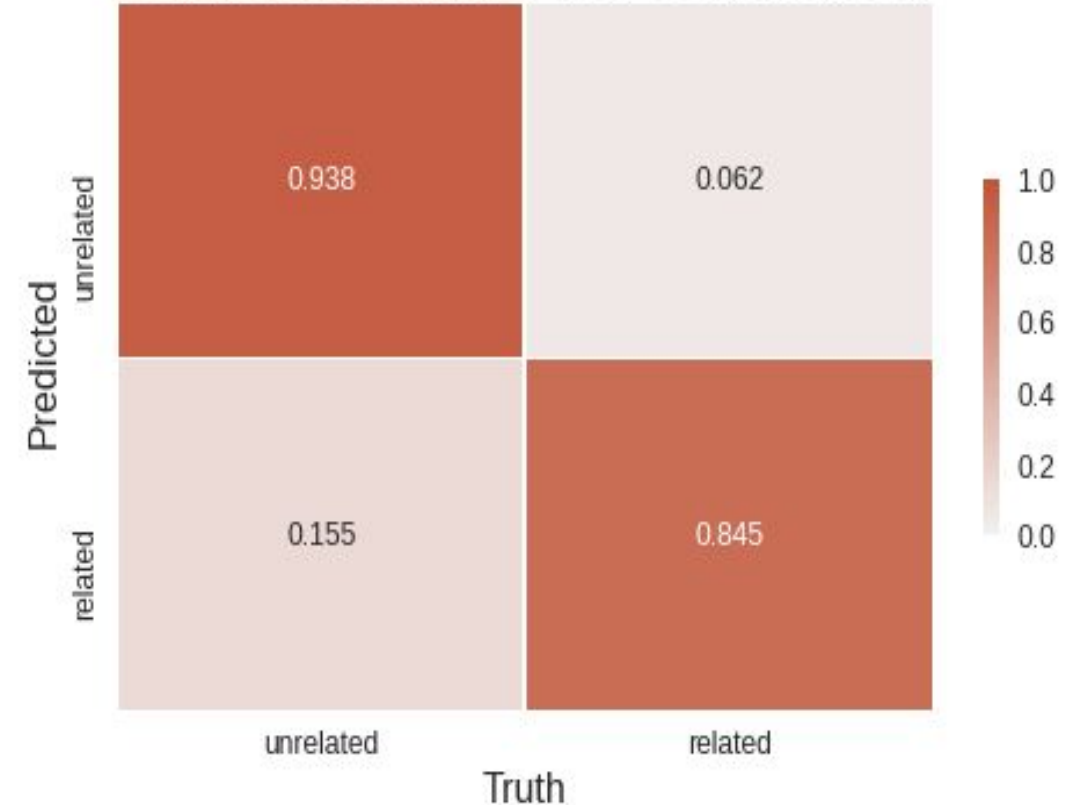
For tokens!

Confusion Matrix for Dev Set



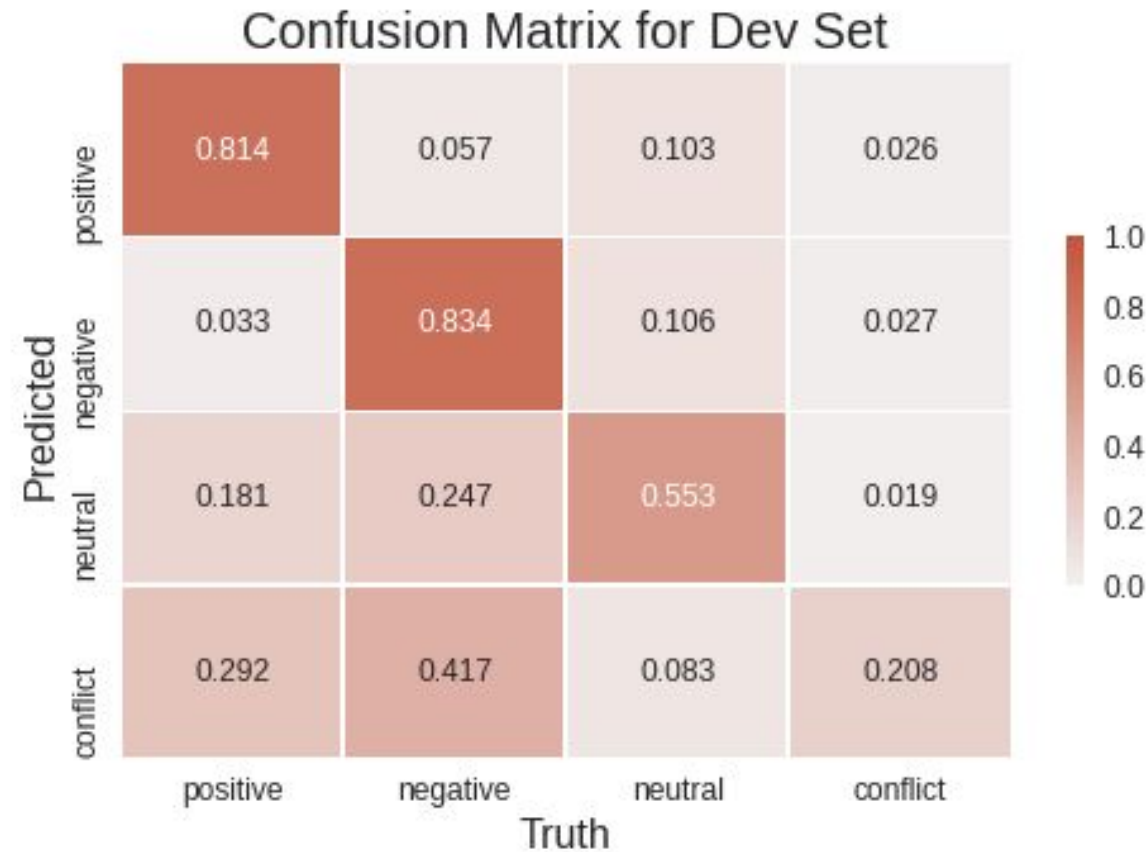
- Task: **Aspect Extraction**
- Dataset: **Mixed**

Confusion Matrix for Test Set (distilbert)

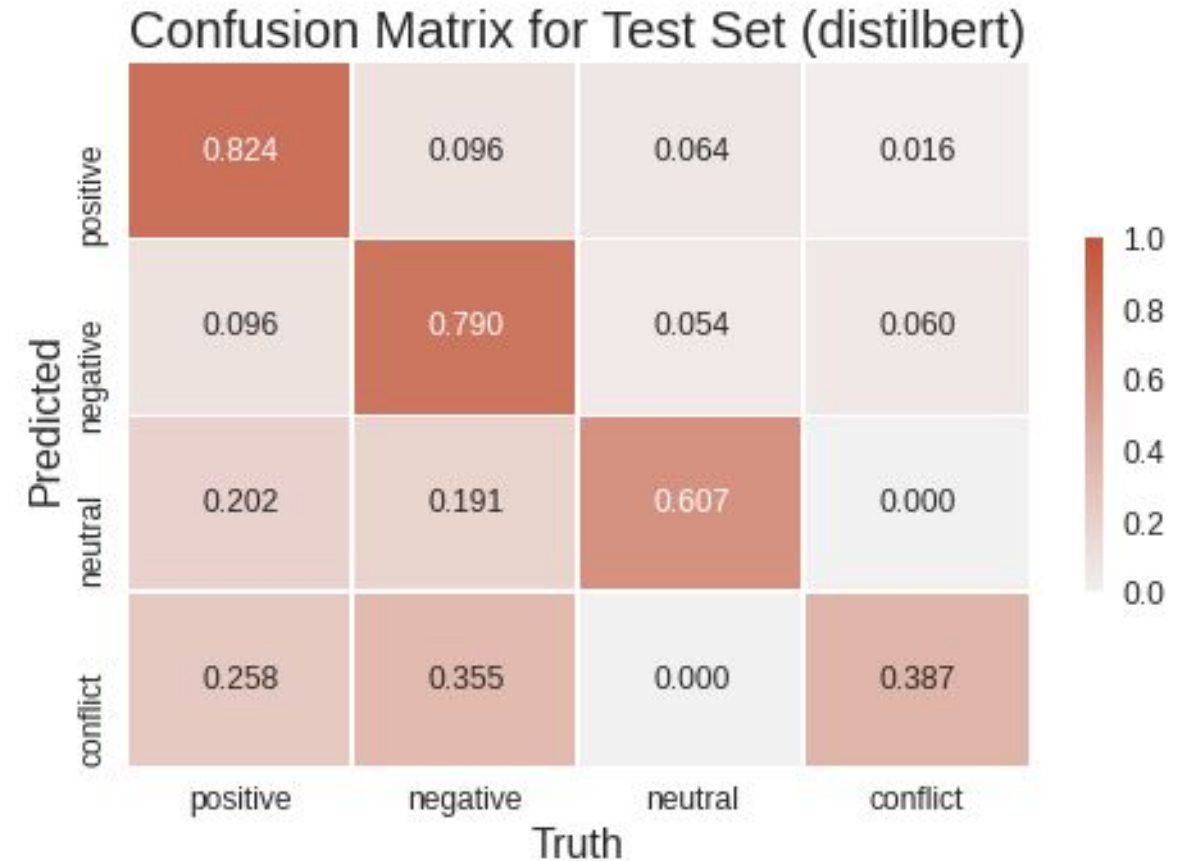


- Task: **Category Extraction**
- Dataset: **Restaurants**

Performances: Confusion matrices for Sentiments



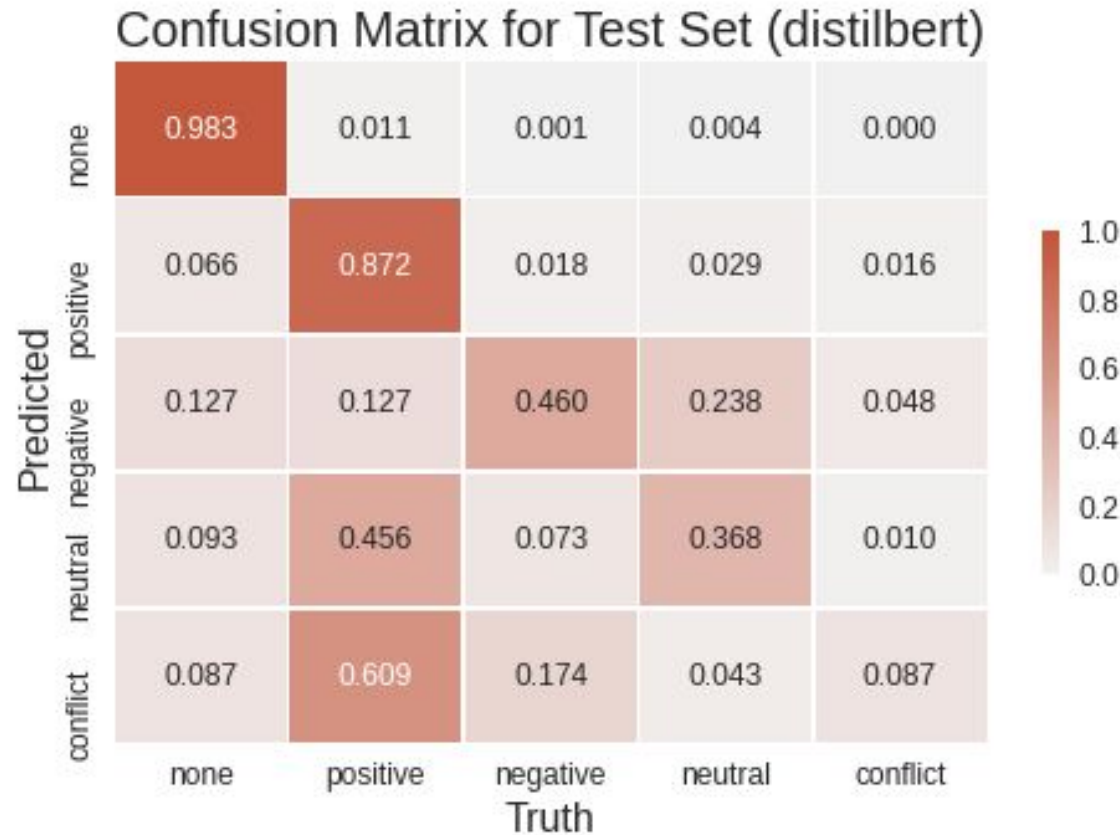
- Task: **Aspect Sentiment**
- Dataset: **Mixed**



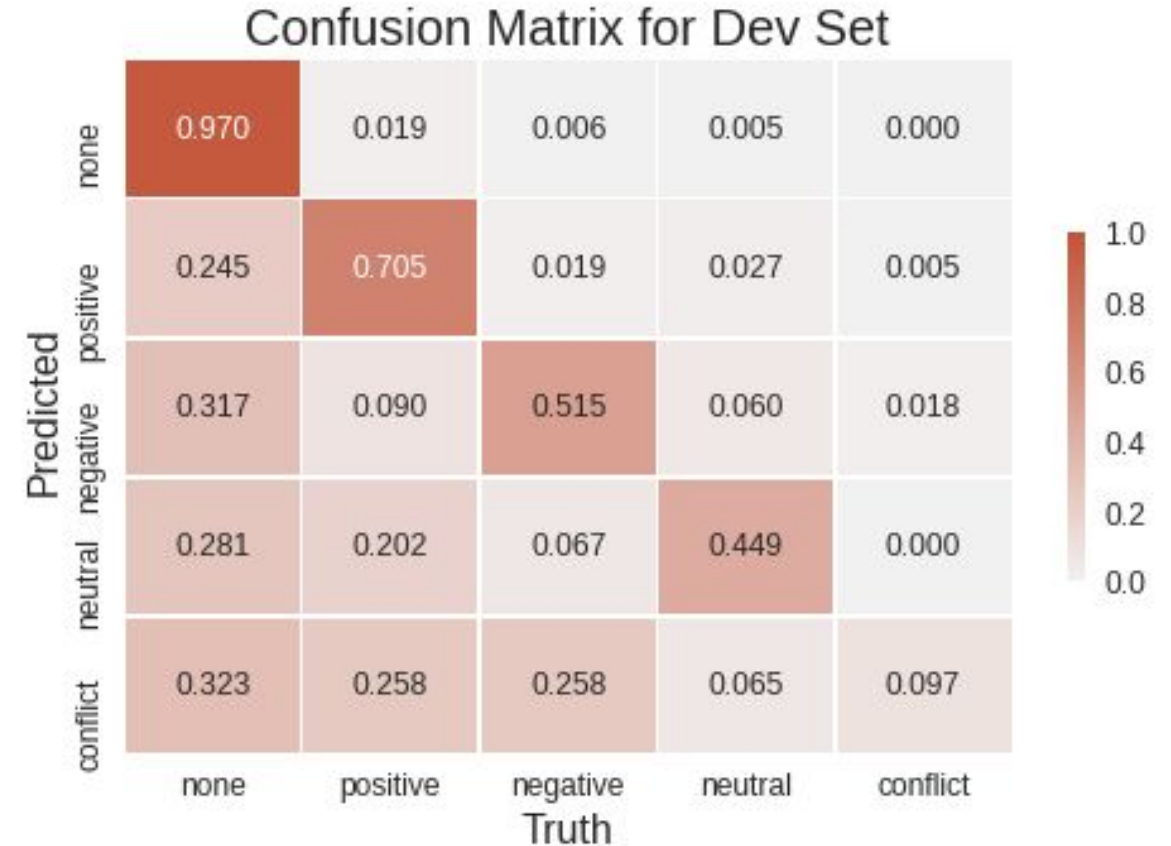
- Task: **Category Sentiment**
- Dataset: **Restaurants**

Performances: Confusion matrices for Combinations

For tokens!



- Task: **a + b**
- Dataset: **Mixed**



- Task: **c + d**
- Dataset: **Restaurants**

Conclusion



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- **Transformers** are powerful instrument in approaching Aspect-Based Sentiment Analysis;
- **Sentence-pair classification** interpretation of a task is appropriate for use of pre-trained BERT models;
- Work on aspects could be improved through the modification of the **decoding** approach;
- Work on categories could be improved through the **conceptual** modifications;
- Precise **fine-tuning** should be under focus;
- For sentiment analysis part **conflict** class is the hardest for recognition.