BERTSCORE: EVALUATING TEXT GENERATION WITH BERT

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Background

Commonly used metrics in language generation tasks

- Evaluates based on n-gram overlap, eg. BLEU.
- Fail to capture long dependencies.
- Fails to penalize ordering changes, eg. "Clouds caused rain", "Rain caused clouds".

Proposed Methodology

Given a reference sentence, $x = \langle x_1, ..., x_k \rangle$ and a candidate sentence, $\hat{x} = \langle \hat{x_1}, ..., \hat{x_l} \rangle$

BERTScore computes similarity of two sentences as *sum of cosine similarities* between their *contextualized* token embeddings.

Strengths

• Different words conveying the same thing, would be judged based on their meanings, rather than exact match.

- Not restricted to capturing dependencies upto n-gram length.
- Token level matching allows weighing tokens according to their importance.

Weaknesses

One particular configuration of BERTScore does not work bet for all tasks.

 BERTScore evaluates based on the order of words. It might not be very suitable for tasks where order of words does not matter.

 BERTScore depends on tokens. It might be misleading if the sentences are not tokenized properly.

Possible Improvements

Embeddings could be generated using models other than BERT

 A parameter could be introduced which lets the user control whether BERTScore should consider the ordering of words or not.

 Can come up with a customized tokenization technique especially for tasks to be evaluated using BERTScore

