

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

Abstract Meaning Representations

(AMR)

A directed graph capturing the

abstract meaning of a sentence

Nodes represent semantic concepts

Edges represent semantic relations

Focus (root node) represents main

assertion of the sentence

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 D 1 - BLEU (†)	•	Syntactic TED-3 (†)	Diversity 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

Datasata	Semantic Similarity		Syntactic Diversity					
Datasets	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

Qualitative Examples

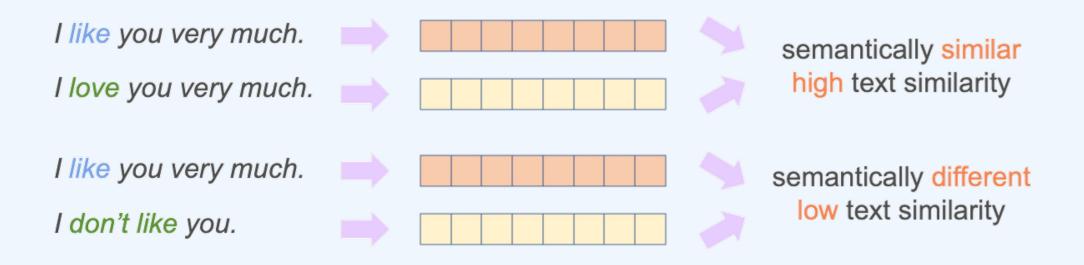
Source Sentence	I know for them to approve this price, they'll need statistical documentation.
PARANMT	I know that in order to accept this award, they'll need a statistical analysis.
PARABANK1	I know that to accept this prize, they're going to need statistical analysis. I know that in order to accept this prize, they're going to need a statistic analysis. I know that if they accept this prize, they're gonna need a statistical analysis.
PARABANK2	I know that to accept that prize, they're going to need a statistical analysis. I know that in order to accept this prize, they will require a statistical analysis. I know they'll require statistical analysis to accept that prize.
PARAAMR	I know they need statistical documentation to approve this price. There is statistic documentation I know they need to approve these prices. They need statistical documentation to approve these prices, I know.

PARABANK2

PARAAMR (ours)

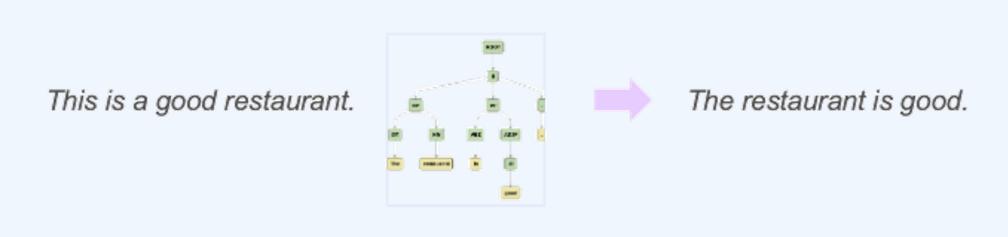
Applications

Learning Sentence Embeddings



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

Syntactically Controlled Paraphrase Generation



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

Data Augmentation for Few-Shot Learning

Dataset	MRPC	QQP	RTE
15-S	hot Learnir	ng	
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-S	hot Learnir	ng	
30-Shot Baseline	68.38	64.93	54.51
PARANMT	67.65	66.20	52.71
PARABANK1	64.46	64.86	53.79

68.38

69.36

64.91

67.03

54.15

55.60