





Neural AMR: Sequence-to-Sequence Models for Parsing and Generation



joint work with Srinivasan Iyer, Mark Yatskar, Yejin Choi, Luke Zettlemoyer











- Rooted Directed Acyclic Graph
- Nodes: concepts (nouns, verbs, named entities, etc)
- Edges: Semantic Role Labels

I have **known** a **planet** that was **inhabited** by a **lazy man**.





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Input: AMR Graph





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• Text Summarization (Liu et al., 2015)



Machine Translation (Jones et al., 2012)



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Existing Approaches

Generate from AMR

- MT-based
 - Flanigan et al. 2016, Pourdamaghani and Knight 2016, Song et al. 2016

Grammar-based

Lampouras and Vlachos 2017, Mille et al. 2017

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Parse to AMR

- Alignment-based
 - Flanigan et al. 2014, 2017 (JAMR)

Grammar-based

 Wang et al. 2016 (CAMR), Pust et al. 2015, Artzi et al. 2015, Damonte et al. 2017, Goodman et al. 2016, Puzikov et al. 2016, Brandt et al. 2017, Nguyen et al. 2017

Neural-based

 Barzdins and Gosko 2016, Peng et al. 2017, Noord and Bos 2017, Buys and Blunsom 2017

Overview

- Sequence-to-sequence architecture
 - End-to-end model w/o intermediate representations
 - Linearisation of AMR graph to string
 - Pre-process

- Paired Training
 - Scalable data augmentation algorithm









Graph —> Depth First Search (Human-authored annotation)



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Pre-processing

Linearization —> Anonymization



Pre-processing

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US officials held an expert group meeting in January 2002 in New York .

Pre-processing

Linearization —> Anonymization



US officials held an expert group meeting in January 2002 in New York .

loc_0 officials held an expert group meeting in month_0 year_0 in loc_1 .

Experimental Setup

AMR LDC2015E86 (SemEval-2016 Task 8)

- Hand annotated MR graphs: newswire, forums
- ~16k training / 1k development / 1k test pairs

Train

Optimize cross-entropy loss



Evaluation

- BLEU n-gram precision (Generation) (Papineni et al., 2002)
- SMATCH score (Parsing) (Cai and Knight, 2013)

Experiments

- Vanilla experiment
 - Limited Language Model Capacity
- Paired Training
 - Data augmentation algorithm

TreeToStr: Flanigan et al, NAACL 2016 **TSP**: Song et al, EMNLP 2016 **PBMT**: Pourdamaghani and Knight, INLG 2016



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(Sennrich et al., ACL 2016)

Reference

US officials held an expert group meeting in January 2002 in New York .

Prediction

United States officials held held a meeting in January 2002.

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Prediction

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Repetition

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US officials held an expert group meeting in January 2002 in New York .

Prediction

United States officials held held a meeting in January 2002.

- Repetition
- Coverage



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Original Dataset: ~16k graph-sentence **pairs**



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Gigaword: ~183M sentences *only*



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Sample sentences with vocabulary overlap











Semi-supervised Learning

Self-training

McClosky et al. 2006

Co-training

- Yarowski 1995, Blum and Mitchell 1998, Sarkar 2001
- Sogaard and Rishoj, 2010

Train AMR Parser P on Original Dataset

AMR (🗐 , 🐾)

Train AMR Parser P on Original Dataset



for i = 0 ... N

Train AMR Parser **P** on Original Dataset



S_i = Sample k 10ⁱ sentences from Gigaword

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AMR (🗐 , 🐾)

Train AMR Parser **P** on Original Dataset



for i = 0 ... N

S_i = Sample k 10ⁱ sentences from Gigaword

Parse S_i sentences with P



Train AMR Parser **P** on Original Dataset





Train AMR Parser **P** on Original Dataset





Train AMR Parser **P** on Original Dataset





Train Generator G on SN

Train **P** on Original Dataset



Train **P** on Original Dataset

















Fine-tune: init parameters from previous step and train on Original Dataset



Fine-tune: init parameters from previous step and train on Original Dataset










Training AMR Generator



Training AMR Generator



Training AMR Generator

















Final Results (Parsing)









How did we do? (Generation)

Reference

US officials held an expert group meeting in January 2002 in New York .

Prediction

In January 2002 United States officials held a meeting of the group experts in New York .

Errors: Disfluency Coverage

How did we do? (Generation)

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In January 2002 United States officials held a meeting of the group experts in New York .

Reference

The report stated **British government** must help to stabilize **weak states** and push for international regulations that would stop **terrorists** using freely available information to create and unleash new forms of biological warfare such as **a modified** version of the influenza **virus**.

Prediction

The report stated that the **Britain government** must help stabilize **the weak states** and push international regulations to stop the use of freely available information to create a form of new biological warfare such as **the modified** version of the influenza.

Errors: Disfluency Coverage

Summary

- Sequence-to-sequence models for Parsing and Generation
- Paired Training: scalable data augmentation algorithm
- Achieve state-of-the-art performance on generating from AMR
- Best-performing Neural AMR Parser
- Demo, Code and Pre-trained Models: <u>http://ikonstas.net</u>



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Bonus Slides





Linearize —> RNN encoding

- Token embeddings



- Token embeddings
- Recurrent Neural Network (RNN)



 \mathbf{h}_5

ARG0-of

- Token embeddings
- Recurrent Neural Network (RNN)
- Bi-directional RNN



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 $h_{5}(s)$

ARG0-of

- Token embeddings
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RNN Encoding —> RNN Decoding (Beam search)



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- init $\mathbf{h}^{(s)}$



RNN Encoding —> RNN Decoding (Beam search)

- init $\mathbf{h}^{(s)}$
- softmax



Holding

Held

US

RNN Encoding —> RNN Decoding (Beam search)

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RNN Encoding —> RNN Decoding (Beam search)



RNN Encoding —> RNN Decoding (Beam search)















