

# CUNI Submission for the MT Task

## Round 2

Covid-19 MLIA @ Eval

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## Introduction

- We participated in all language pairs (EN → AR, DE, EL, ES, IT, FR, SV)
- All our systems are constrained
- Training data for Round 2, no filtering
- Different settings:
  1. Multilingual
  2. Transfer learning

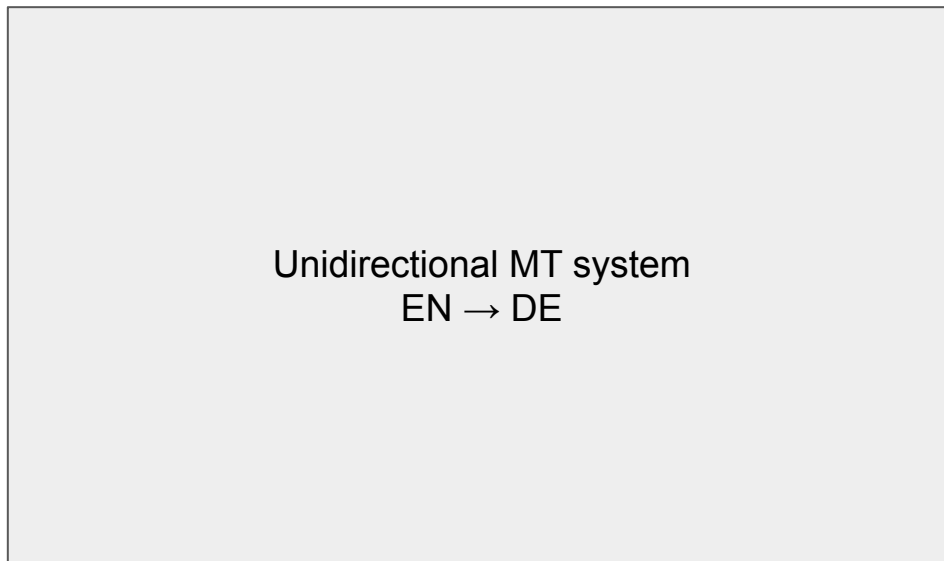
# System Description

- Same architecture
  - 6-layer Transformer with 8 heads
  - Vocabulary size 50k - shared among all languages
- Multilingual
  - One system, seven languages
  - Joint training
  - Target language embeddings
- Transfer learning
  - Pretraining on one parent language pair
  - Pretraining on a sequence of two language pairs
  - Pretraining on a mix of all language pairs

# Transfer learning

Train until  
convergence

1.



**Legend:**



One batch  
of parallel  
sentences

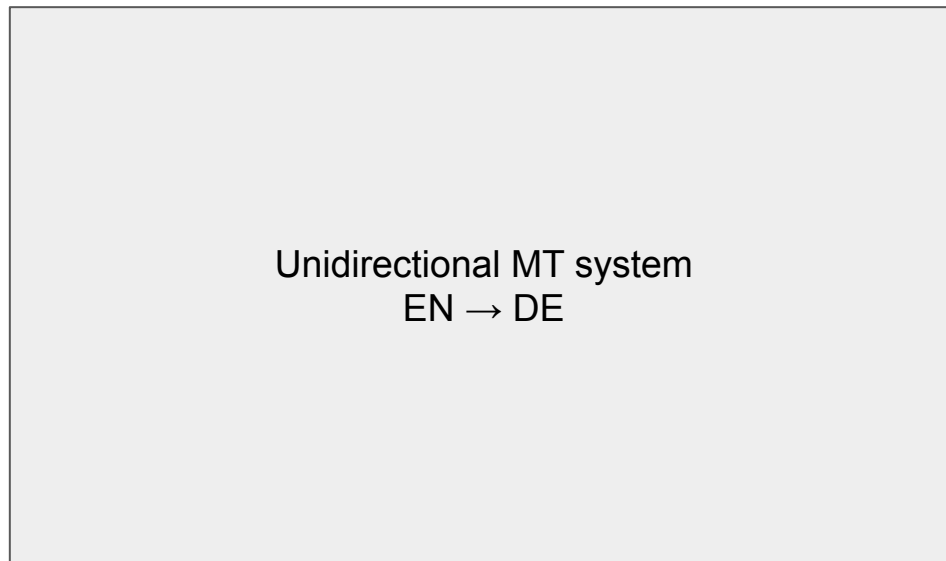


Training

# Transfer learning

Swap training corpus and fine-tune

2.



Unidirectional MT system  
EN → DE

**Legend:**



One batch of parallel sentences

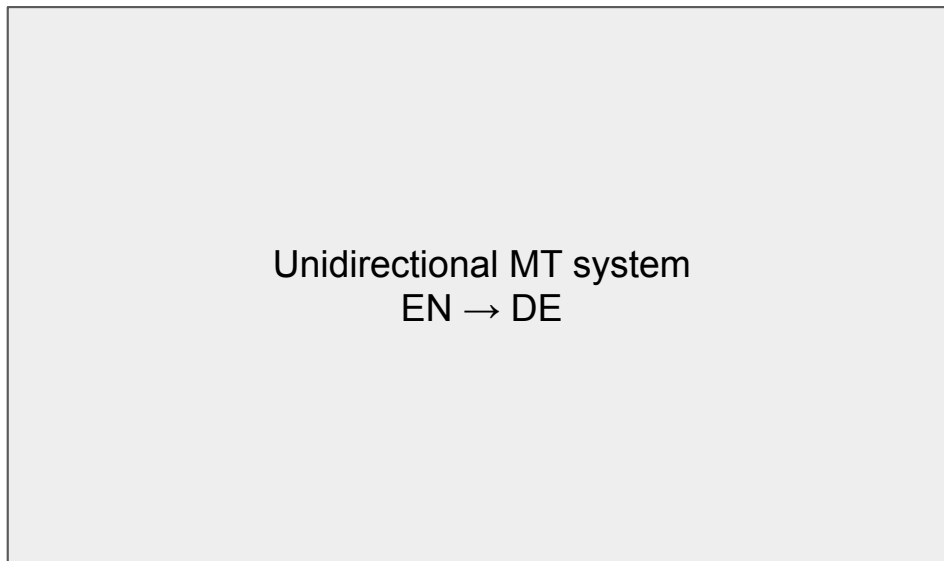
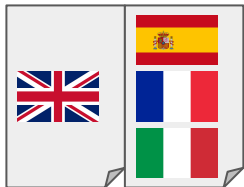


Training

# Transfer learning with multilingual pretraining

Train until  
convergence

1.



**Legend:**



One batch  
of parallel  
sentences

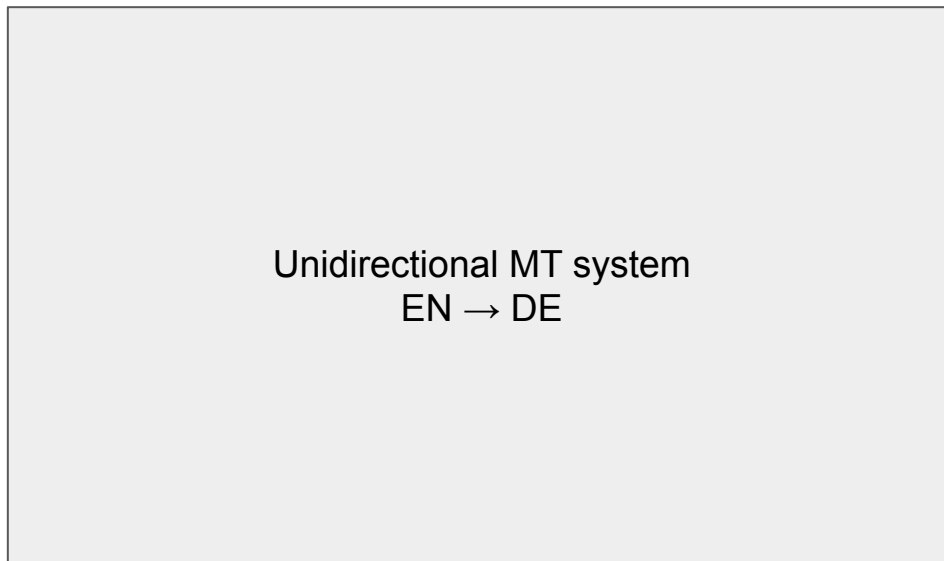


Training

# Transfer learning with multilingual pretraining

Swap training corpus and fine-tune

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Unidirectional MT system  
EN → DE

**Legend:**

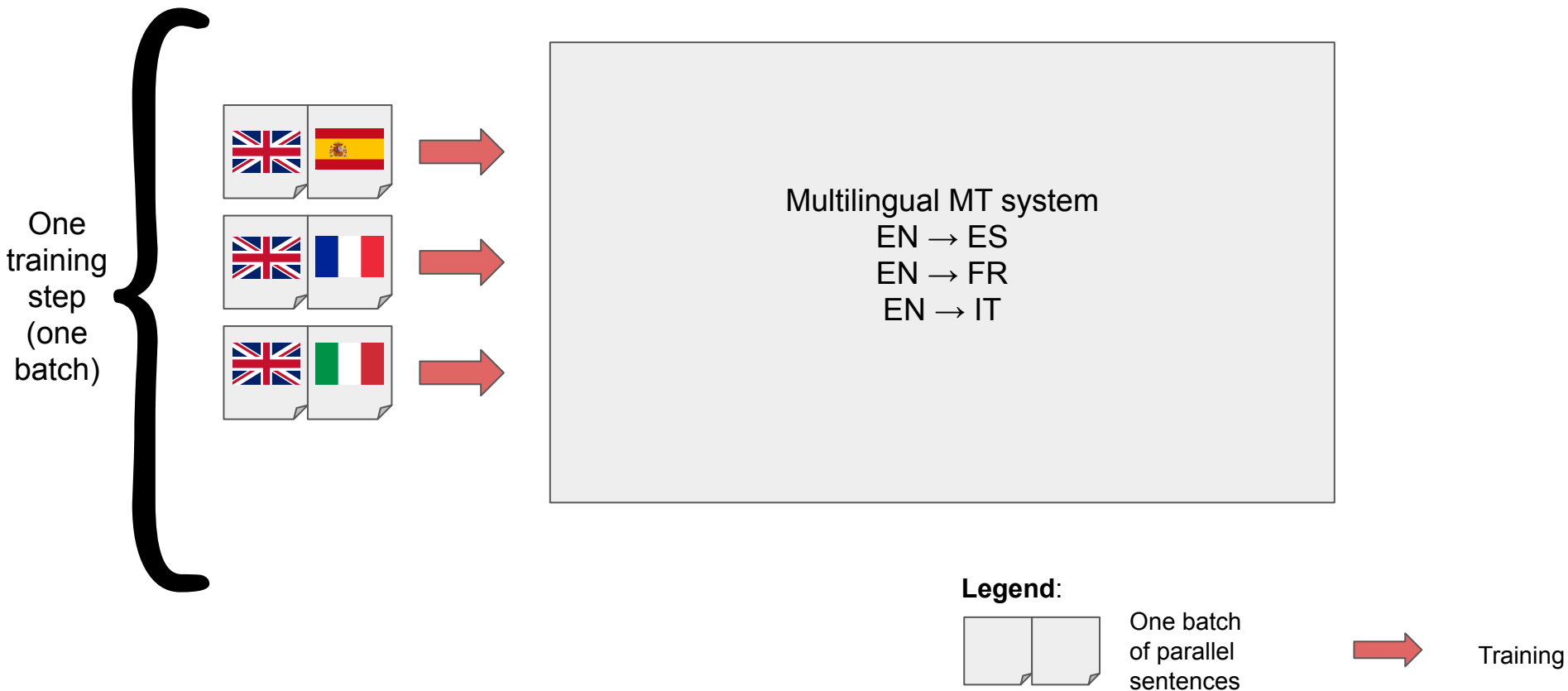


One batch of parallel sentences



Training

# Multilingual training





## Results: Transfer learning

	ar	de	el	es	fr	it	sv
ar		38.50	41.18	56.09	57.12	45.06	16.90
de	23.67		43.09	56.31	57.56	46.15	18.74
el	21.51	39.12		56.11	57.58	45.8	18.08
transfer from							
es	23.27	39.89	43.58		<b>57.87</b>	<b>47.38</b>	<b>18.75</b>
fr	<b>24.82</b>	<b>40.21</b>	<b>43.86</b>	<b>56.58</b>		47.35	18.71
it	24.49	39.18	41.93	56.1	57.45		18.38
sv	23.87	38.38	43.11	55.77	57.38	45.28	

Table 1: Transfer learning for various language pairs (child - horizontal axis, parent - vertical axis). BLEU scores on dev set.

## Results: Transfer learning

	ar	de	el	es	fr	it	sv
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sv	23.87	38.38	43.11	55.77	57.38	45.28	
all	<b>24.95</b>	39.83	<b>44.05</b>	54.76	55.43	47.39	<b>19.4</b>

## Results (BLEU scores on test set)

EN → AR	
Multilingual	17
<b>Transfer</b>	<b>19.1</b>

EN → DE	
Multilingual	27
<b>Transfer</b>	<b>27.5</b>

EN → EL	
Multilingual	34.9
<b>Transfer</b>	<b>32.4</b>

EN → ES	
Multilingual	42.1
<b>Transfer</b>	<b>48.4</b>

EN → FR	
Multilingual	44.1
Transfer	na

EN → IT	
Multilingual	35.2
<b>Transfer</b>	<b>38.6</b>

EN → SV	
<b>Multilingual</b>	<b>14.7</b>
Transfer	13.9

## Conclusion

- All language pairs significantly benefit from transfer learning
- In our setup, transfer learning is more effective than multilingual training
- The larger the parent dataset, the better
- Multilingual pretraining is beneficial for low-resource language pairs
- Possible improvements:
  - data preprocessing
  - individual vocabularies
  - back-translation

**Thank you for your attention!**

Any questions?