

ParaAMR: A Large-Scale Syntactically Diverse Paraphrase Dataset by AMR Back-Translation

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Paraphrase Generation

Paraphrase generation benefits many NLP applications

- Question answering
- Chatbots
- Data augmentation
- Robustness

Challenge: Large-Scale High Quality Paraphrase Data

Human-annotated dataset

- MRPC, PAN, Quora
- High quality but limited scale

Automatically generated dataset

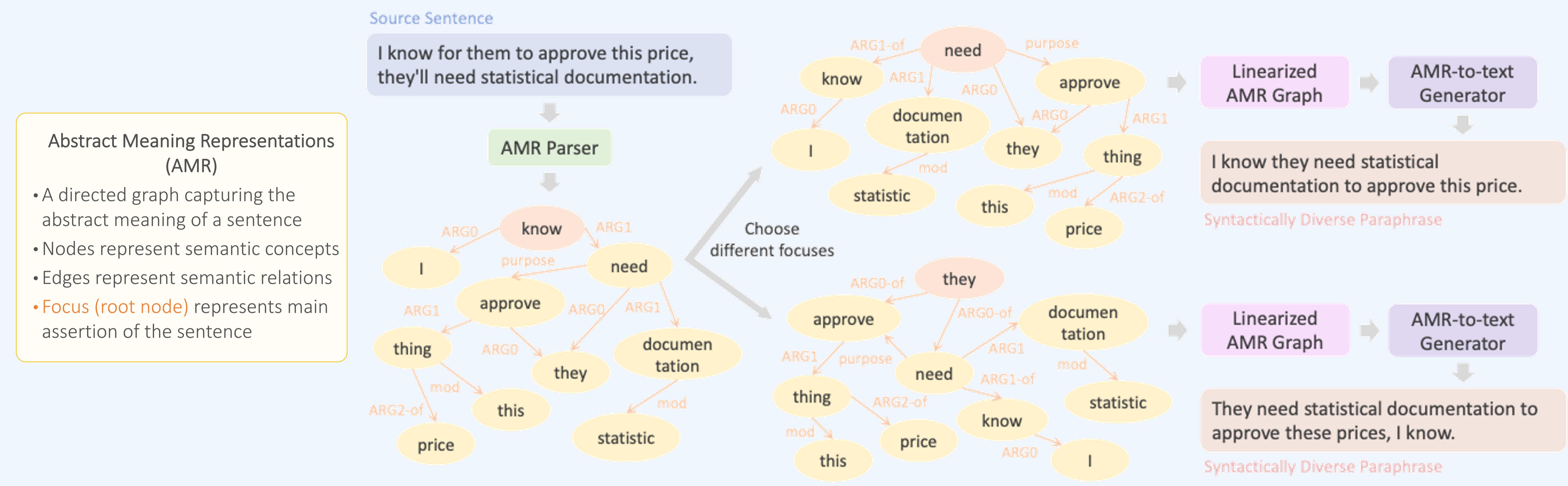
- ParaNMT, ParaBank 1, ParaBank 2 (back-translation)
- Large scale but lack of syntactic diversity

ParaAMR

<https://github.com/uclanlp/ParaAMR>



Generate Syntactically Diverse Paraphrases by AMR Back-Translation



Quantitative Analysis

Automatic Scores

| Dataset | Semantic Similarity (↑) | Lexical Diversity | | Syntactic Diversity | |
|------------------------------------|-------------------------|---------------------|--------------|---------------------|--------------|
| | 1 - BLEU (↑) | 1 - \cap/\cup (↑) | TED-3 (↑) | TED-F (↑) | |
| PARANMT (Wieting and Gimpel, 2018) | 84.28 | 70.71 | 45.78 | 3.28 | 13.94 |
| PARABANK1 (Hu et al., 2019a) | 81.77 | 78.19 | 52.59 | 3.59 | 14.53 |
| PARABANK2 (Hu et al., 2019b) | 82.50 | 88.82 | 59.61 | 4.04 | 17.41 |
| PARAAMR (Ours) | 82.05 | 87.86 | 53.10 | 5.86 | 22.07 |

Human Evaluation Scores

| Datasets | Semantic Similarity | | | | Syntactic Diversity | | | |
|------------------------------------|---------------------|------|------|-------------|---------------------|------|------|-------------|
| | 3(%) | 2(%) | 1(%) | Average | 3(%) | 2(%) | 1(%) | Average |
| PARANMT (Wieting and Gimpel, 2018) | 28.7 | 46.7 | 24.6 | 2.04 | 16.7 | 45.0 | 38.3 | 1.78 |
| PARABANK1 (Hu et al., 2019a) | 26.8 | 49.0 | 24.2 | 2.03 | 15.1 | 47.8 | 37.1 | 1.78 |
| PARABANK2 (Hu et al., 2019b) | 26.8 | 50.3 | 22.9 | 2.04 | 14.2 | 51.8 | 34.0 | 1.80 |
| PARAAMR (Ours) | 26.5 | 47.2 | 26.3 | 2.00 | 18.2 | 53.8 | 28.0 | 1.90 |

Applications

Learning Sentence Embeddings

I like you very much. semantically similar
I love you very much. high text similarity
I like you very much. semantically different
I don't like you. low text similarity

| Dataset | Pearson's r | Spearman's r |
|----------------|-------------------------|-------------------------|
| PARANMT | 74.38 \pm 0.70 | 73.80 \pm 0.42 |
| PARABANK1 | 74.80 \pm 1.33 | 74.56 \pm 1.02 |
| PARABANK2 | 75.39 \pm 0.29 | 75.17 \pm 0.25 |
| PARAAMR (ours) | 77.70 \pm 0.40 | 75.72 \pm 0.43 |

Syntactically Controlled Paraphrase Generation

This is a good restaurant. The restaurant is good.

| Dataset | Quora | MRPC | PAN |
|----------------|-------------------------|-------------------------|-------------------------|
| PARANMT | 47.38 \pm 0.39 | 45.24 \pm 0.61 | 39.45 \pm 0.50 |
| PARABANK1 | 46.21 \pm 0.26 | 44.52 \pm 0.18 | 39.85 \pm 0.11 |
| PARABANK2 | 46.86 \pm 0.45 | 45.17 \pm 0.39 | 40.20 \pm 0.56 |
| PARAAMR (ours) | 48.50 \pm 0.11 | 47.38 \pm 0.19 | 40.30 \pm 0.10 |

Data Augmentation for Few-Shot Learning

| Dataset | MRPC | QQP | RTE |
|------------------|------------------|--------------|--------------|
| | 15-Shot Learning | | |
| 15-Shot Baseline | 59.93 | 63.18 | 54.05 |
| PARANMT | 49.26 | 63.54 | 55.68 |
| PARABANK1 | 59.56 | 63.72 | 54.59 |
| PARABANK2 | 58.46 | 63.54 | 54.05 |
| PARAAMR (ours) | 62.87 | 64.08 | 52.97 |

| 30-Shot Learning | | | |
|------------------|--------------|--------------|--------------|
| 30-Shot Baseline | MRPC | QQP | RTE |
| PARANMT | 68.38 | 64.93 | 54.51 |
| PARABANK1 | 67.65 | 66.20 | 52.71 |
| PARABANK2 | 64.46 | 64.86 | 53.79 |
| PARAAMR (ours) | 68.38 | 64.91 | 54.15 |
| PARAAMR (ours) | 69.36 | 67.03 | 55.60 |