

Deep evaluation of hybrid architectures: Use of different metrics in MERT weight optimization

Cristina España-Bonet, Gorka Labaka,
Arantza Díaz de Ilarraza, Kepa Sarasola, Lluís Màrquez

Free/Open-source Rule-based Machine Translation

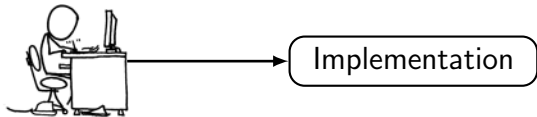
Gothenburg, June 14th, 2012

Overview

- 1 Motivation
- 2 SMatxinT, a hybrid translator
- 3 Systems' evaluation
- 4 Conclusions

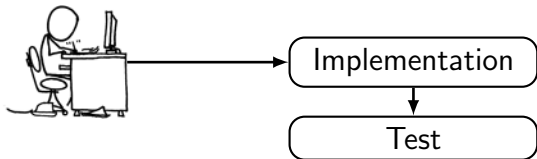
Motivation

Evaluation during system development



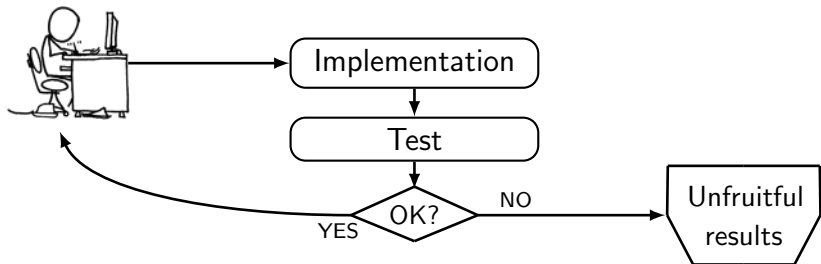
Motivation

Evaluation during system development



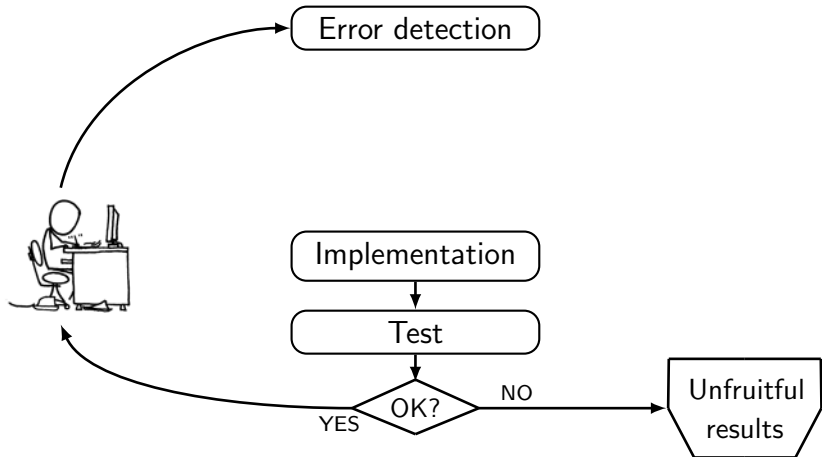
Motivation

Evaluation during system development



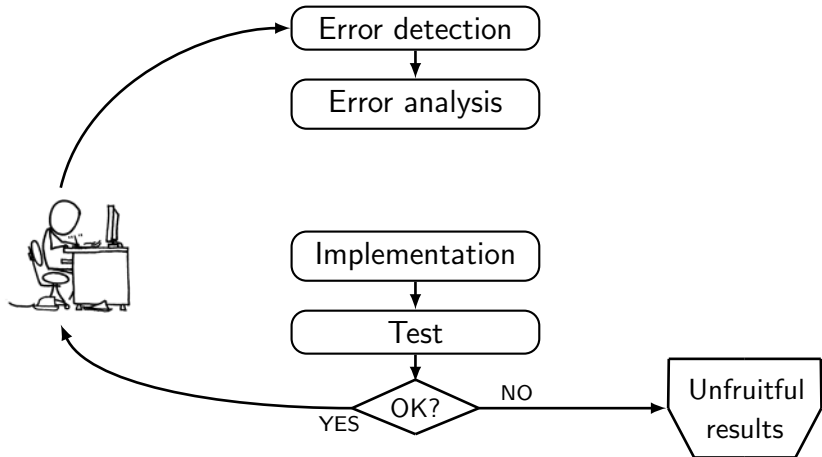
Motivation

Evaluation during system development



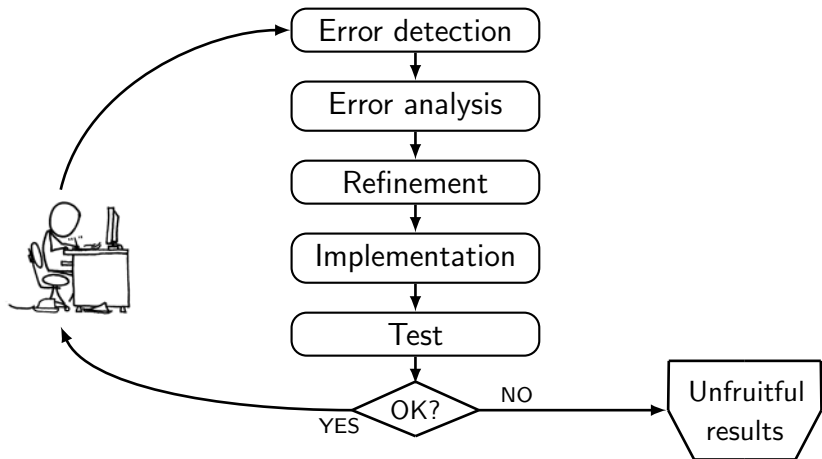
Motivation

Evaluation during system development



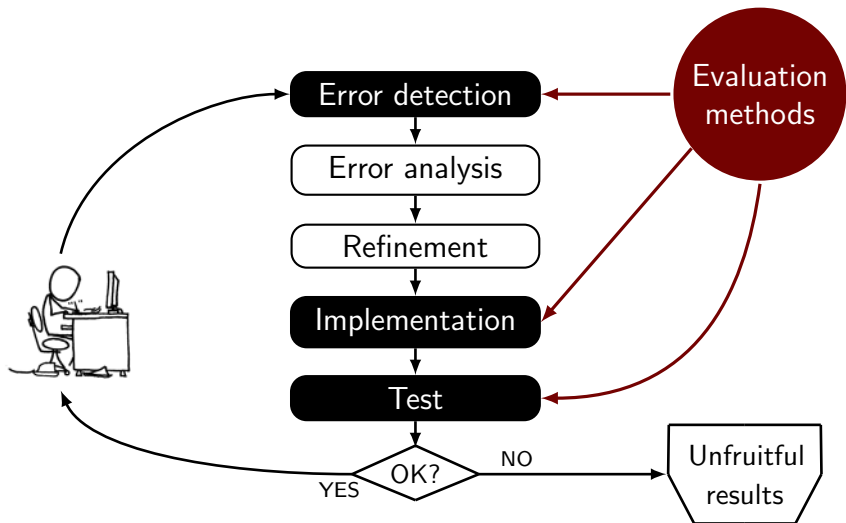
Motivation

Evaluation during system development



Motivation

Evaluation during system development



Motivation

Need for an automatic evaluation

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

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

Motivation

Need for an automatic evaluation

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

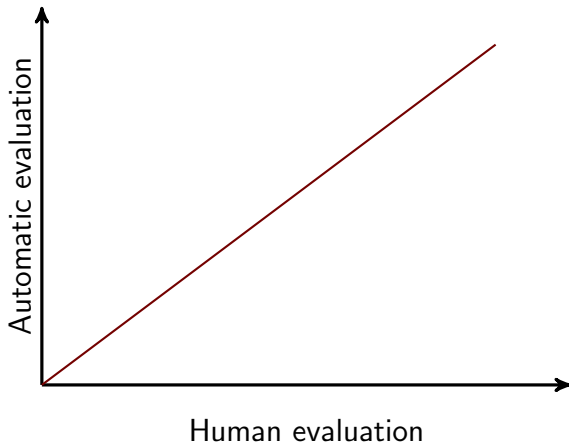
- **Error analysis**
- **System optimisation**
- **System comparison**

Besides, they are

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

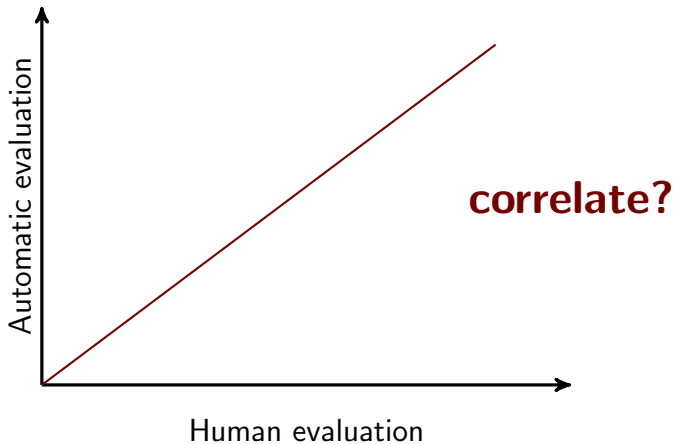
Motivation

But...



Motivation

But...



Motivation

Our System Evaluation



Automatic evaluation

Motivation

Our System Evaluation



Automatic evaluation



Manual evaluation

Motivation

Problem

System development using a metric that
does not correlate with human ranking

Are we worsening the system?

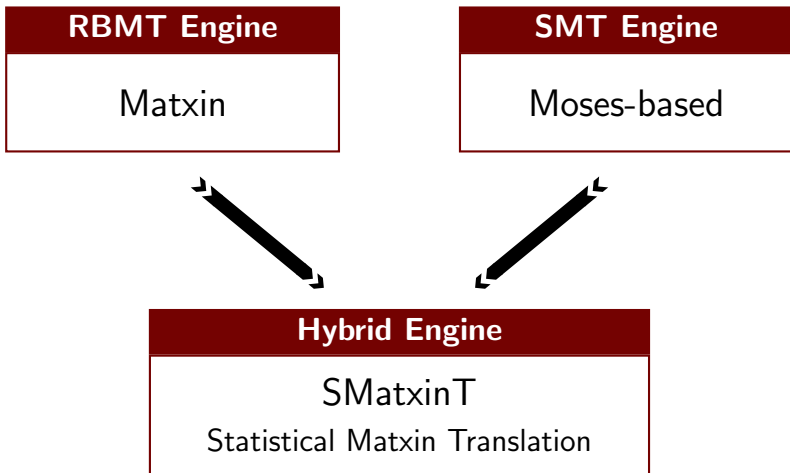
SMatxinT, a hybrid translator

Overview

- 1 Motivation
- 2 SMatxinT, a hybrid translator
 - Individual systems
 - Hybrid system
- 3 Systems' evaluation
- 4 Conclusions

SMatxinT, a hybrid translator

Hybrid engine components



SMatxinT, a hybrid translator

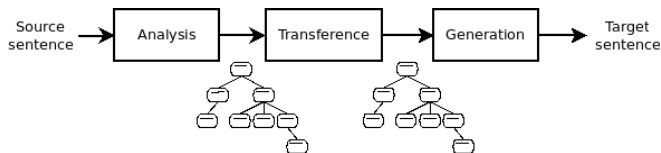
SMT System

- **Basque segmentation**
- **Language model:** 3-gram interpolated Kneser-Ney discounting, SRILM Toolkit
- **Alignments:** GIZA++ Toolkit
- **Translation model:** Moses package
- **Weights optimization:** MERT against BLEU
- **Decoder:** Moses

SMatxinT, a hybrid translator

RBMT System

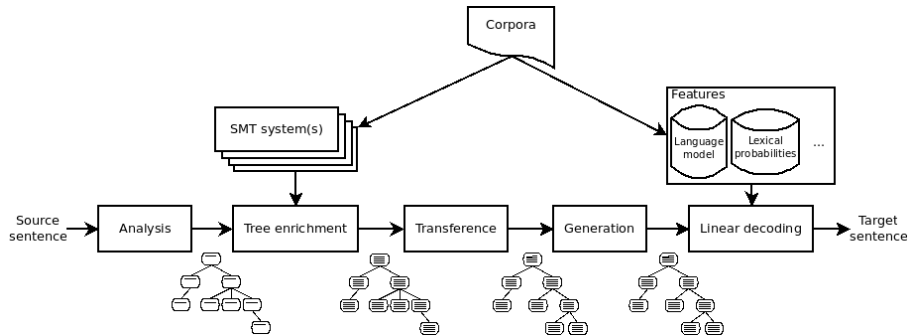
Matxin, Open-Source Rule-Based MT system



- Chunk-based dependency tree
(Dependency trees + chunk boundaries)

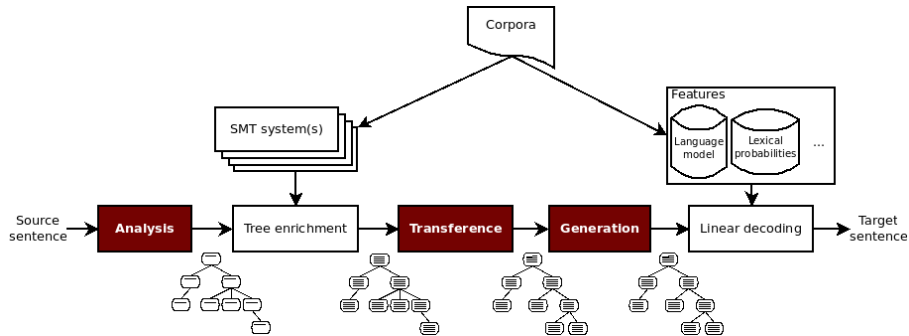
SMatxinT, a hybrid translator

SMatxinT System



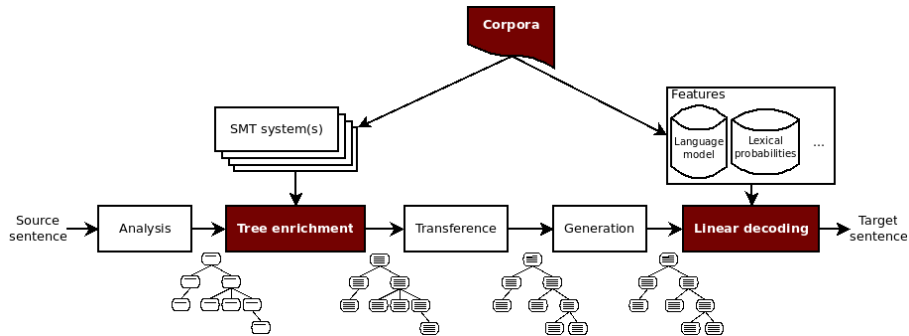
SMatxinT, a hybrid translator

SMatxinT System



SMatxinT, a hybrid translator

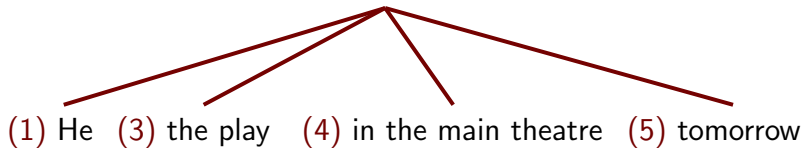
SMatxinT System



SMatxinT, a hybrid translator

Tree enrichment

(2) is going to see



SMatxinT, a hybrid translator

Tree enrichment

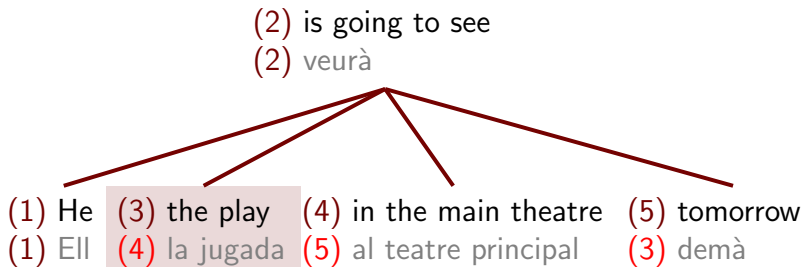
(2) is going to see

(2) veurà

(1) He (3) the play (4) in the main theatre (5) tomorrow
(1) Ell (4) la jugada (5) al teatre principal (3) demà

SMatxinT, a hybrid translator

Tree enrichment

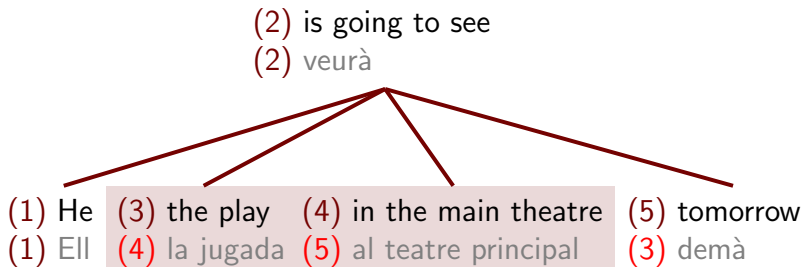


SMT: l'obra

...

SMatxinT, a hybrid translator

Tree enrichment



SMT: l'obra

SMT: l'obra

l'obra al cinema principal

l'obra al teatre principal

...

SMatxinT, a hybrid translator

Tree enrichment: purpose

The RBMT system

- ensures syntactic correctness,
- and takes care of long distance reordering.

Additional richness of phrases:

- Short phrases to improve lexical selection
- Long phrases to overcome wrong syntactic analysis

SMatxinT, a hybrid translator

Linear decoding

_____→

(1) (2) (5) (3) (4)
He is going to see tomorrow the play in the main theatre

SMatxinT, a hybrid translator

Linear decoding

(1) (2) (5) (3) (4)
He is going to see tomorrow the play in the main theatre

Ell	veurà	demà	la jugada	al teatre principal
-----	-------	------	-----------	---------------------

SMatxinT, a hybrid translator

Linear decoding

(1) (2) (5) (3) (4)
He is going to see tomorrow the play in the main theatre

Ell	veurà	demà	la jugada	al teatre principal
Ell ϕ	veurà mirarà ...	demà	l'obra la jugada ...	al teatre principal al cinema principal al teatre del centre

SMatxinT, a hybrid translator

Linear decoding

(1) (2) (5) (3) (4)
He is going to see tomorrow the play in the main theatre

Ell	veurà	demà	la jugada	al teatre principal
Ell ϕ	veurà mirarà ...	demà	l'obra la jugada ...	al teatre principal al cinema principal al teatre del centre
Ell ϕ	veurà mirarà ...	demà	l'obra al cinema del centre l'obra al teatre principal ...	

SMatxinT, a hybrid translator

Linear decoding

(1) (2) (5) (3) (4)
He is going to see tomorrow the play in the main theatre

Ell	veurà	demà	la jugada	al teatre principal
-----	-------	------	-----------	---------------------

Ell	veurà	demà	l'obra	al teatre principal
ϕ	mirarà		la jugada	al cinema principal
	al teatre del centre

Ell	veurà	demà	l'obra al cinema del centre	
ϕ	mirarà		l'obra al teatre principal	
	

...

Anirà a veure demà l'obra al teatre principal

Ell mirarà demà la jugada al teatre principal

...

SMatxinT, a hybrid translator

Linear decoding: Features

Standard SMT features

- Language model
- Word penalty
- Phrase penalty

SMatxinT, a hybrid translator

Linear decoding: Features

Standard SMT features

- Language model
- Word penalty
- Phrase penalty

Source/consensus features

- Counter ($1...n$)
- SMT ($1/e$)
- RBMT ($1/e$)
- Both ($e^\#$)

SMatxinT, a hybrid translator

Linear decoding: Features

Standard SMT features

- Language model
- Word penalty
- Phrase penalty

Source/consensus features

- Counter ($1 \dots n$)
- SMT ($1/e$)
- RBMT ($1/e$)
- Both ($e^\#$)

Lexical features

- Corpus lexical probabilities ($e_u \rightarrow e_s$ & $e_s \rightarrow e_u$)
- Dictionary lexical probabilities ($e_u \rightarrow e_s$ & $e_s \rightarrow e_u$)

SMatxinT, a hybrid translator

The log-linear model & MERT

Features, $h_m(f, e)$

$$\hat{e} = \operatorname{argmax}_e \log P(e|f) = \operatorname{argmax}_e \sum \lambda_m h_m(f, e)$$

MERT

λ_m that minimise $\Delta_{err} \equiv \operatorname{metric}(\hat{e}) - \operatorname{metric}(e_{ref})$

- 1 Motivation
- 2 SMatxinT, a hybrid translator
- 3 Systems' evaluation
 - Setting
 - Automatic evaluation
 - Manual evaluation
- 4 Conclusions

Language pair

- Spanish–Basque

Training corpus

- Administrative documents and TV programs descriptions
- 491,853 parallel sentences

Development and test corpora

- *Elhuyar dev&test*: Administrative documents (1500 snt)
- *NEWS*: News (1500 sentences, 2 references)

Individual systems

- SMT
- Matxin

Individual systems

- SMT
- Matxin

Hybrid systems

- SMatxinT with different MERT

λ_m that minimise $\text{metric}(\hat{e}) - \text{metric}(e_{ref})$

metric: BLEU, BLEU_C, METEOR

Individual systems

- SMT
- Matxin

Hybrid systems

- SMatxinT with different MERT

λ_m that minimise $\text{metric}(\hat{e}) - \text{metric}(e_{ref})$

metric: BLEU, **BLEU_C**, METEOR

Hybrid systems

- SMatxinT with different MERT

λ_m that minimise $\text{metric}(\hat{e}) - \text{metric}(e_{ref})$

metric: BLEU, BLEU_C, METEOR

$$\text{BLEU}_C = (\text{BLEU} + \text{BLEU}_{PoS})/2$$

Hybrid systems

- SMatxinT with different MERT

λ_m that minimise $\text{metric}(\hat{e}) - \text{metric}(e_{ref})$

metric: BLEU, BLEU_C, METEOR

$$\text{BLEU}_C = (\text{BLEU} + \text{BLEU}_{PoS})/2$$

Control system

- Google

Systems' evaluation

In-domain automatic evaluation

	BLEU	METEOR	TER	BLEU _c
Matxin	6.07	27.20	83.49	19.65
SMT	16.50	37.49	70.39	27.64

Systems' evaluation

In-domain automatic evaluation

	BLEU	METEOR	TER	BLEU _c
Matxin	6.07	27.20	83.49	19.65
SMT	16.50	37.49	70.39	27.64
Google	8.19	28.02	78.43	20.73
SMatxinT _{BL}	16.09	38.24	69.92	27.95
SMatxinT _{BL_c}	15.36	38.24	70.78	27.33
SMatxinT _{MTR}	15.87	37.77	67.77	27.53

Systems' evaluation

Out-of-domain automatic evaluation

	BLEU	METEOR	TER	BLEU _c
Matxin	12.67	36.10	69.16	31.98
SMT	15.84	37.70	66.52	31.01

Systems' evaluation

Out-of-domain automatic evaluation

	BLEU	METEOR	TER	BLEU _c
Matxin	12.67	36.10	69.16	31.98
SMT	15.84	37.70	66.52	31.01
Google	12.36	32.57	70.44	29.08
SMatxinT _{BL}	16.61	39.24	64.50	32.77
SMatxinT _{BL_c}	17.11	39.94	63.84	33.39
SMatxinT _{MTR}	16.76	39.30	62.83	32.50

Systems' evaluation

Manual evaluation

- 100 sentences in-domain, 100 sentences out-of domain
- 2 evaluators for each sentence

1st. experiment: 5 systems to rank per sentence

8.: El Supremo ordena juzgar a patrones de cayucos interceptados en alta mar

() ⇒ gorenak epaitzeko agindu dio eredu cayucos interceptados en alta mar

() ⇒ gorenak cayucos-interceptados eredu itsaso zabalean epaitzeko agindu

() ⇒ Supremok cayucos geldituen patroiak itsas zabalean epaitzea ordenatzen du

() ⇒ gorenak cayucos-interceptados eredu en alta mar epaitzeko agindu du

() ⇒ gorenak epaitzeko agindu dio patrones de cayucos interceptados en alta mar

Systems' evaluation

Manual evaluation

- Ranking allows ties

1.: La oposición cree que el unico relevo necesario es el del alcalde

- () ⇒ oposizioaren ustez beharrezkoa da txanda bakarra alkatearen
- () ⇒ oposizioak sinesten du beharrezko txanda bakarra alkatetik dela
- () ⇒ oposizioaren ustez beharrezkoa da txanda bakarra alkatearen
- () ⇒ oposizioaren ustez beharrezkoa da txanda bakarra alkatearen
- () ⇒ oposizioaren ustez beharrezkoa da txanda bakarra alkatearen

Systems' evaluation

Manual evaluation

- Ranking allows ties

1.: La oposición cree que el unico relevo necesario es el del alcalde

- () \Rightarrow oposizioaren ustez beharrezkoa da txanda bakarra alkatearen
- () \Rightarrow oposizioak sinesten du beharrezko txanda bakarra alkatetik dela
- () \Rightarrow oposizioaren ustez beharrezkoa da txanda bakarra alkatearen
- () \Rightarrow oposizioaren ustez beharrezkoa da txanda bakarra alkatearen
- () \Rightarrow oposizioaren ustez beharrezkoa da txanda bakarra alkatearen

- Two measures:

Ranking mean, from $[1,5]$ to $[1,1]$ if ties

Normalisation to $[0,1]$

Systems' evaluation

Manual evaluation

	in-domain		out-of-domain	
	Rank	Norm	Rank	Norm
Matxin	2.07	0.396	1.70	0.275
SMT	2.51	0.532	2.60	0.625

Systems' evaluation

Manual evaluation

	in-domain		out-of-domain	
	Rank	Norm	Rank	Norm
Matxin	2.07	0.396	1.70	0.275
SMT	2.51	0.532	2.60	0.625
SMatxinT_{BL}	2.16	0.423	2.21	0.485
SMatxinT_{BL_c}	2.08	0.399	2.11	0.445
SMatxinT_{MTR}	2.09	0.403	2.12	0.470

Systems' evaluation

Manual evaluation

2nd experiment: Discrete ranking (instead of mean values)

Each system is qualified for each sentence as

best

intermediate

worst

all-draw

Systems' evaluation

Manual evaluation

in-domain

	Best	Intermediate	Worst	All-draw
Matxin	24 (34+42)	9 (26+19)	20 (38+32)	0 (2+7)
SMT	9 (22+23)	7 (31+23)	30 (45+47)	0 (2+7)
SMatxinT _{BL}	8 (27+19)	22 (52+43)	8 (19+31)	0 (2+7)
SMatxinT _{BL_c}	12 (27+18)	29 (55+45)	7 (16+30)	0 (2+7)
SMatxinT _{MTR}	6 (28+19)	24 (54+47)	6 (16+27)	0 (2+7)

Systems' evaluation

Manual evaluation

out-of-domain

	Best	Intermediate	Worst	All-draw
Matxin	47 (51+64)	4 (22+12)	10 (25+19)	0 (2+5)
SMT	7 (20+11)	6 (21+25)	41 (57+59)	0 (2+5)
SMatxinT _{BL}	11 (28+15)	27 (44+43)	21 (26+37)	0 (2+5)
SMatxinT _{BL_c}	12 (27+17)	28 (50+44)	15 (21+34)	0 (2+5)
SMatxinT _{MTR}	11 (26+16)	26 (46+42)	18 (26+37)	0 (2+5)

System evaluation

- Evaluation and comparison of SMT, Matxin and SMatxinT.
- Human evaluation and lexical metrics do not correlate in this case.
- BLEU_c does a slightly better job than BLEU.

System development

- Development (MERT) should use metrics that correlate with human assessments in order to improve translations.
- $\text{SMat}_{\text{inT}}_{\text{BL}_c}$ best hybrid, but minimal differences.

System development

- Development (MERT) should use metrics that correlate with human assessments in order to improve translations.
- $\text{SMat}_{\text{inT}_{\text{BL}_c}}$ best hybrid, but **minimal differences**.
- **Why?**

MERT development

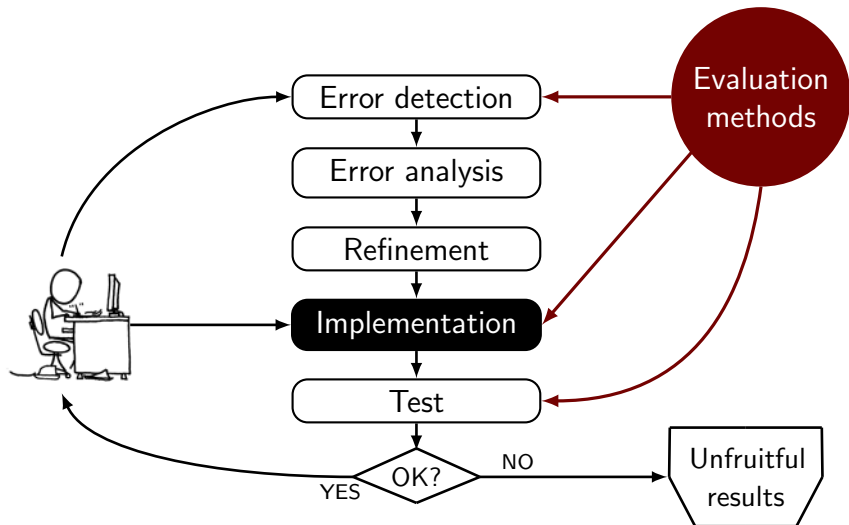
- We do not provide MERT with features sensible to the metric.

Next steps

- Include more linguistic features in SMatxinT.
- Define a metric that includes these features to be used with MERT.

Conclusions

Current and next steps



Conclusions

Thanks for your attention

Thank you!

Deep evaluation of hybrid architectures: Use of different metrics in MERT weight optimization

Cristina España-Bonet, Gorka Labaka,
Arantza Díaz de Ilarraza, Kepa Sarasola, Lluís Màrquez

Free/Open-source Rule-based Machine Translation

Gothenburg, June 14th, 2012