

CUNI Submission for the MT Task

Covid-19 MLIA @ Eval

Ivana Kvapilíková, Ondřej Bojar

Charles University
Institute of Formal and Applied Linguistics



Introduction

- We participated in all language pairs (EN → DE, EL, ES, IT, FR, SV)
- All our systems are constrained
- Different settings:
 1. Standard MT (Baseline)
 2. Online back-translation
 3. Multilingual
 4. Transfer learning

System Description

- Same architecture
 - 6-layer Transformer
 - 1024 hidden dimension
 - GELU activations
 - Language embeddings
- Training
 - Dropout 0.2
 - Vocabulary size 30k
- Inference
 - Beam search with beam size 4

Baseline

One training step (one batch)



Unidirectional MT system
EN → DE

Legend:



One batch of parallel sentences



Synthetic

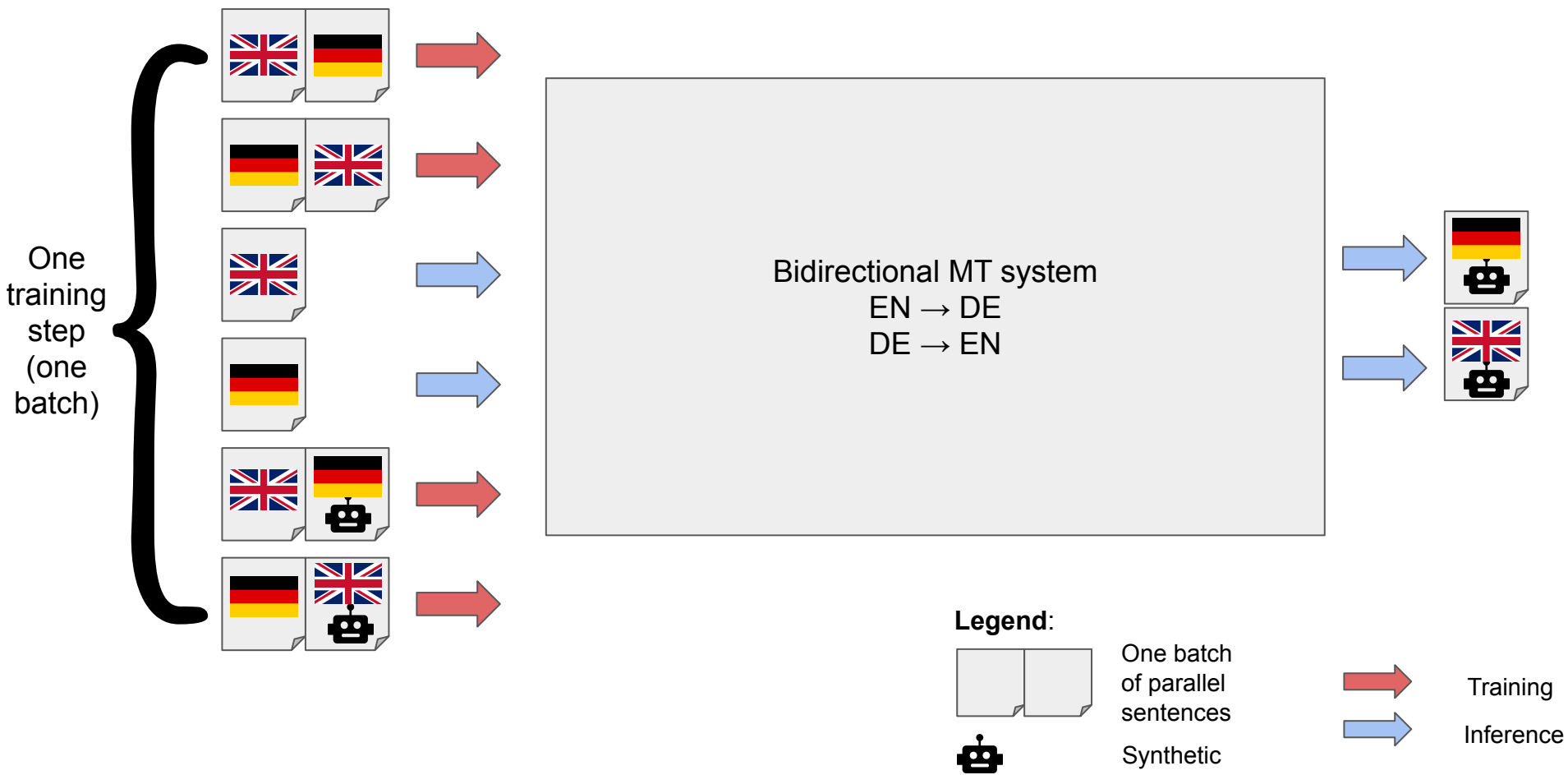


Training



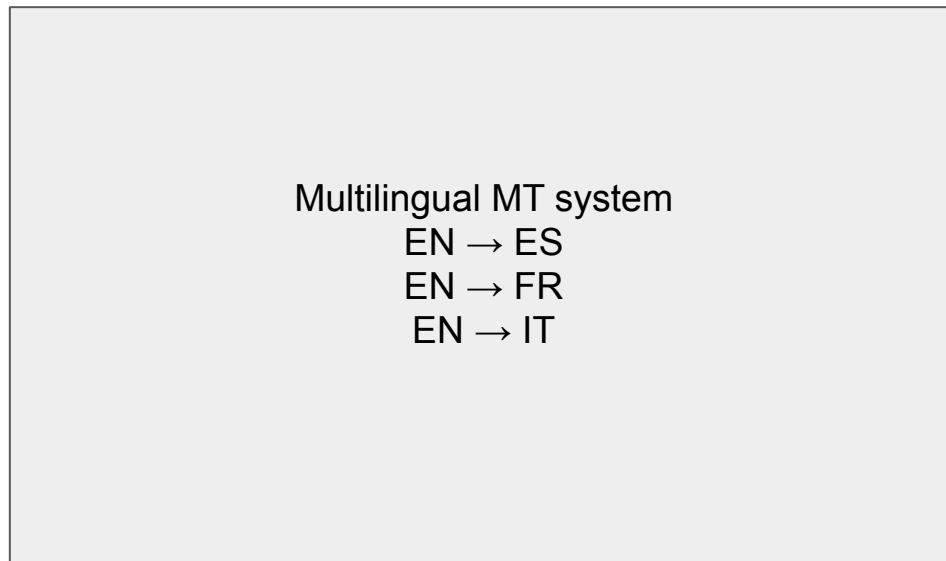
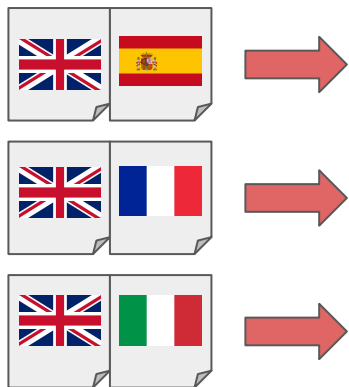
Inference

Online back-translation



Multilingual training

One training step (one batch)



Legend:



One batch of parallel sentences



Synthetic



Training

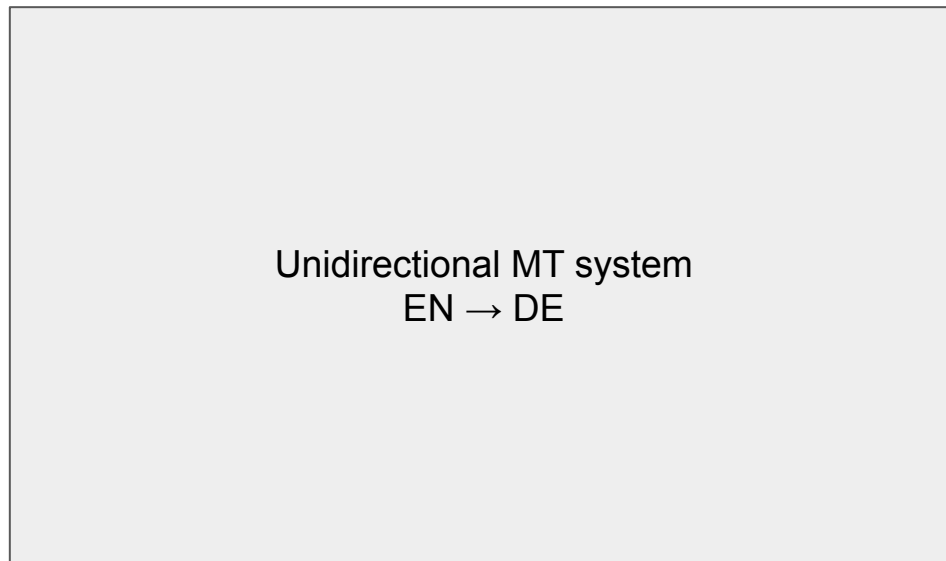


Inference

Transfer learning

Train until
convergence

1.



Legend:



Corpus of
parallel
sentences



Synthetic



Training

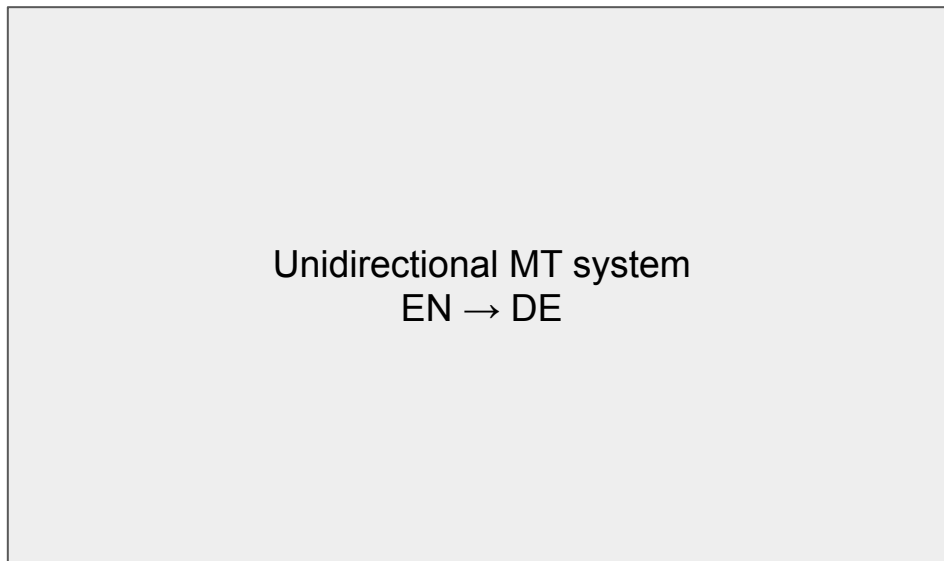


Inference

Transfer learning

Swap training corpus and fine-tune

2.



Legend:



Corpus of parallel sentences



Synthetic



Training

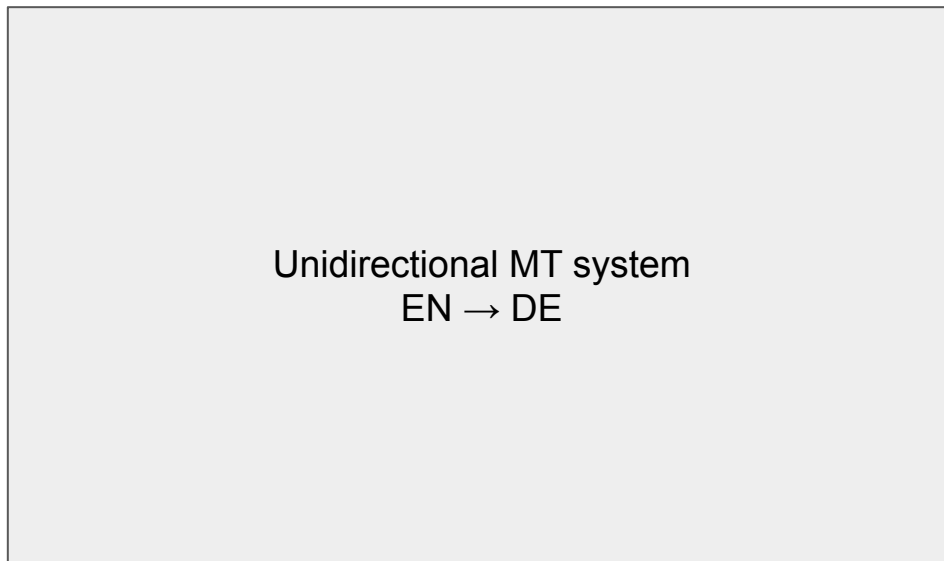


Inference

Transfer learning “with a detour”

Train until
convergence

1.



Unidirectional MT system
EN → DE

Legend:



Corpus of
parallel
sentences



Synthetic



Training

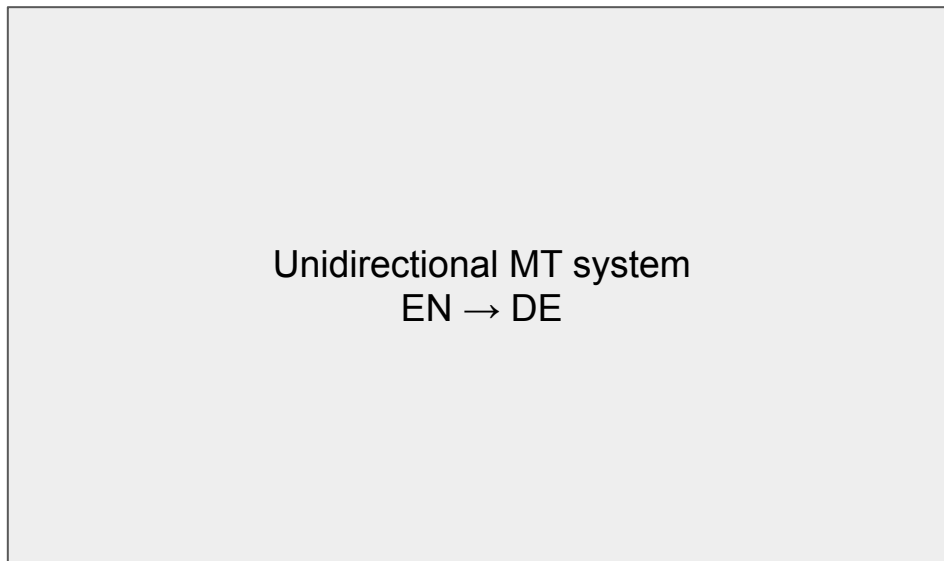


Inference

Transfer learning “with a detour”

Swap training corpus and fine-tune until convergence

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Legend:



Corpus of parallel sentences



Synthetic



Training

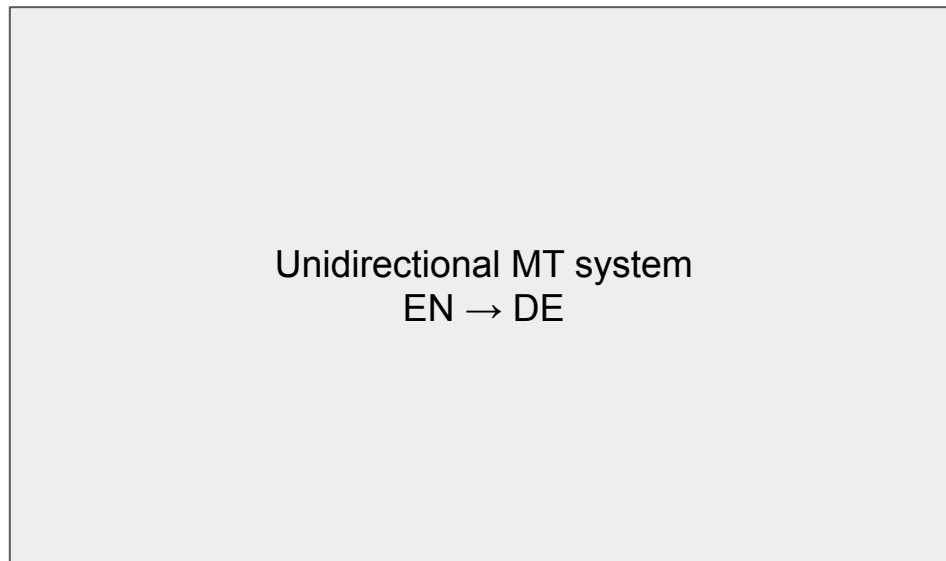


Inference

Transfer learning “with a detour”

Swap
training
corpus
and
fine-tune

3.



Legend:



Corpus of
parallel
sentences



Synthetic



Training



Inference

Results (BLEU scores on dev set)

EN → DE	
Baseline	20.76
Online BT	21.52
Multilingual	na
Transfer	22.60

EN → EL	
Baseline	22.7
Online BT	22.3
Multilingual	na
Transfer	23.29

EN → ES	
Baseline	40.46
Online BT	40.94
Multilingual	40.15
Transfer	41.34

EN → FR	
Baseline	35.57
Online BT	38.46
Multilingual	36.07
Transfer	35.10

EN → IT	
Baseline	30.97
Online BT	33.17
Multilingual	32.76
Transfer	33.07

EN → SV	
Baseline	19.13
Online BT	20.61
Multilingual	na
Transfer	22.60

Results (BLEU scores on dev set)

EN → DE		EN → EL	
Baseline	20.76	Baseline	22.7
Online BT	21.52	Online BT	22.3
Multilingual	na	Multilingual	na
Transfer	22.60	Transfer	23.29

Round 1 winner
31.6 BLEU
on test set

EN → ES	
Baseline	40.46
Online BT	40.94
Multilingual	40.15
Transfer	41.34

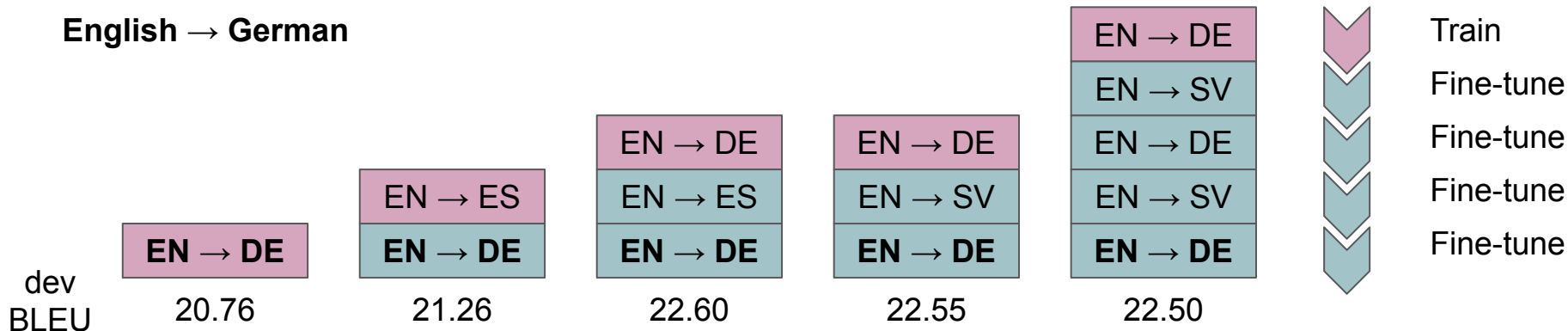
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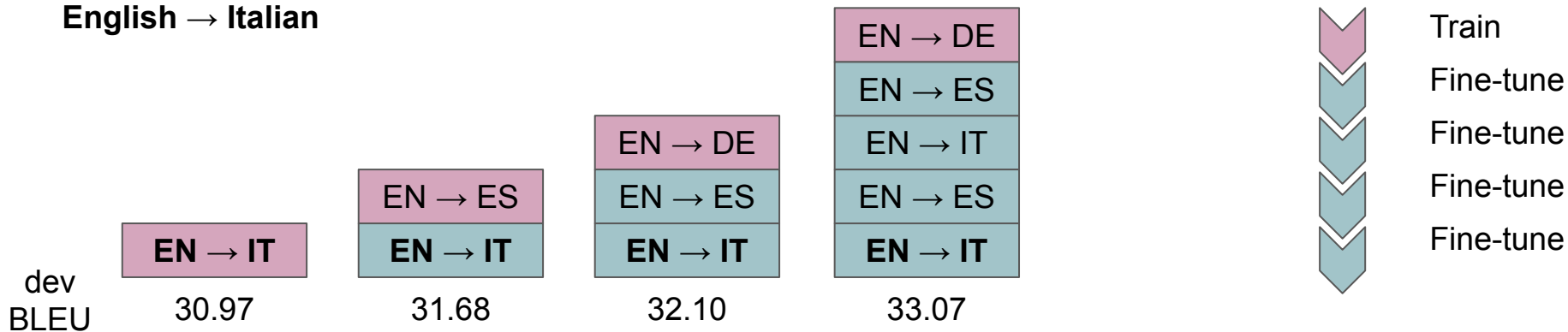
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Baseline	19.13
Online BT	20.61
Multilingual	na
Transfer	22.60

Analysis: Transfer learning

English → German



English → Italian



Conclusion

- No single winning approach
- All language pairs except for English-French significantly benefit from transfer learning
- Transfer learning more effective than multilingual training
- "Detours" during transfer learning help the final model (e.g. transferring via two languages)
- Next steps
 - Further investigate transfer learning combinations
 - Use a pretrained model from general domain (unconstrained)

Thank you for your attention!

Any questions?