

Sentiment Analysis

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RoBERTa & other models, tokenization/encoding strategies, stop-word removal

Background 01

Sentiment Analysis

❖ **Sentiment Analysis**

- NLP technique
- Focused on understanding polarity of data
 - how positive or negative (or neutral)
- Often used in business to gauge customer interest / satisfaction

❖ **Natural Language Processing (NLP)**

- The ability of a program to understand a human language (natural language) in a written or spoken context.
- Uses Artificial Intelligence

Dataset

- ❖ Sentiment140
 - 1.6 mil tweets annotated (0 = negative, 4 = positive)
 - Fields
 - target
 - ids
 - date
 - user
 - text
 - 50/50 split of negative & positive sentiment

Methods 02

NLTK Sentiment Vader

- ❖ **Valence Aware Dictionary and sEntiment Reasoner**
- ❖ **Lexicon and rule-based model**
 - Key words are given a value
 - VADER provides estimates based on key words that appear
- ❖ **Downside:**
 - Difficult to upkeep
 - If no keywords appear, VADER will evaluate data as neutral
- ❖ **Process:**
 - Vader is pretrained
 - Using sentiment140 dataset, used Vader to calculate Polarity Scores

Bidirectional Encoder Representations from Transformers (BERT)

❖ Transformer:

- every output element connected to every input element, weightings btwn. them dynamically calculated during training
- self-attention mechanism to **figure out how important all the other “words” in the sentence are** w.r.t. to the aforementioned “word”
 - e.g. “I am hungry.”

❖ BERT:

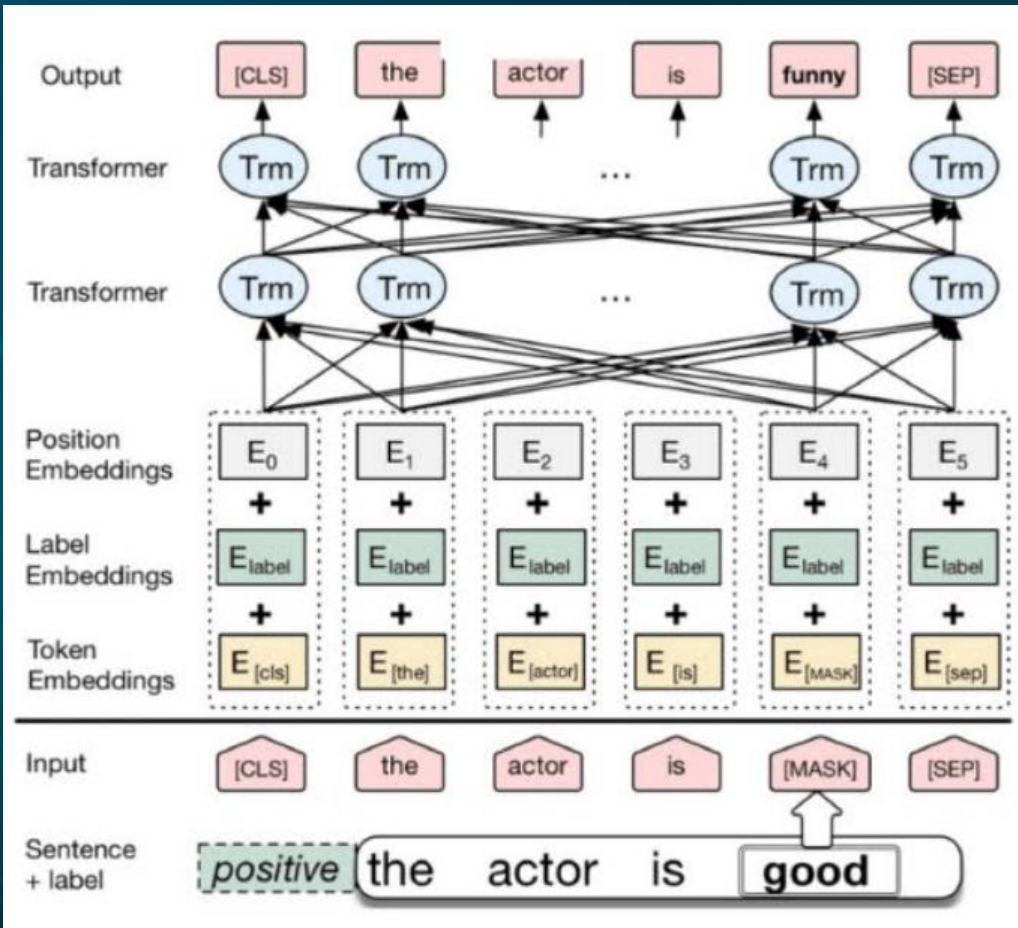
- Pre-trained with text from Wikipedia (Masked Language Modeling & Next Sentence Prediction)
- Can “read both directions” at once

BERT Tokenization

```
>>> from transformers import BertTokenizer

>>> tokenizer = BertTokenizer.from_pretrained("bert-base-uncased")
>>> tokenizer.tokenize("I have a new GPU!")
["i", "have", "a", "new", "gp", "##u", "!"]
```

BERT Architecture



Results & Analysis

03

NLTK Sentiment Vader vs. BERT

❖ Precision:

- Measures exactness
- VADER:
 - Positive: 0.64
 - Negative: 0.78
- BERT:
 - Positive: 0.89
 - Negative: 0.89

❖ Recall:

- Measures completeness
- VADER:
 - Positive: 0.89
 - Negative: 0.44
- BERT:
 - Positive: 0.89
 - Negative: 0.89

NLTK Sentiment Vader vs. BERT

❖ F1 Score:

- Harmonic mean of Precision & Recall
- VADER:
 - Positive: 0.75
 - Negative: 0.56
- BERT:
 - Positive: 0.89
 - Negative: 0.89

❖ Accuracy:

- VADER: 0.68
- BERT: 0.89

Analysis

❖ Evaluation of Performance:

- recall: BERT
- precision: BERT
- f1-score: BERT
- accuracy: BERT

- ❖ Due to its bidirectional transformer encoder architecture and its overall intricate deep learning design (better suited for bigger datasets, i.e. sentiment140), theoretically speaking it should perform better

Conclusion & Future Work

04

Conclusion

❖ Conclusion:

- Bigger datasets should generally employ/utilize deep learning models
- Advancement of NLP time & effort -> BERT, RoBERTa, DistilBERT, XLNet

❖ Future Work:

- Limited due to lack of GPU
- Tokenizers & Models (RoBERTa, other HuggingFace models)
- Stop Word removal (“the”, “by”, etc.)
- Different Hyperparameters (learning rates, epochs)
- Different Optimization & Error/Loss Functions (AdamW & Cross Entropy Loss)

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Q/A