

Multilingual Semantic Networks for Data-driven Interlingua Seq2Seq Systems

Cristina España-Bonet and Josef van Genabith
[UdS & DFKI]

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12th May 2018

Goal1 Improve MT specially for low-resourced languages

Goal2 (future) Translation with monolingual data
and/or parallel from other pairs

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Framework (Multilingual) NMT

Approach Interlinguality

Resources Multilingual Knowledge Graphs

Outline

1 Introduction

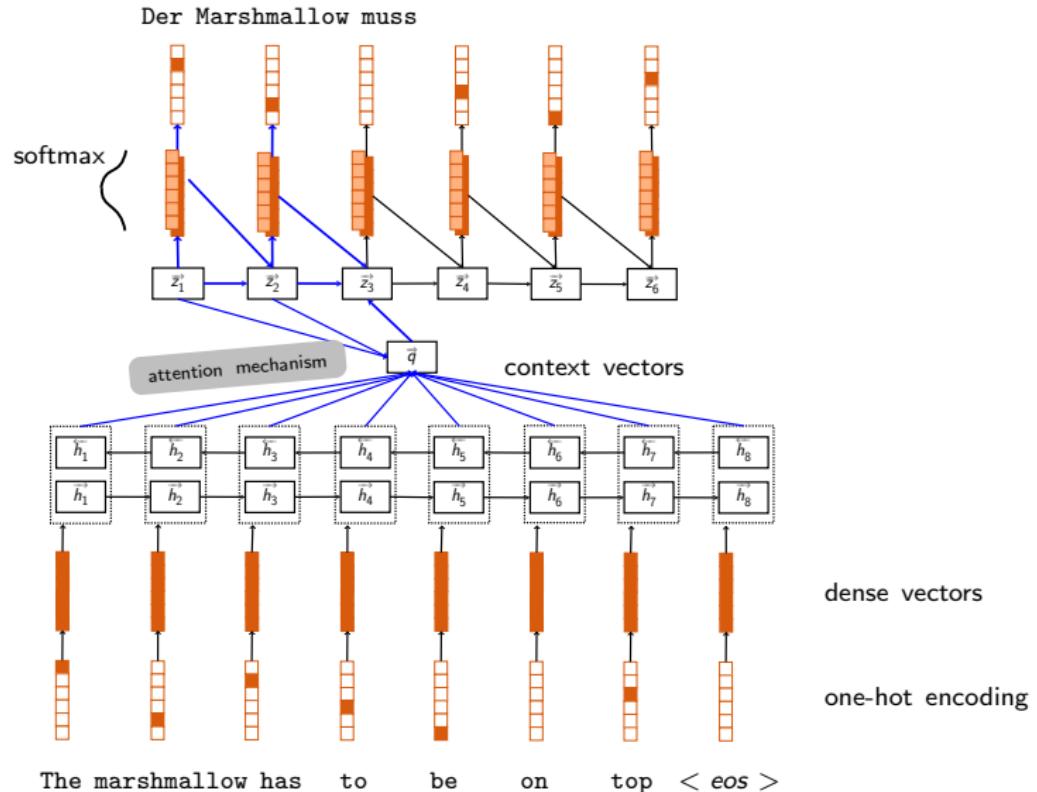
2 Approach

3 Current Results

4 Conclusions

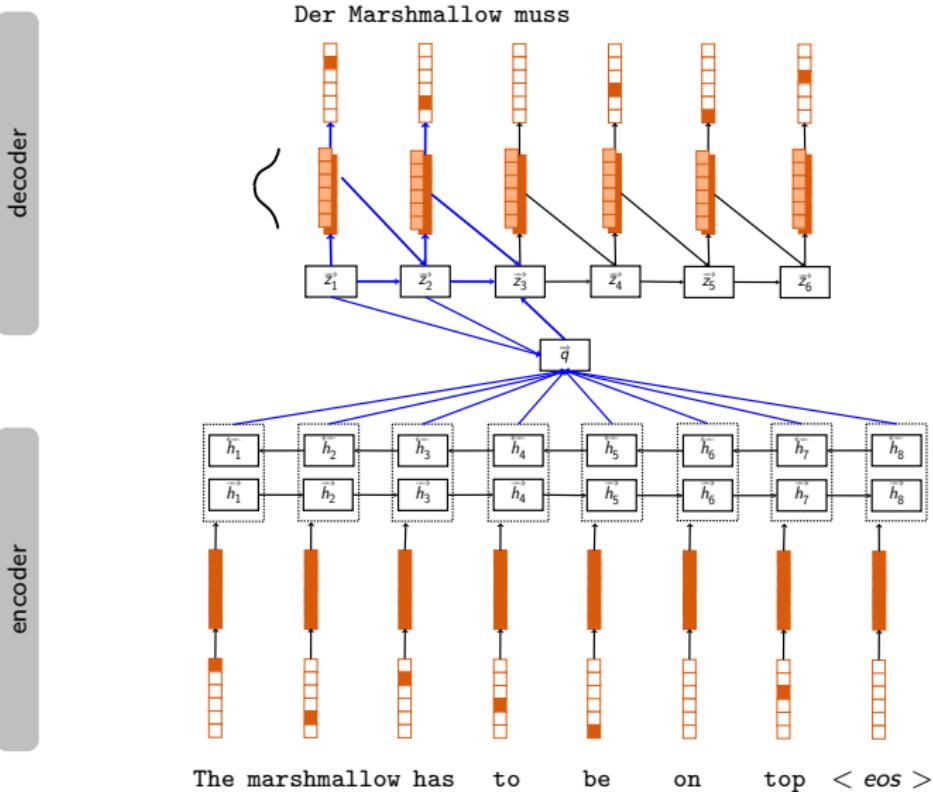
Introduction

(Multilingual) NMT



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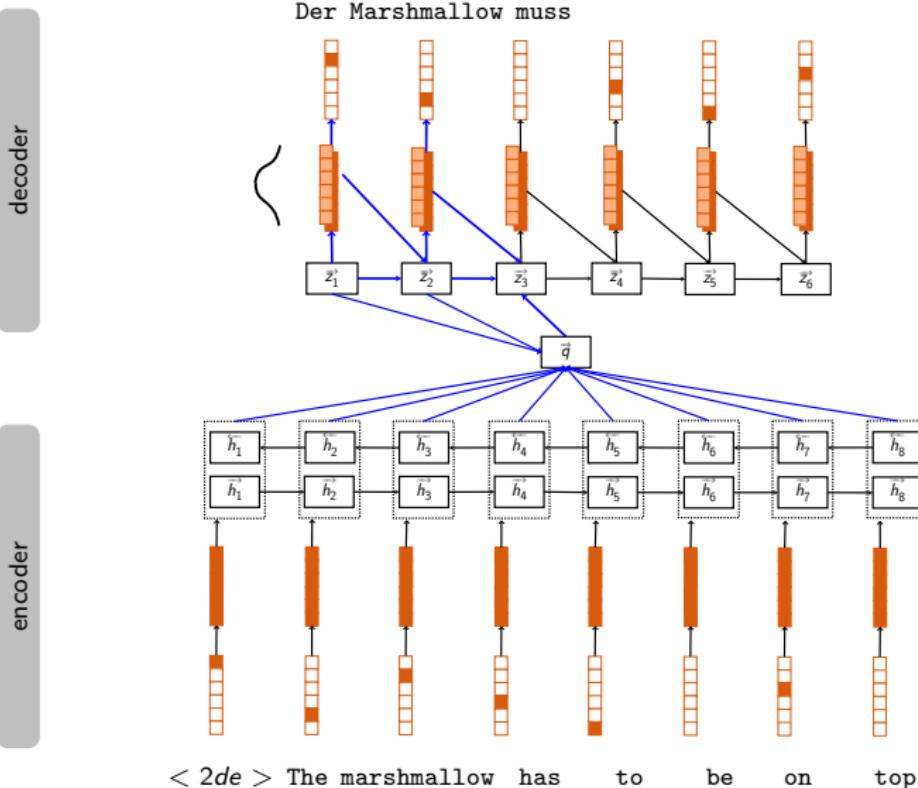


Vocabulary
Marshmallow,
muss, oben,
drauf, sein...

Vocabulary
marshmallow,
has, to,
be, top...

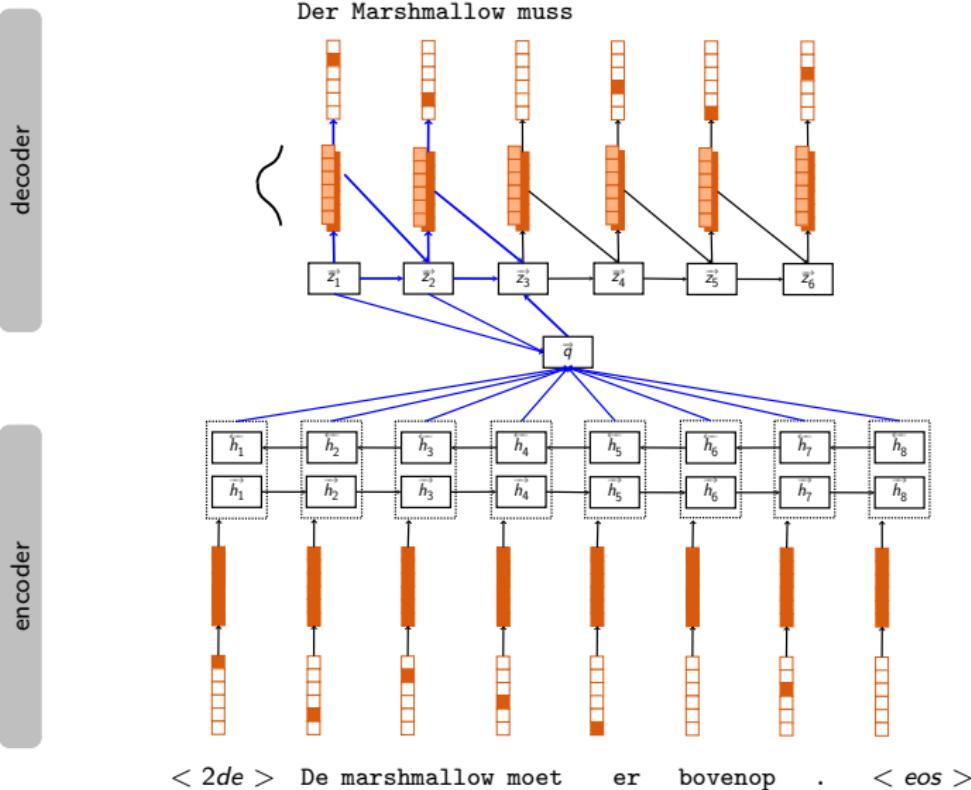
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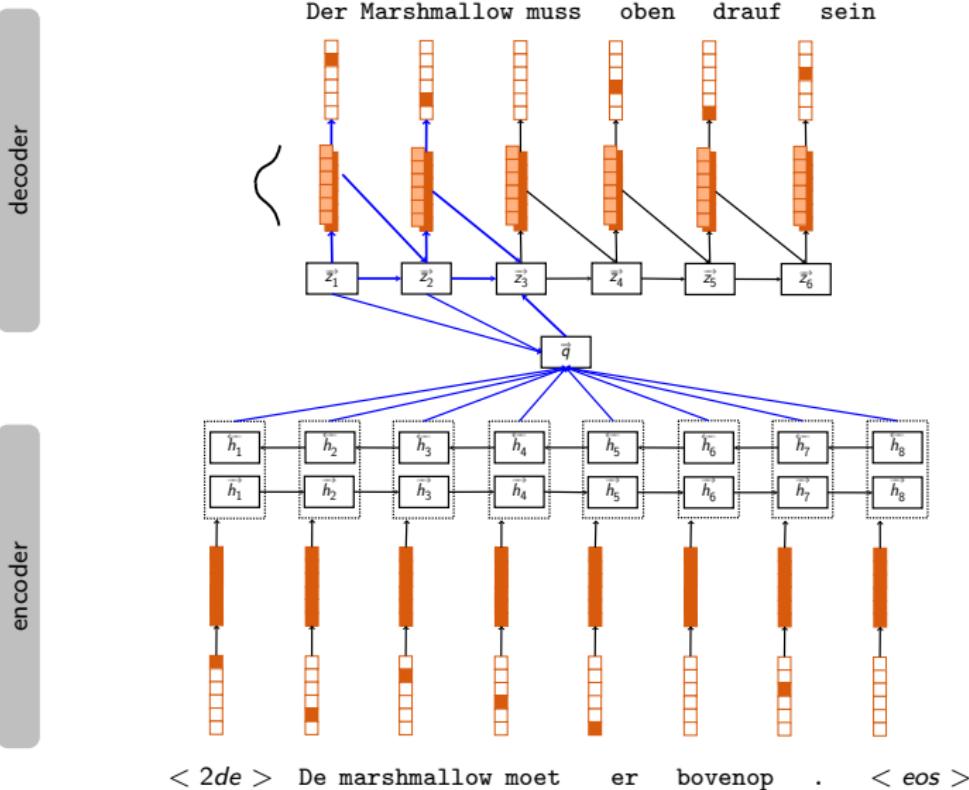
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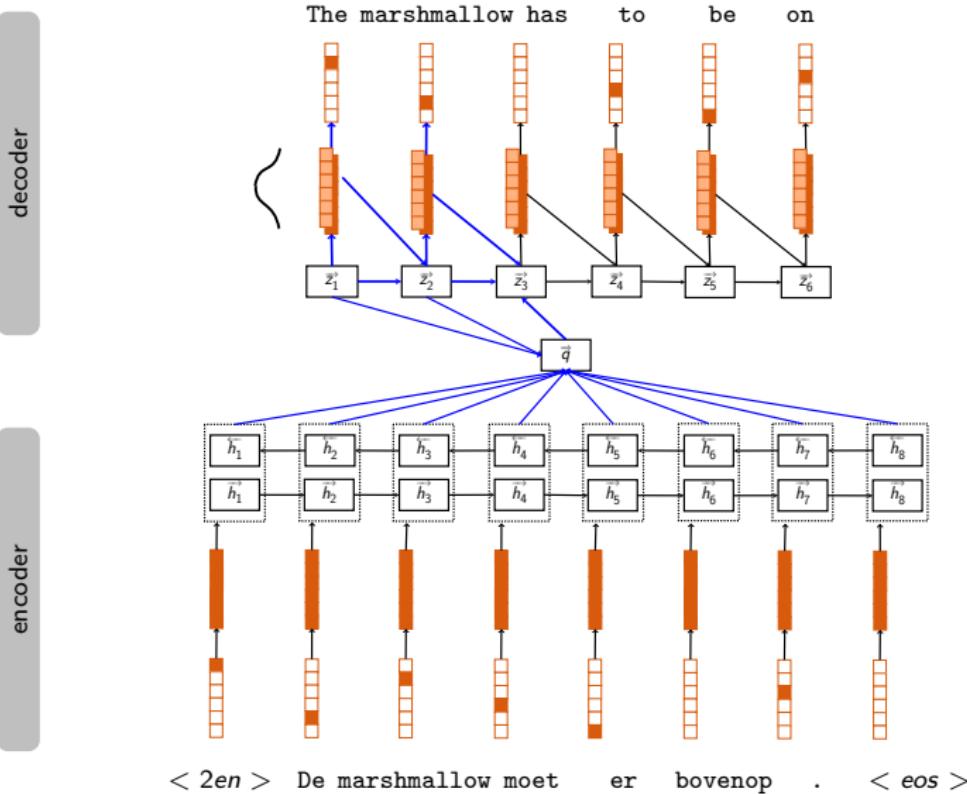
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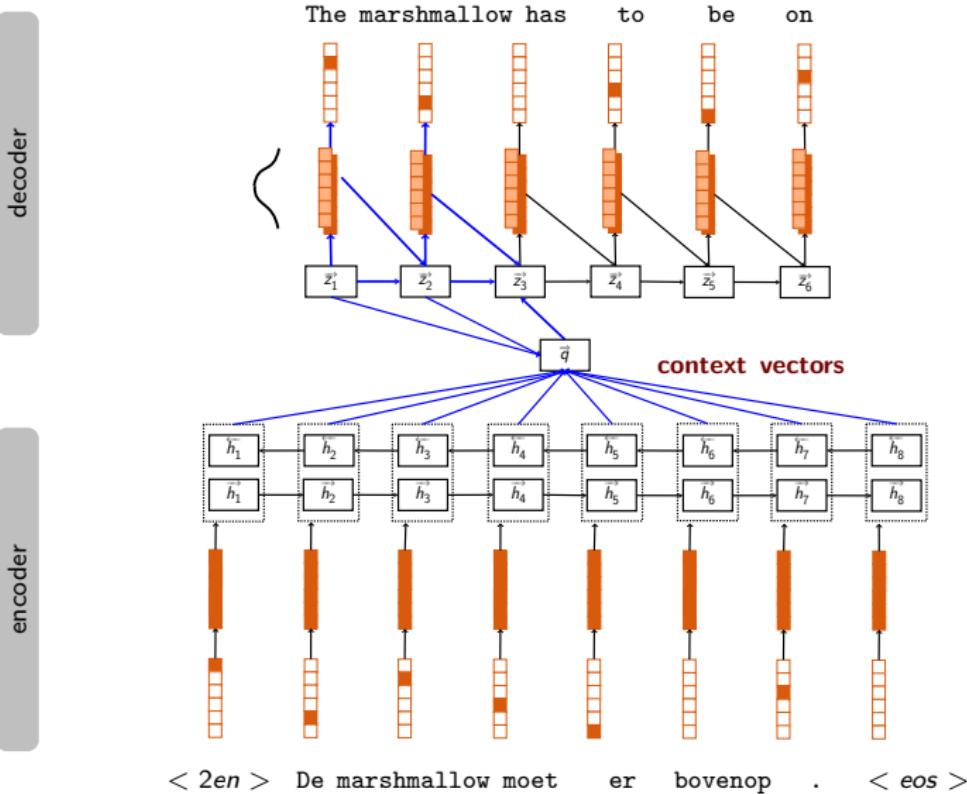


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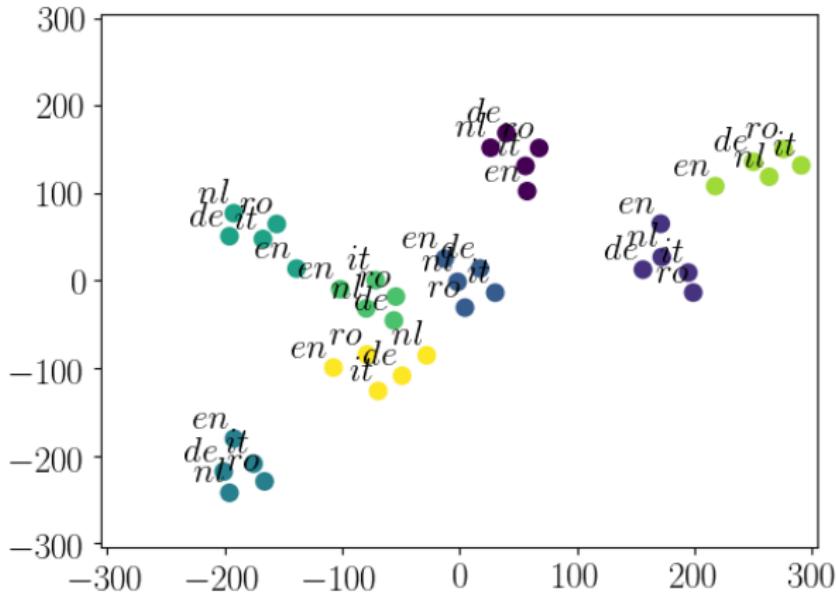
Introduction

(Multilingual) NMT



Introduction

Interlingual Semantic Space



ML-NMT $\{de, en, nl, it, ro\} \rightarrow \{de, en, nl, it, ro\}$ with TED talks

Introduction

Purposes of MSNs in NMT

- Help to cluster by semantics
 - Improve general translation
- Reduce UNKs with help of other languages
 - Help translation of under-resourced languages
- Translation from unseen languages
 - Beyond-zero-shot translation
- (semi-)Abstract representation for sentences
 - Data-driven “Interlingual” translation

Approach

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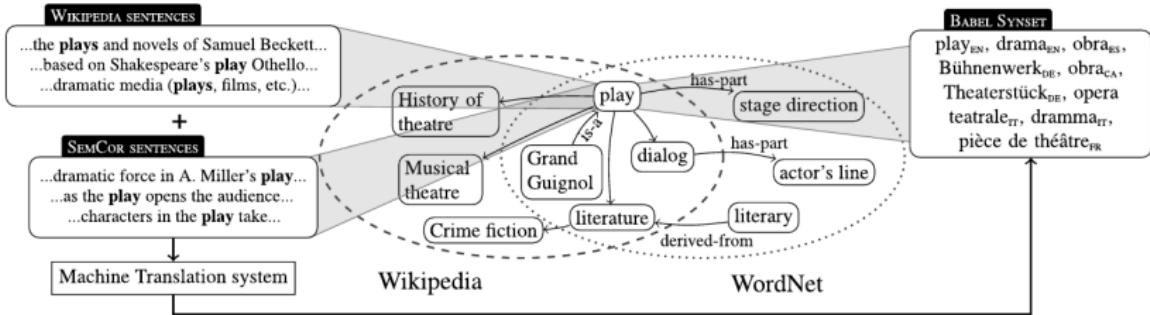
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Approach

Multilingual Semantic Network: *BabelNet*

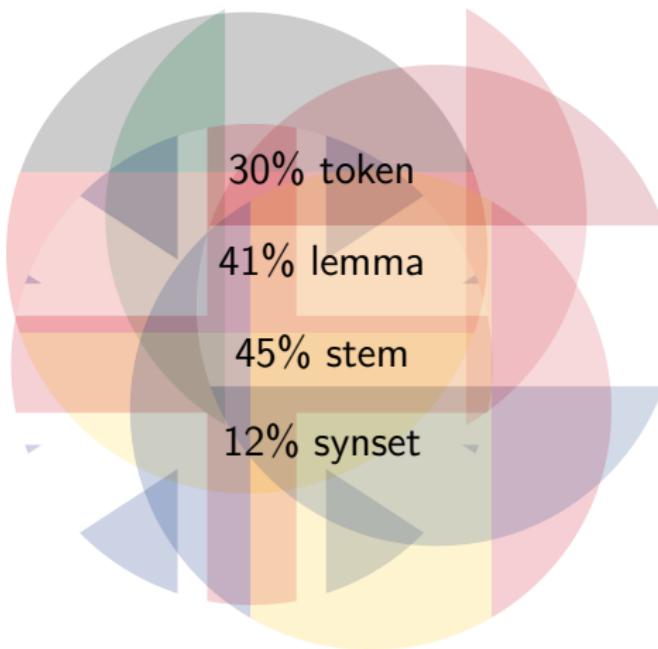


[Navigli & Ponzetto, 2012]

- 1 Synsets as cross-lingual lexicons (for 271 languages!)
- 2 Relations among synsets

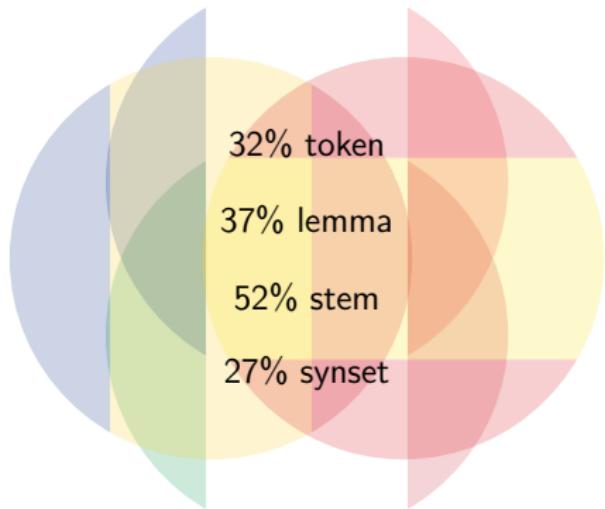
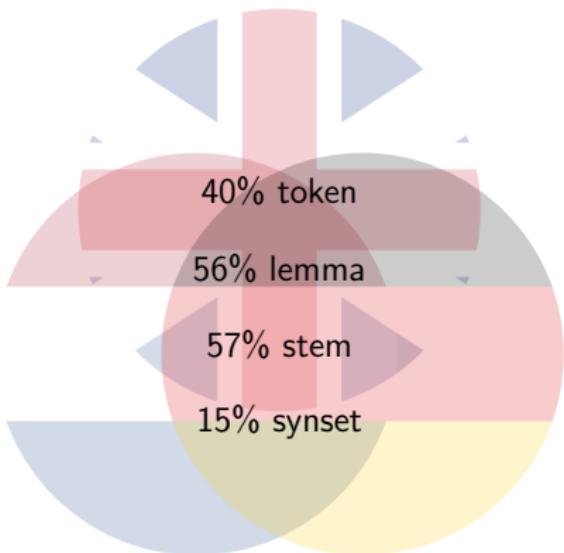
Approach

Multilingual Corpus: en-de-nl-ro-it TED



Approach

Multilingual Corpus: TED, Germanic & Latin Languages



Approach

Simple Approaches: Source Sentence Encodings

■ word2word translation w:

<2en> es war ein riesiger Erfolg

<2en> è stato un enorme successo

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Simple Approaches: Source Sentence Encodings

- **word2word translation** *w:*

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- **synset2word translation** *b:*

PRONOUN VERB DETERMINER ADJECTIVE bn:15350982n

bn:00083181v bn:00083181v DETERMINER bn:00102268a

bn:00078365n

Approach

Simple Approaches: Synset Annotation

- Enrich content words (nouns —including NEs, foreign words and numerals—, adjectives, adverbs and verbs) with synsets
- Retrieve a synset according to the lemma and PoS of a word
- Select the BabelNet ID according to BabelNet's ordering
- Mark negation particles with a tag NEG
- Keep (coarse) PoS for the remaining tokens

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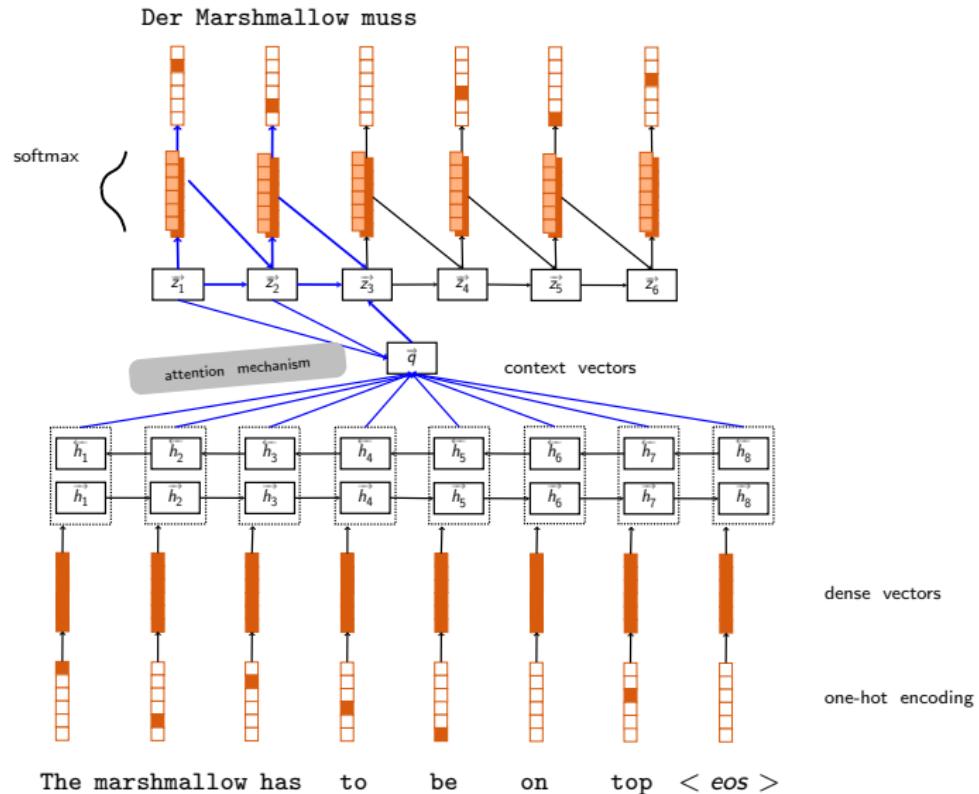
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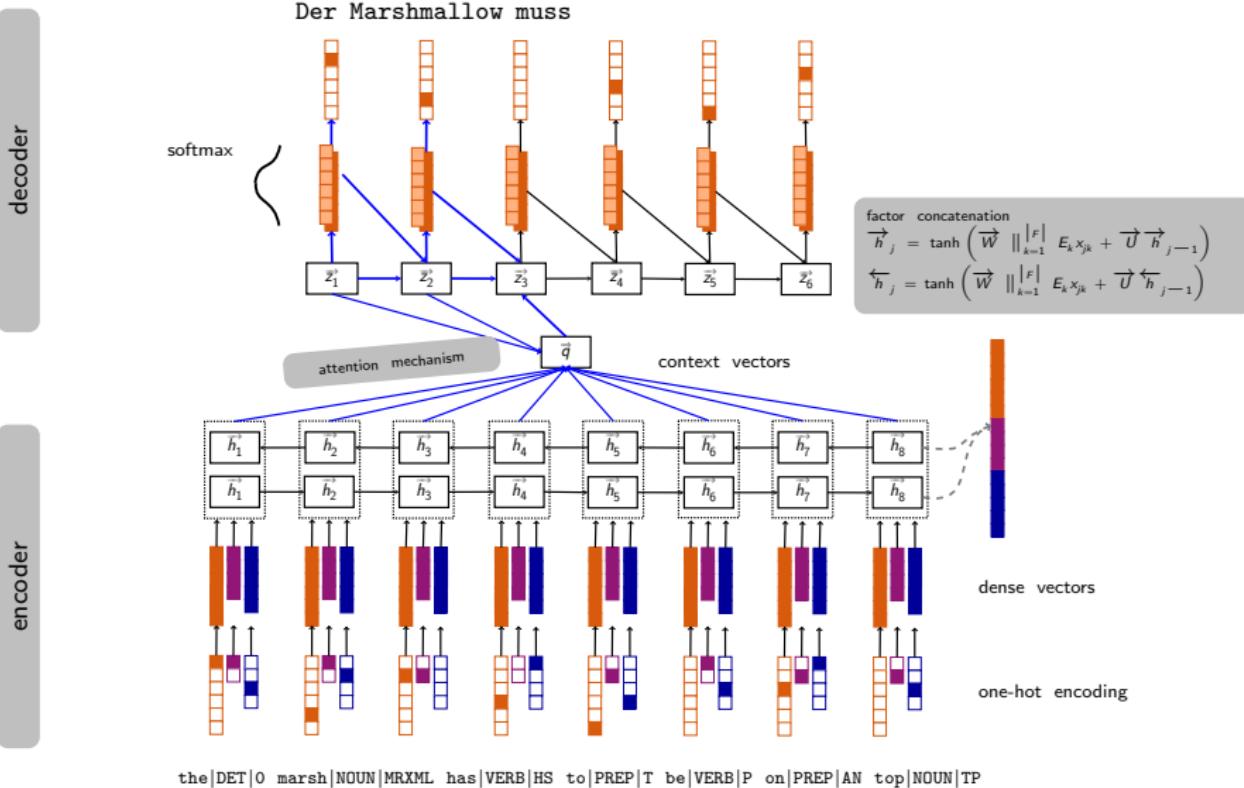
Approach

(Multilingual) NMT



Approach

Factored NMT



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Current Results

ML-NMT System Characteristics

■ Data

Corpus for *w* and *wb*, *all2all* (2,113,917 parallel fragments)

Corpus for *b*, *bn2en* (868,226 parallel fragments)

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ML-NMT System Characteristics

■ Data

Corpus for w and wb , *all2all* (2,113,917 parallel fragments)

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■ Architecture

Nematus with 150k Vocab. (+2k BPE for w and wb)

Adadelta, 506D embeddings, 800 hidden units, mini-batch size 100

Robust results with 4-ensemble

Current Results

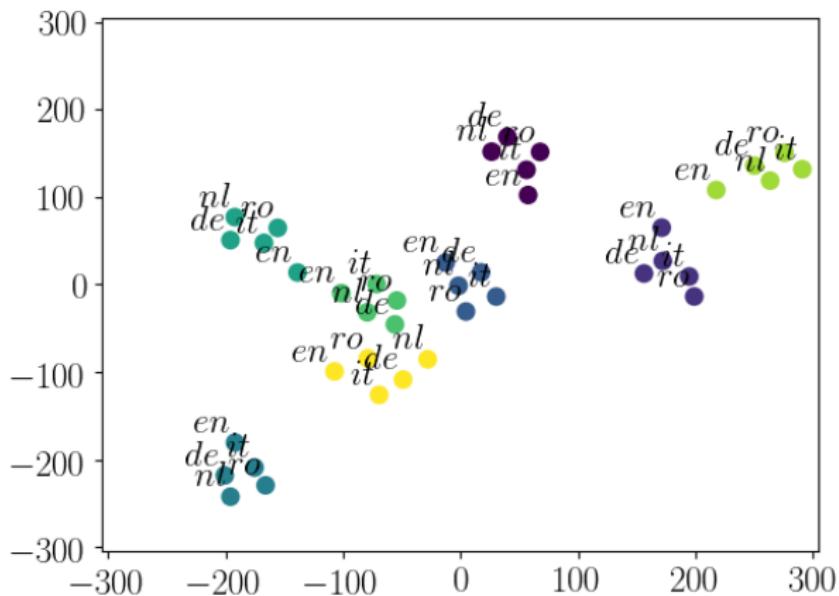
Automatic Evaluation, ML System

TED ML-NMT test set 2010

	BLEU			METEOR		
	w	wb	b	w	wb	b
<i>de2en</i>	32.6	33.0	17.5	33.1	33.5	24.2
<i>it2en</i>	33.5	33.2	21.4	33.9	34.0	27.4
<i>nl2en</i>	36.2	36.6	15.0	34.7	34.9	21.5
<i>ro2en</i>	34.3	34.8	19.6	34.4	34.6	25.9

Current Results

Automatic Evaluation, ML System



■ Too ideal corpus!

Current Results

Automatic Evaluation, IL System

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<i>ro2en</i>	34.3	34.8	19.6	34.4	34.6	25.9
<i>fr2en</i>	2.4	5.1	7.3	11.2	16.7	17.5
<i>es2en</i>	3.1	6.7	11.3	12.0	18.4	20.7

Current Results

Automatic Evaluation, IL System

de: Es|- war|bn:00083181v ein|- riesiger|- Erfolg|bn:15350982n

en: And|- it|- was|bn:00083181v a|- huge|bn:00098905a
success|bn:00075023n

it: Ed|- è|bn:00083181v stato|bn:00083181v un|- enorme|bn:00102268a
successo|bn:00078365n

nl: En|- het|- was|bn:00083181v een|- groot|- succes|bn:06512571n

ro: Sî|bn:00012706n a|- fost|bn:00083181v un|- mare|bn:00098342a
succes|bn:00075024n

Current Results

Automatic Evaluation, IL System

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- Ambiguity issues (+ multilingual ambiguity!)
- Lot of room for improvement

Conclusions

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Conclusions

Summary

- We have presented 2 ways of using **semantic information (synsets)** in a seq2seq architecture
- Improvements come specially in **beyond-zero-shot translation**
- The weakest detected point is the current need for **multilingual equivalence** and **multilingual WSD**
- The same methodology can be used in any **seq2seq** architecture (extension to CL-QA)

Conclusions

Future Work

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Data-driven “Interlingual” Translation

- Multilingual Word Sense Disambiguation
- Interlingual representation for non-content words

Thanks!

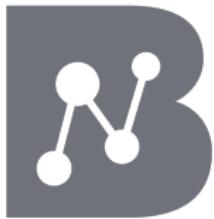
ありがとうございます。

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BabelNet

- Multilingual encyclopedic dictionary
- Semantic network
- 271 languages
- 14 million entries

Extra Slides

Multilingual Corpus: TED in numbers

	West Germanic Languages			Latin Languages			
	en	de	nl	ro	it	es	fr
Sentences	545,270	303,668	444,287	225,980	513,693	151,631	140,717
Tokens	9,768,374	5,148,199	6,894,438	3,732,679	8,367,940	2,494,336	2,473,040
uToken	141,013	221,459	187,148	213,670	200,697	148,366	131,015
uLemma	73,048	101,003	85,846	72,535	52,525	52,052	53,088
uStem	50,128	94,126	85,560	54,227	44,691	35,307	40,504
uM3	57,630	79,029	60,534	30,576	32,828	31,840	32,234
uBN	28,445	34,022	27,720	24,375	27,172	23,567	23,856

Extra Slides

Multilingual Corpus: TED in numbers

Language (iso code)	BabelNet				TED corpus	
	Lemmas	Synsets	Senses	Synonym/Synset	Synsets	Coverage (%)
English (<i>en</i>)	11,769,205	6,667,855	17,265,977	2.59	28,445	27.25
French (<i>fr</i>)	5,301,989	4,141,338	7,145,031	1.73	–	–
German (<i>de</i>)	5,109,948	4,039,816	6,864,767	1.70	34,022	23.50
Spanish (<i>es</i>)	5,022,610	3,722,927	6,490,447	1.74	–	–
Dutch (<i>nl</i>)	4,416,028	3,817,696	6,456,175	1.69	27,720	26.25
Italian (<i>it</i>)	4,087,765	3,541,031	5,423,837	1.53	27,172	29.00
Romanian (<i>ro</i>)	3,009,318	2,697,720	3,384,256	1.25	24,375	27.25