# MODERN: Modeling Discourse Entities and Relations for Coherent Machine Translation

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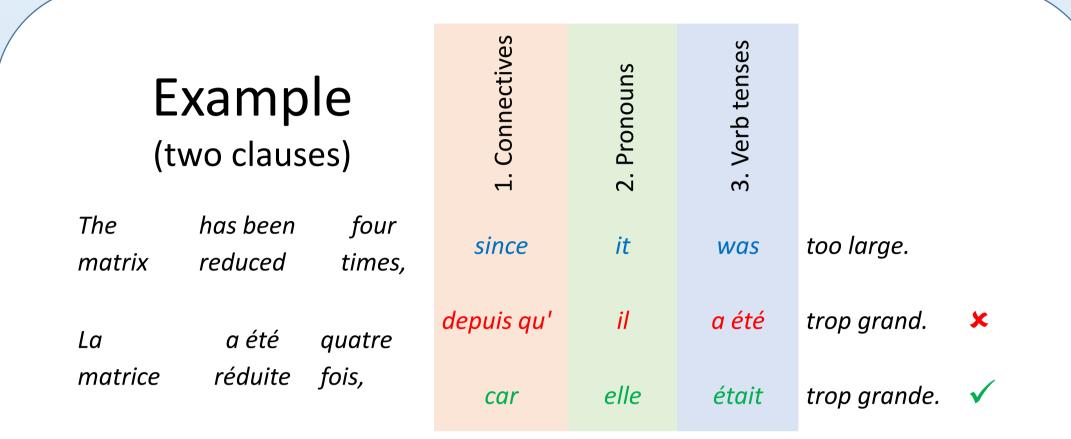
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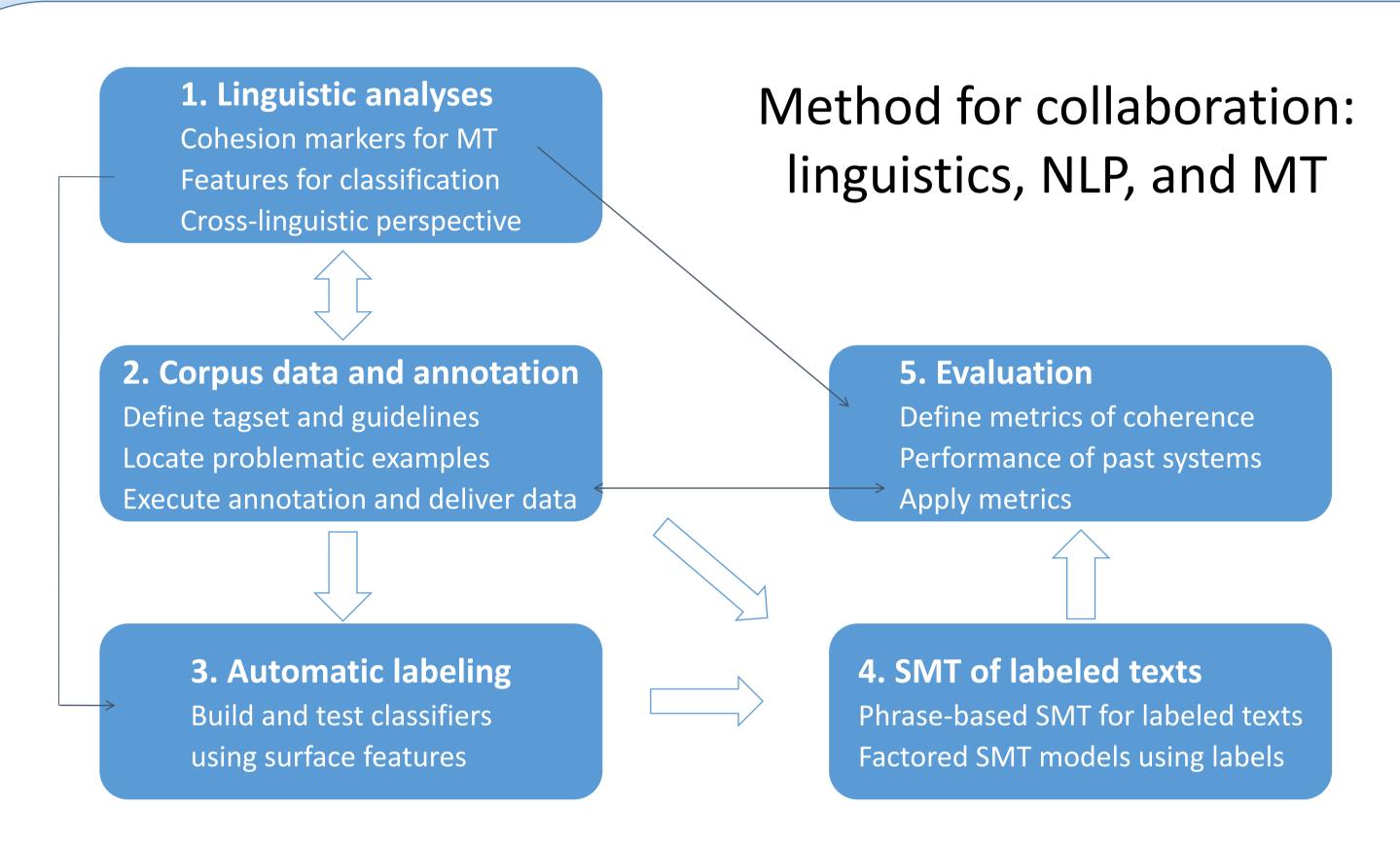
#### Motivation

- MT is efficient, has good coverage, is quite intelligible, but it always translates sentence by sentence and uses local features  $\rightarrow$  it does not propagate information across sentences or even clauses
- Such information is crucial for the correct and coherent translation of complex sentences or entire texts
  - referring information: noun phrases, pronouns
  - verbs: tense, mode, aspect
  - discourse relations, as signaled by discourse connectives
  - other features, not targeted here: style, register, politeness
- This information is not (yet) accurately captured or used by mainstream MT systems (SMT or RBMT)



Current machine translation systems: red

Using longer-range dependencies: green



# Support and related initiatives

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FNSNF

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**Continues COMTIS SNSF** Sinergia project grant n. 127510 www.idiap.ch/project/comtis

- Resources available at www.idiap.ch/dataset
- Contributed to the creation of the **DiscoMT** workshops on Discourse and Machine Translation (in conjunction with ACL 2013 and EMNLP 2015)
- Involved in the TextLink EU COST network: **Structuring Discourse in Multilingual Europe**

#### Discourse connectives

- Goal: model the human translation of connectives, annotate parallel data, disambiguate source-side connectives automatically, use the result for SMT, and measure the improvement: EN → FR, DE, IT, AR
- **Cross-lingual studies of discourse connectives** Zufferey S. and Cartoni B. (2014). A multifactorial analysis of explicitation in <u>translation</u>. *Target*, vol. 26(3), p. 361-384. Zufferey S. & Cartoni B. (2012). English and French causal connectives in contrast. Languages in Contrast. 12(2): 232-250.
- Annotation through translation spotting on Europarl Cartoni B., Zufferey S., Meyer T. (2013) - Annotating the meaning of discourse connectives by looking at their translation: The translationspotting technique. Dialogue and Discourse, vol. 4, n. 2, pp. 65-86.
- Automatic disambiguation and use for SMT with Moses Meyer T., Hajlaoui N., and Popescu-Belis A. (2015). Disambiguating Discourse Connectives for Statistical Machine Translation. IEEE/ACM Trans.
- on Audio, Speech and Language Processing, vol. 23, n. 7, p.1184-1197. Design of a reference-based evaluation metric (ACT) Hajlaoui N. and Popescu-Belis A. (2013). Assessing the Accuracy of Discourse Connective Translations: Validation of an Automatic Metric. Proc. of CICLing 2013, Springer LNCS 7817, p.236-247.
- Additional goal: model the implicitation of discourse connectives, i.e. conveying a discourse relation without using an explicit discourse connective
- **Cross-lingual modeling**

implicitly

Hoek, J., and Zufferey, S. (2015). Factors influencing the implicitation of discourse relations across languages. Proc. of the 11th Joint ACL-ISO Workshop on Interoperable Semantic Annotation (ISA-11), London. Hoek, J., Evers-Vermeul, J., & Sanders, T. (2015). The role of expectedness in the implicitation and explicitation of discourse relations. Proc. of the Second DiscoMT workshop at EMNLP 2015, Lisbon, Portugal.

**Automatic selection of connectives to translate** 

Meyer T. and Webber B. (2013). Implicitation of Discourse Connectives in (Machine) Translation. In Proc. of the First DiscoMT Workshop at ACL 2013, Sofia, Bulgaria, pages 19-26.

## Verb tenses

- Goal: model verb tense, mode and aspect; infer features for translation; study the EN  $\rightarrow$  FR divergency of Simple Past translated e.g. as *imparfait*, passé simple, or passé composé
- Cross-language modeling of verb TMA: linguistics and pragmatics Grisot C., Moeschler J. and Cartoni B. (2012). <u>Jusqu'où les temps verbaux sont-ils procéduraux?</u> Nouveaux cahiers de linguistique française, vol. 30, p. 119-139. Grisot C. (2016). Temporal coherence in discourse: theory and application for Machine

<u>Translation</u>. Blochowiak J. et al. eds., *Formal models in the study of language*, Springer.

- Annotation of narrativity, disambiguation and use for MT Grisot C. and Meyer T. (2014). Cross-Linguistic Annotation of Narrativity for English/French Verb <u>Tense Disambiguation</u>. Proc. of LREC 2014 (9th International Conference on Language Resources and Evaluation), Reykjavik. Meyer T., Grisot C. and Popescu-Belis A. (2013). Detecting Narrativity to Improve English to
- French Translation of Simple Past Verbs. Proc. of the First DiscoMT Workshop at ACL 2013, Sofia, Bulgaria, pages 33-42. Automatic alignment, disambiguation and MT of EN/FR verbs
- Loaiciga S., Meyer T. and Popescu-Belis A. (2014). English-French Verb Phrase Alignment in Europarl for Tense Translation Modeling. Proc. of LREC 2014 (9th International Conference on Language Resources and Evaluation), Reykjavik.

# Pronouns

- **Goal**: improve pronoun translation by addressing translation divergencies, in particular, for EN  $\rightarrow$  FR, the possible translations of it (as il, elle, ça, ce, etc.) and they (as ils, elles, etc.) – participated in the DiscoMT '15 and WMT '16 shared tasks
- Modeling uncertainty on coreference resolution and definition of a coreference-aware decoder Luong N.Q., Miculicich L., and Popescu-Belis A. (2015). Pronoun translation and prediction with or without coreference links. In Proc. of the Second Workshop on Discourse in Machine Translation, p. 94-100, Lisbon, Portugal. Luong N.Q. and Popescu-Belis A. (in preparation). Improving Pronoun Translation by Modeling Coreference Uncertainty.
- Defining a pronoun language model to re-rank SMT output Luong N. Q. and Popescu-Belis A. (2016). A Contextual Language Model to Improve Machine Translation of Pronouns by Re-ranking Translation Hypotheses. Proc. of EAMT 2016 (19th Annual Conference of the European Association for Machine Translation), Riga, Latvia, 13 p.
- Validating a reference-based evaluation metric
- Miculicich L. and Popescu-Belis A. (in preparation). Validation of an Automatic Metric for the Accuracy of Pronoun Translation (APT).

## Noun phrases

semantic cues.

- **Goal**: improve the consistency of noun phrase translation, in particular for repeated nouns, including compounds, for DE  $\rightarrow$  EN/FR & ZH  $\rightarrow$  EN
- Enforcing consistent translation of compound elements: XY ... Y Mascarell L., Fishel M., Korchagina N., and Volk M. (2014). Enforcing Consistent Translation of German Compound Coreferences. Proc. of KONVENS 2014 (12th German Conference on Natural Language Processing), Hildesheim, Germany. Pu X., Mascarell L., Popescu-Belis A., Fishel M., Luong N.Q., and Volk M. (2015). Leveraging Compounds to Improve Noun Phrase Translation from Chinese and German. Proc. of the ACL-IJCNLP
- 2015 Student Research Workshop, Beijing, China. Enforcing consistent translation of nouns:  $X ... X \rightarrow N_1 ... N_2$ Pu X. and Mascarell L. (in preparation). Consistent translation of repeated nouns using syntactic and