

The Second Workshop on Applying Machine Learning Techniques to Optimise the Division of Labour in Hybrid Machine Translation

COLING 2012Workshop and Shared Task



Organizers:

- Christian Federmann (German Research Center for Artificial Intelligence (DFKI))
- Dr. Maite Melero (Barcelona Media (BM))
- Dr. Marta R. Costa-jussà (Barcelona Media (BM))
- Prof. Toni Badia (Universitat Pompeu Fabra and Barcelona Media (BM))
- Dr. Tsuyoshi Okita (Dublin City University (DCU))
- Prof. Josef van Genabith (Dublin City University (DCU) and Centre for Next Generation Localisation (CNGL))

Machine Translation Paradigms



- ◆ RB-MT Rule-Based Machine translation
- ◆ EB-MT Example-Based Machine Translation
- SMT Statistical Machine Translation
- PB-SMT Phrase-Based Statistical Machine Translation
- HPB-SMT Hierachical Phrase-Based Statistical Machine Translation
- SB-SMT Syntax-Based Statistical Machine Translation

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Observation: Different systems have different strengths (e.g. easy training of SMT vs. good grammar of RB-MT)

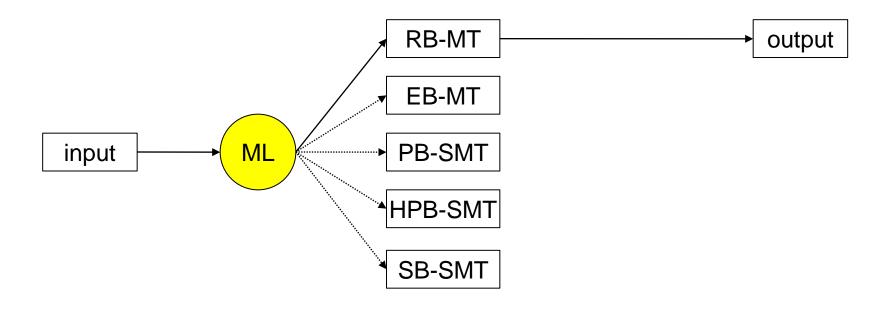
Hypothesis: can hybrid systems combine best of all?

How: Machine Learning

Hybrid MT: Pre-Translation System Selection



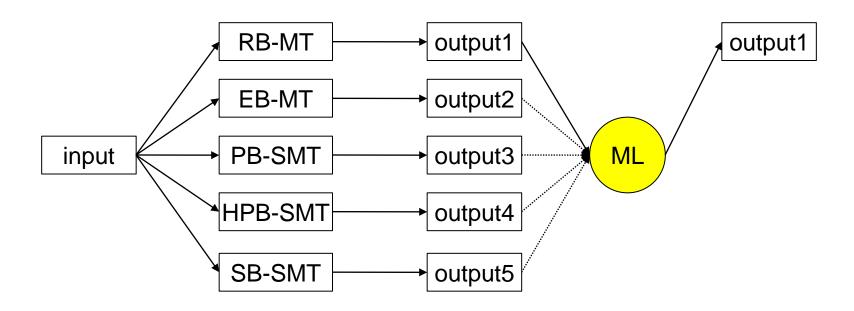
- multiple MT engines/systems available
- machine learning techniques to decide which system is best to translate input sentence



Hybrid MT: Post-Translation System Selection META NET



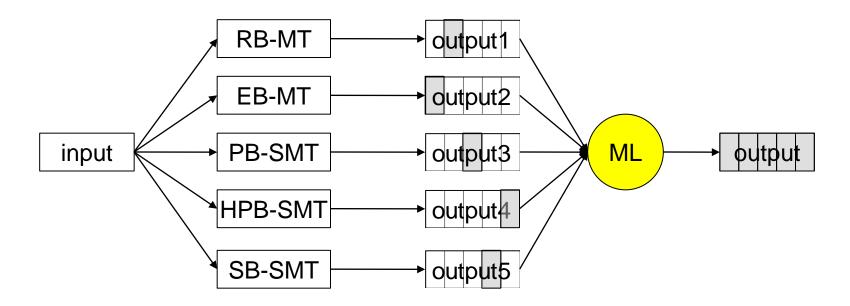
- multiple MT engines/systems available
- the input sentence is translated by all the systems and the best translation is selected based on the analysis of their ouptuts



Hybrid MT: Post-Translation System Selection META NET



- multiple MT engines/systems available
- all systems are used to produce multiple translations of the input sentence, they are broken down to smaller pieces and these are recombined to get a better output





- Based on system combination:
- Multiple systems based on different paradigms used to produce annotated nbest outputs:

 - Metis (rule based): Spanish → English, German → English
 - Apertium (rule based): Spanish ↔ English
 - Lucy (rule based): Spanish, German ↔ English
 - Joshua (hierarchical phrase based): all language pairs ↔ English
 - TectoMT (deep syntax based): Czech ↔ English
- Annotation: words, phrases, subtrees, chunks scored by different models (depending on the system)
- Decoding: machine learning to use strings + meta-data for better output



- Many more
 - Statistical post-editing: RBMT > SMT or SMT > SMT
 - Pre-ordering
 - **♦**
- System combination:
 - parallel, sequential, ..., but not just MEMT
 - probabilities in RBMT etc.

Optimising the Division of Labour in Hybrid MT



 Objectives: To provide a systematic investigation and exploration of the space of possible choices in Hybrid MT, in order to provide optimal support for Hybrid MT design, using sophisticated machine-learning (ML) technologies.

Partners:

- DFKI Deutsche Forschungszentrum für Künstliche Intelligenz (Germany)
- BM Barcelona Media (Spain)
- DCU Dublin City University (Ireland)



- Barcelona, Spain 2011
- Mumbai, India 2012 COLING
 - ◆ A very hard task
 - ♦ Heterogeneous and often incompatible meta-data
 - Difficult for ML

- WS a combination
 - Regular papers
 - Shared task
- Wider focus on general machine learning for/in MT



9:00 Josef van Genabith - Welcome and introductory remarks

Regular Papers:

9:15 **Hybrid Adaptation of Named Entity Recognition for Statistical Machine Translation** Vassilina Nikoulina, Agnes Sandor, Marc Dymetman

9:40 Confusion Network Based System Combination for Chinese Translation Output: Word-Level or Character-Level? Maoxi Li, Mingwen Wang

10:05 Using Cross-Lingual Explicit Semantic Analysis for Improving Ontology Translation Kartik Asooja, Jorge Gracia, Nitish Aggarwal, Asunción Goméz Pérez, presented by Mihael Arcan



Shared Task

- 10:30 **System Combination with Extra Alignment Information** Xiaofeng Wu
- 10:50 **Topic Modeling-based Domain Adaptation for System Combination**Antonio Toral
- 11:10 Sentence-Level Quality Estimation for MT System Combination Raphaël Rubino
- 11:30 Tea break
- 11:45 **Neural Probabilistic Language Model for System Combination**Tsuyoshi Okita
- 12:05 **System Combination Using Joint, Binarised Feature Vectors**Christian Federmann
- 12:25 Results of the ML4HMT-12 Shared Task

Christian Federmann, Tsuyoshi Okita, Maite Melero, Marta Ruiz Costa-Jussà, Toni Badia, Josef van Genabith



12:30 Discussion Panel

Panelists: Jan Hajič, Qun Liu, Hans Uszkoreit, Josef van Genabith Topics include:

- The Future of Hybrid MT: is there a single-paradigm winner?
- Will we see increasing usage of additional, potentially highly sparse, features?
- Will research efforts in Machine Translation and Machine Learning converge?
- How do we evaluate progress in terms of translation quality for Hybrid MT?
- What are the baselines? Can Human Judgment be integrated?

12:50 Invited talk: **Deep Linguistic Information in Hybrid Machine Translation**Jan Hajič · Institute of Formal and Applied Linguistics · Charles University in Prague

13:30 Lunch