

# Translation Quality Evaluation in the Molto Project (II)

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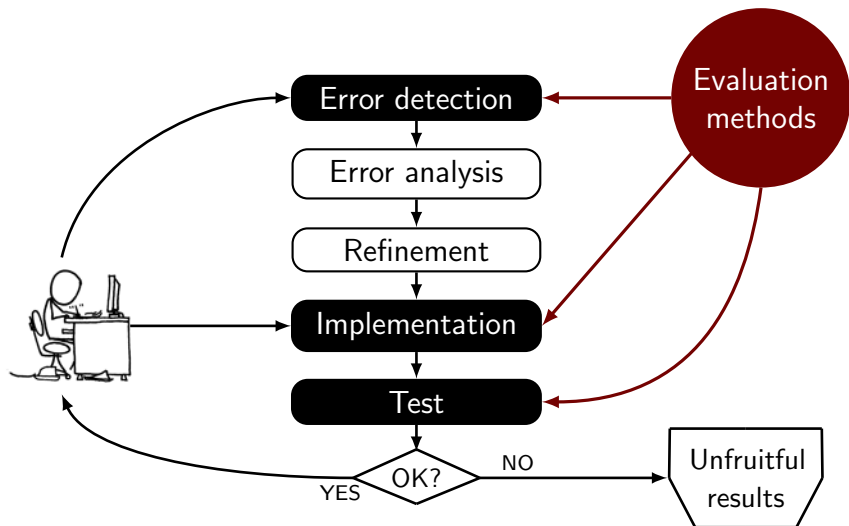
– First year project meeting –

Göteborg, March 9th, 2011

- 1 Introduction
- 2 Manual Evaluation
- 3 Automatic Evaluation
  - Motivation
  - The Asiya Software
  - Case of Study: Patents
- 4 Conclusions

# Automatic Evaluation: Motivation

*Importance for system development*



# Automatic Evaluation: Motivation

*What can be achieved with automatic evaluation?*

Automatic metrics notably **accelerate** the development cycle of MT systems:

- **Error analysis**
- **System optimisation**
- **System comparison**

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*What can be achieved with automatic evaluation?*

Automatic metrics notably **accelerate** the development cycle of MT systems:

- **Error analysis**
- **System optimisation**
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**Besides**, they are

- **Costless** (vs. costly)
- **Objective** (vs. subjective)
- **Reusable** (vs. non-reusable)

# Automatic Evaluation: Motivation

*What can be damaged with automatic evaluation?*

- **System overtuning** when system parameters are adjusted towards a given metric.
- **Blind system development** when metrics are unable to capture system improvements.
- **Unfair system comparisons** when metrics are unable to reflect difference in quality between MT systems.

### **Metrics based on lexical similarity**

(most of the metrics!)

- **Edit Distance:** WER, PER, TER
- **Precision:** BLEU, NIST, WNM
- **Recall:** ROUGE, CDER
- **Precision/Recall:** GTM, METEOR, BLANC, SIA

### Metrics based on lexical similarity

(most of the metrics!)

- **Edit Distance:** WER, PER, TER
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- **Precision/Recall:** GTM, METEOR, BLANC, SIA

Nowadays, **BLEU** is accepted as **the standard** metric.



# Automatic Evaluation: Motivation

## *Limits of lexical similarity*

The reliability of lexical metrics depends very strongly on the **heterogeneity/representativity** of reference translations.

e: This sentence **is** going to be difficult to evaluate.

Ref1: The evaluation of the translation **is** complicated.

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The reliability of lexical metrics depends very strongly on the **heterogeneity/representativity** of reference translations.

e: This sentence is going to be difficult to evaluate.

Ref1: The evaluation of the translation is complicated.

Ref2: The sentence will be hard to qualify.

Ref3: The translation is going to be hard to evaluate.

Ref4: It will be difficult to punctuate the output.

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The reliability of lexical metrics depends very strongly on the **heterogeneity/representativity** of reference translations.

e: This sentence is going to be difficult to evaluate.

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Ref2: The sentence will be hard to qualify.

Ref3: The translation is going to be hard to evaluate.

Ref4: It will be difficult to punctuate the output.

Lexical similarity is **nor a sufficient neither a necessary condition** so that two sentences convey the same meaning.

# Automatic Evaluation: Motivation

## *Going over lexical similarity*

**Extension** of the reference material:

- Using **lexical variants** such as morphological variations or synonymy lookup or using **paraphrasing** support

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Comparing other **linguistic features** than words:

- **Syntactic** similarity: shallow parsing, full parsing (constituents /dependencies).
- **Semantic** similarity: named entities, semantic roles, discourse representations, textual entailment.

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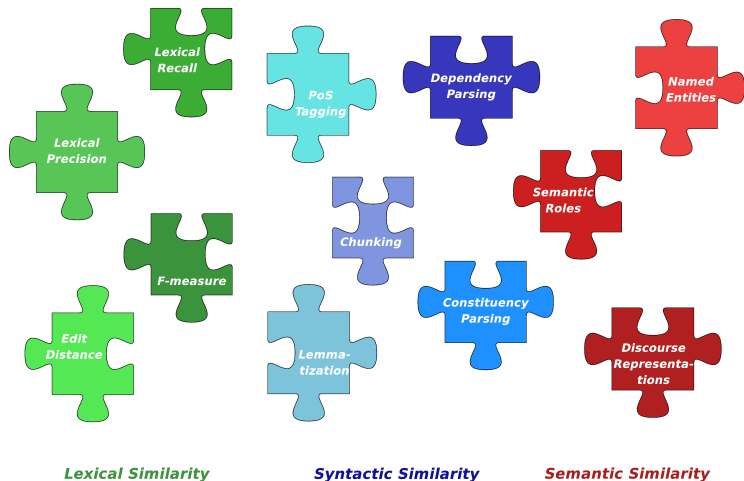
Comparing other **linguistic features** than words:

- **Syntactic** similarity: shallow parsing, full parsing (constituents /dependencies).
- **Semantic** similarity: named entities, semantic roles, discourse representations, textual entailment.

**Combination** of the existing metrics.

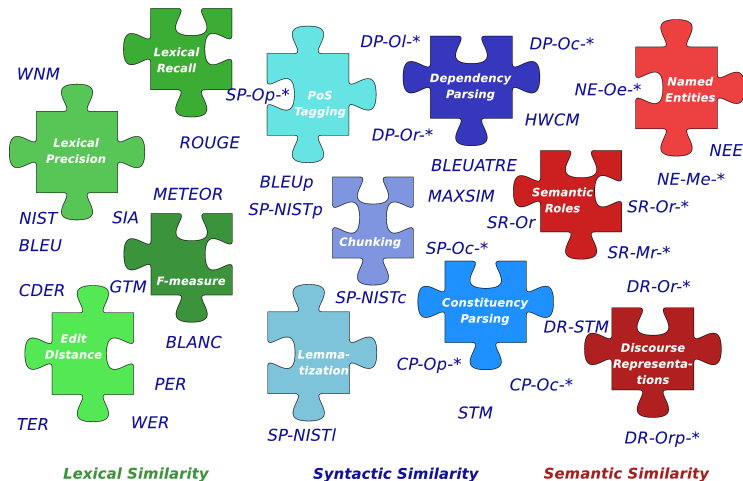
# Automatic Evaluation: Motivation

## *Towards Heterogeneous Automatic MT Evaluation*



# Automatic Evaluation: Motivation

## *Towards Heterogeneous Automatic MT Evaluation*





## ASIYA

Asiya has been designed to assist both **system** and metric **developers** by offering a rich repository of metrics and meta-metrics.

`http://www.lsi.upc.edu/~nlp/Asiya/`

# Automatic Evaluation: The Asiya Software

## *Language-dependent evaluation*

The number of available metrics depends on the available **linguistic procesors**. Currently implemented:

**English:** Lexical, Syntactic and Semantic similarity

**Spanish:** Lexical and Syntactic similarity

**German, French and others:** Lexical similarity

# Automatic Evaluation: The Asiya Software

## *Language-dependent evaluation*

The number of available metrics depends on the available **linguistic procesors**. Currently implemented:

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**Spanish:** Lexical and Syntactic similarity

**German, French and others:** Lexical similarity

*Soon!* Widening Spanish, German, French, Czech & Catalan

(FAUST project, FP7-ICT-2009-4-247762)

Web interface (OPENMT2 project, TIN2009-14675-C03)

# Automatic Evaluation: Case of Study, Patents

## *Evaluation of the SMT systems*

### **System 1** (MOLTO SMT baseline)

- **Corpus.** Chemical domain, A61P patents
- **Translation Engine.** Moses-based translator

### **System 2**

- **Google**

### **System 3**

- **Bing**

# Automatic Evaluation: Case of Study, Patents

## *English, German & French: Lexical similarity*

Lexical metrics available in **Asiya**:

```
metrics_BLEU = BLEU, BLEU-1, BLEU-2, BLEU-3, BLEU-4,  
               BLEUi-2, BLEUi-3, BLEUi-4  
metrics_GTM = GTM-1, GTM-2, GTM-3  
metrics_METEOR = METEOR-ex, METEOR-pa, METEOR-st, METEOR-sy  
metrics_NIST = NIST, NIST-1, NIST-2, NIST-3, NIST-4, NIST-5,  
              NISTi-2, NISTi-3, NISTi-4, NISTi-5  
metrics_O = O1  
metrics_PER = -PER  
metrics_ROUGE = ROUGE-1, ROUGE-2, ROUGE-3, ROUGE-4,  
               ROUGE-L, ROUGE-S*, ROUGE-SU*, ROUGE-W  
metrics_TER = -TER, -TERbase, -TERp, -TERp-A  
metrics_WER = -WER
```

# Automatic Evaluation: Case of Study, Patents

## English, German & French: Lexical similarity

Lexical metrics available in **Asiya**:

```
metrics_BLEU = BLEU, BLEU-1, BLEU-2, BLEU-3, BLEU-4,  
               BLEUi-2, BLEUi-3, BLEUi-4  
metrics_GTM = GTM-1, GTM-2, GTM-3  
metrics_METEOR = METEOR-ex, METEOR-pa, METEOR-st, METEOR-sy  
metrics_NIST = NIST, NIST-1, NIST-2, NIST-3, NIST-4, NIST-5,  
              NISTi-2, NISTi-3, NISTi-4, NISTi-5  
metrics_O = O1  
metrics_PER = -PER  
metrics_ROUGE = ROUGE-1, ROUGE-2, ROUGE-3, ROUGE-4,  
               ROUGE-L, ROUGE-S*, ROUGE-SU*, ROUGE-W  
metrics_TER = -TER, -TERbase, -TERp, -TERp-A  
metrics_WER = -WER
```

{-WER,-PER,-TER,BLEU,NIST,ROUGE-W,GTM-2,METEOR-pa}

# Automatic Evaluation: Case of Study, Patents

*English-German Translations, scores*

METRIC	DE2EN			EN2DE		
	Bing	Google	Domain	Bing	Google	Domain
1-WER	0.52	0.64	<b>0.72</b>	0.42	0.51	<b>0.69</b>
1-PER	0.66	0.76	<b>0.82</b>	0.56	0.64	<b>0.77</b>
1-TER	0.59	0.67	<b>0.76</b>	0.45	0.53	<b>0.71</b>
BLEU	0.43	0.58	<b>0.65</b>	0.33	0.45	<b>0.58</b>
NIST	8.25	9.67	<b>10.12</b>	6.53	8.05	<b>9.40</b>
ROUGE-W	0.40	0.48	<b>0.52</b>	0.34	0.41	<b>0.48</b>
GTM-2	0.30	0.40	<b>0.47</b>	0.25	0.32	<b>0.43</b>
METEOR-pa	0.60	0.69	<b>0.74</b>	0.36	0.45	<b>0.57</b>
<b>ULC</b>	0.09	0.29	<b>0.41</b>	0.03	0.19	<b>0.43</b>

# Automatic Evaluation: Case of Study, Patents

## *English-German Translations, examples*

Why such good scores?

---

<b>DE</b>	Verwendung nach Anspruch 23 , worin das molare Verhältnis von Arginin zu Ibuprofen 0,60 : 1 beträgt .
<b>EN</b>	The use of claim 23 , wherein the molar ratio of arginine to ibuprofen is 0.60 : 1 .

---



# Automatic Evaluation: Case of Study, Patents

## English-German Translations, examples

Why such good scores?

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**DE**      Verwendung nach Anspruch 23 , worin das molare Verhältnis von Arginin zu Ibuprofen 0,60 : 1 beträgt .

**EN**      **The use** of claim 23 , wherein the molar ratio of arginine to ibuprofen is 0.60 : 1 .

---

**Domain**    The use of claim 23 , wherein the molar ratio of arginine to ibuprofen is 0.60 : 1 .

**Google**    The **method** of claim 23 , wherein the molar ratio of arginine to ibuprofen 0.60 : 1 **is** .

**Bing**      ~~The~~ Use of claim 23 , wherein the molar ratio of arginine to ibuprofen is 0.60 : 1 .

---

# Automatic Evaluation: Case of Study, Patents

## *English-German Translations, examples*

What's wrong?

---

<b>DE</b>	(±)-N-(3-Aminopropyl)-N,N-dimethyl-2,3-bis(syn-9-tetradecenyl-1-oxy)-1-propanaminiumbromid
<b>EN</b>	(±)-N-(3-aminopropyl)-N,N-dimethyl-2,3-bis(syn-9-tetradecenyl-1-oxy)-1-propanaminium bromide

---

# Automatic Evaluation: Case of Study, Patents

## English-German Translations, examples

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<b>EN</b>	(±)-N-(3-aminopropyl)-N,N-dimethyl-2,3-bis(syn-9-tetradecenyloxy)-1-propanaminium bromide

---

<b>Domain</b>	(±)-N-(3-Aminopropyl)-N,N-dimethyl-2,3-bis(syn-9-tetradecenyloxy)-1-propanaminiumbromid
<b>Google</b>	(±)-N-(3-aminopropyl)-N , N-dimethyl-2 , 3-bis (syn-9-tetradecenyloxy) is 1- propanaminiumbromid
<b>Bing</b>	(±)-N-(3-Aminopropyl)-N,N-dimethyl-2,3-bis(syn-9-tetradecenyloxy)-1-propanaminiumbromid

---

# Automatic Evaluation: Case of Study, Patents

*English-French Translations, scores*

METRIC	FR2EN			EN2FR		
	Bing	Google	Domain	Bing	Google	Domain
1-WER	0.54	0.66	<b>0.78</b>	0.57	0.63	<b>0.73</b>
1-PER	0.71	0.78	<b>0.86</b>	0.68	0.75	<b>0.82</b>
1-TER	0.59	0.70	<b>0.80</b>	0.60	0.66	<b>0.74</b>
BLEU	0.45	0.62	<b>0.70</b>	0.43	0.53	<b>0.62</b>
NIST	8.52	10.01	<b>10.86</b>	8.39	9.21	<b>9.96</b>
ROUGE-W	0.41	0.50	<b>0.54</b>	0.39	0.45	<b>0.49</b>
GTM-2	0.32	0.43	<b>0.53</b>	0.31	0.36	<b>0.45</b>
METEOR-pa	0.61	0.72	<b>0.77</b>	0.57	0.65	<b>0.71</b>
<b>ULC</b>	0.07	0.28	<b>0.44</b>	0.10	0.23	<b>0.39</b>

# Automatic Evaluation: Case of Study, Patents

*German-French Translations, scores*

METRIC	DE2FR			FR2DE		
	Bing	Google	Domain	Bing	Google	Domain
1-WER	0.42	0.52	<b>0.76</b>	0.30	0.43	<b>0.65</b>
1-PER	0.58	0.68	<b>0.77</b>	0.46	0.59	<b>0.74</b>
1-TER	0.47	0.56	<b>0.68</b>	0.32	0.46	<b>0.66</b>
BLEU	0.29	0.43	<b>0.56</b>	0.24	0.39	<b>0.53</b>
NIST	6.72	8.21	<b>9.10</b>	5.35	7.30	<b>8.88</b>
ROUGE-W	0.31	0.38	<b>0.45</b>	0.29	0.37	<b>0.44</b>
GTM-2	0.24	0.30	<b>0.41</b>	0.21	0.28	<b>0.41</b>
METEOR-pa	0.45	0.56	<b>0.64</b>	0.26	0.39	<b>0.51</b>
<b>ULC</b>	0.03	0.22	<b>0.41</b>	-0.03	0.19	<b>0.44</b>

# Conclusions

## *Evaluation within MOLTO*

- MOLTO uses both manual and automatic evaluation.
- For a fast development process automatic metrics are very useful.
- But, for the automatic evaluation one needs reference translations.
- Manual evaluation assures a high quality final evaluation.

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# Conclusions

## *System evaluation with Asiya*

```
Asiya.pl -eval single,ulc -g sys Asiya.config
```



# Conclusions

## *System evaluation with Asiya*

```
Asiya.pl -eval single,ulc -g sys Asiya.config
```

```
input=raw
```

```
SRCLANG=de
```

```
TRGLANG=en
```

```
SRCCASE=cs
```

```
TRGCASE=cs
```

```
#SRC =====  
src=./data/patsA61P.test.de  
#REF =====  
ref=./data/patsA61P.test.en  
#OUT =====  
sys=./data/patsA61P.test.trans.de2en  
sys=./data/patsA61P.test.trad.google.de2en  
sys=./data/patsA61P.test.trad.bing.de2en  
#-----
```

# Conclusions

## *System evaluation with Asiya*

```
Asiya.pl -eval single,ulc -m metrSet Asiya.config
```

```
SRCLANG=de
```

```
TRGLANG=en
```

```
#SRC =====
```

```
src=./data/patsA61P.test.de
```

```
#REF =====
```

```
ref=./data/patsA61P.test.en
```

```
#OUT =====
```

```
sys=./data/patsA61P.test.trans.de2en
```

```
#-----
```

```
metrSet=1-PER 1-TER 1-WER BLEU-4 CP-0c-* CP-Op-* CP-STM-9 DP-HWC-c-4  
DP-HWC-r-4 DP-HWC-w-4 DP-0c-* DP-0l-* DP-Or-* DR-Or-* DR-Orp-* DR-STM-9  
GTM-1 GTM-2 GTM-3 MTR-exact MTR-stem MTR-wnstm MTR-wnsyn NE-Me-* NE-Oe-*  
NE-Oe-** NIST-5 RG-L RG-S* RG-SU* RG-W-1.2 SP-0c-* SP-Op-* SP-cNIST-5  
SP-iobNIST-5 SP-lNIST-5 SP-pNIST-5 SR-Mr-* SR-Mrv-* SR-Or SR-Or-* SR-Orv
```

