

Can Cognate Prediction Be Modelled as a Low-Resource Machine Translation Task?

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Words in related languages evolved from the same ancestor

Figure 1: Cognates in our languages of interest





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 - Seq2seq correspondences $[b, u, o, n, o] \rightarrow [b, u, e, n, o]$
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 - Seq2seq correspondences $[b, u, o, n, o] \rightarrow [b, u, e, n, o]$
 - with very little data
 - ... similar to low resource MT?
- Theoretical comparison
 - Units and sample length
 - Modelled relations
 - Ambiguity management

Cognate prediction \leftrightarrow MT task?

Experiments

- Cognate data (ES-IT-LA)
- NMT (RNN, Transformers), SMT



Figure 2: Results of our different setups for a language direction

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- Data augmentation: pretrain



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- Cognate data (ES-IT-LA)
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Conclusion

About the methods

- $\bullet~\text{SMT}$ » NMT in low resource settings
- RNN » other methods when enough data
- Data augmentation:
 - ullet through monolingual lexicons igodot
 - though multilinguality 😀



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• Because of ambiguity, need to predict n-best



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About the tasks

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 \rightarrow Cognate prediction \leftrightarrow low-resource MT to an extent (needs to be studied with its specificities).



Thank you for listening, more info in the paper! Poster session at LChange'21 Code and data at: github.com/clefourrier/CopperMT