

Can the Internet help Improve Machine Translation?

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MT output: Assassinated a diplomat Russian and kidnapped other four in Baqdad



MT output: Assassinated a diplomat Russian and kidnapped other four in Bagdad

Could hire experts to correct machine translations



MT output: Assassinated a diplomat Russian and kidnapped other four in Bagdad

Could hire experts to correct machine translations

Expensive + time consuming



MT output: Assassinated a diplomat Russian and kidnapped other four in Bagdad

 Not feasible for large amounts of data (google, yahoo, etc.)

Does not generalize to new sentences



My Solution

SL: Asesinado un diplomático ruso y secuestrados otros cuatro en Bagdad

TL: Assassinated a diplomat Russian and kidnapped other four in Bagdad

- Get non-expert bilingual speakers to provide correction feedback online
 - Make correcting translations easy and fun
 - 5-10 minutes a day → Large amounts of correction data



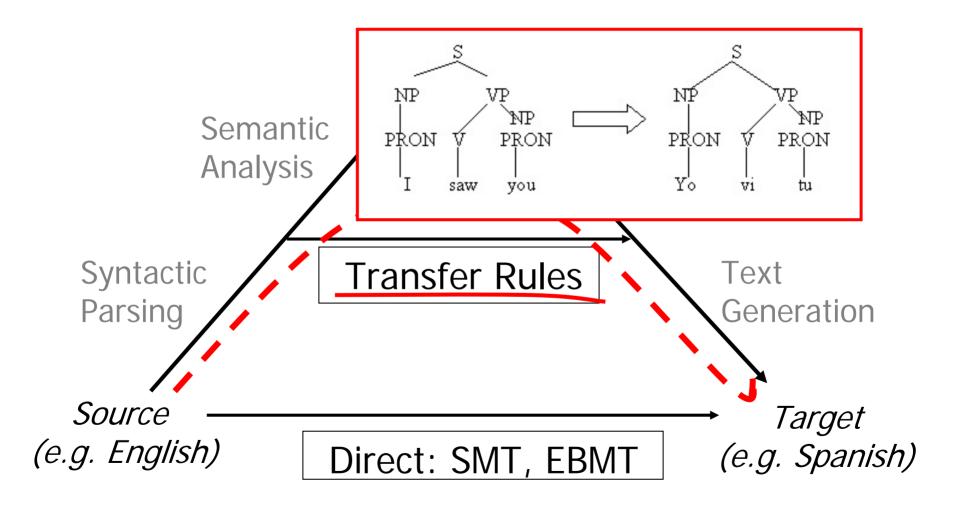
My Solution

SL: Asesinado un diplomático ruso y secuestrados otros cuatro en Bagdad

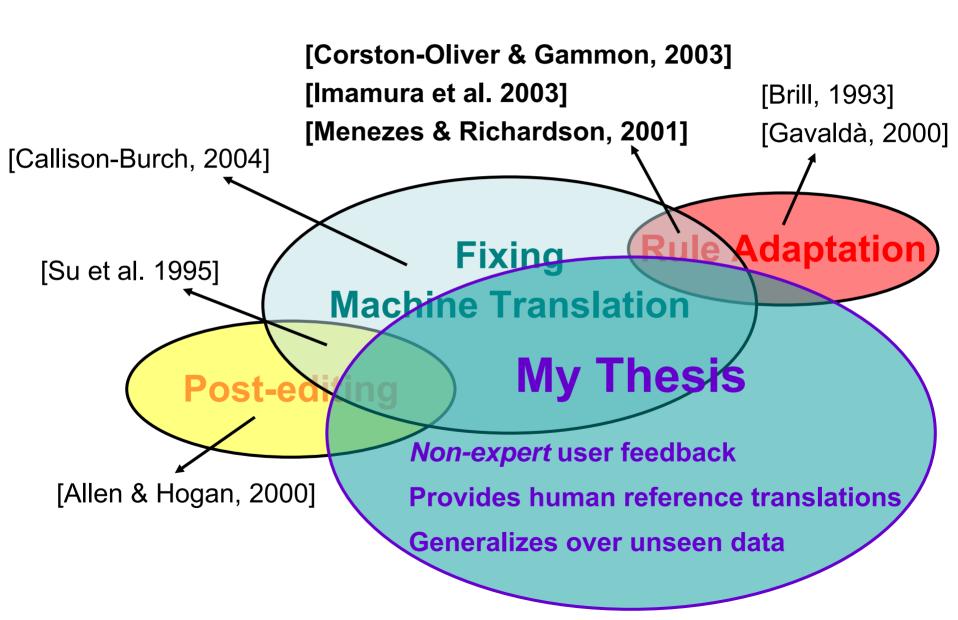
TL: Assassinated a diplomat Russian and kidnapped other four in Bagdad

- Get non-expert bilingual speakers to provide correction feedback online
 - Make correcting translations easy and fun
 - 5-10 minutes a day → Large amounts of correction data
- Feed corrections back into the MT system, so that they can be generalized
 - → System will translate new sentences better

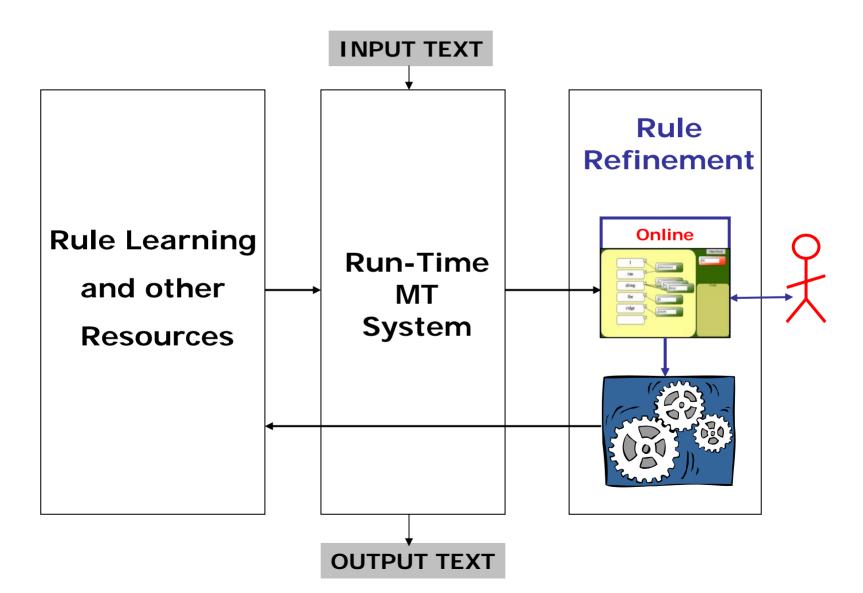
MT Approaches



Related Work

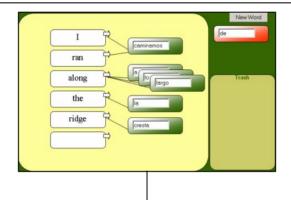


System Architecture

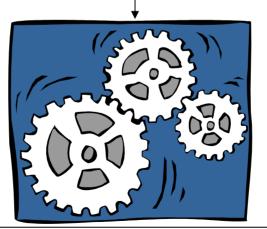


Main Technical Challenge

Simple user edits to MT output

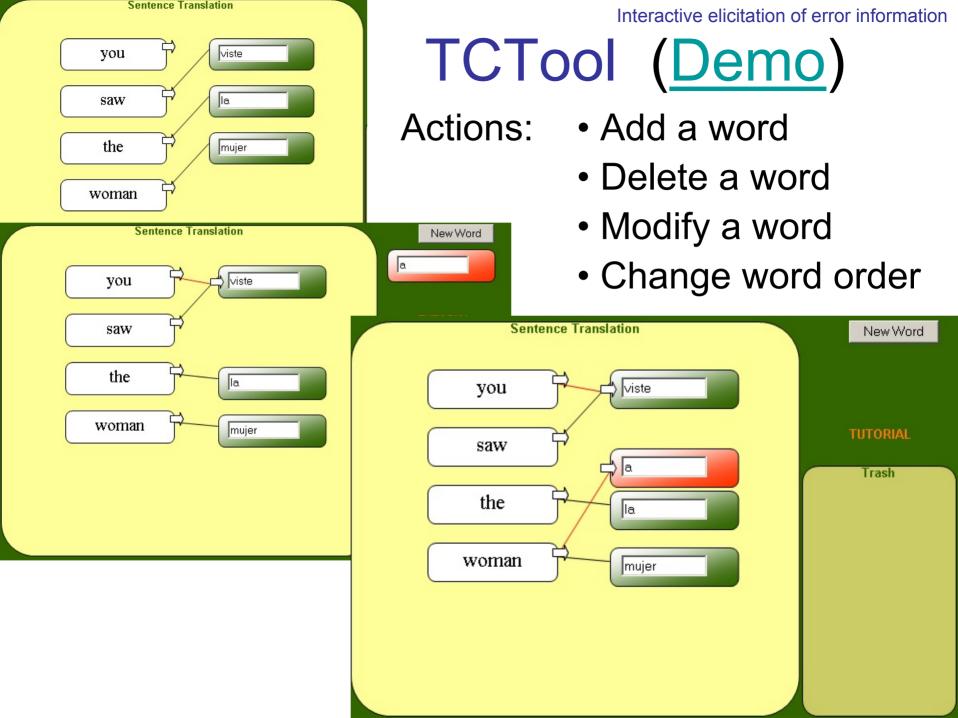


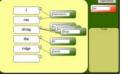
Mapping between



Blame assignment
Rule Modifications
Lexical Expansions

Improved Translation Rules





Eng2Spa User Study [LREC 2004]

	precision	recall
error detection	90%	89%
error classification	72%	71%

 MT error classification → 9 linguisticallymotivated classes [Flanagan, 1994], [White et al. 1994]:

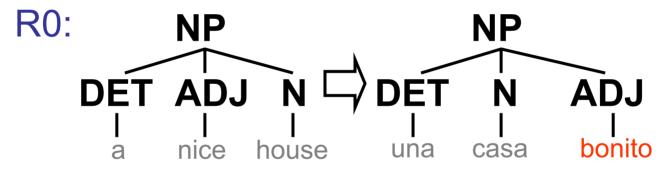
word order, sense, agreement error (number, person, gender, tense), form, incorrect word and no translation

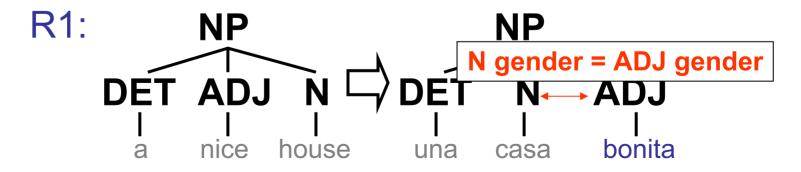


Types of Refinement Operations

1. Refine a translation rule:

R0 → R1 (<u>change</u> R0 to make it more specific or more general)





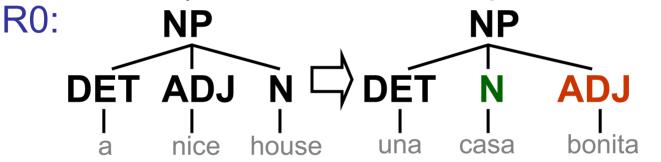


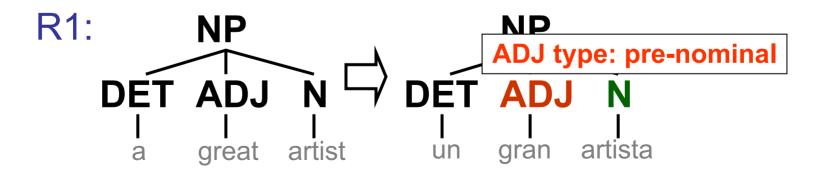
Types of RefinementOperations (2)

2. Bifurcate a translation rule:

 $R0 \rightarrow R0$ (same, general rule)

→ R1 (add a new more specific rule)







Formalizing Error Information

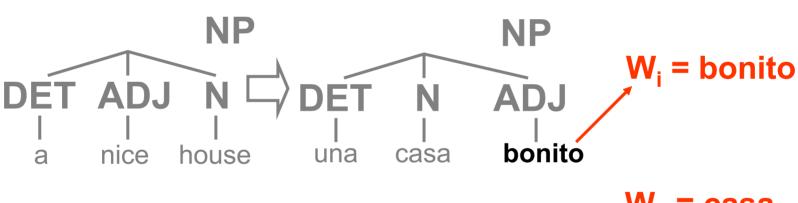
$$W_i = error$$

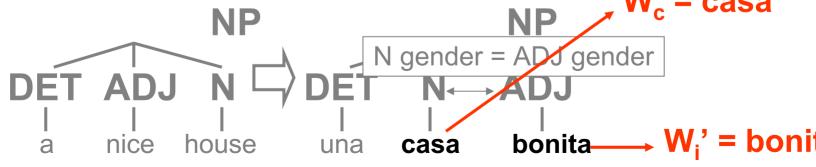
$$\overrightarrow{TL_m} = (\mathbf{W_1}, \dots \mathbf{W_i}, \dots, \mathbf{W_n})$$

$$W_i$$
' = correction

$$W_c$$
 = clue word

$$\overrightarrow{TL'_m} = (\mathbf{W}_1, \dots, \mathbf{W}_i, \dots, \mathbf{W}_c, \dots, \mathbf{W}'_n)$$







Triggering Feature Detection

Comparison at the feature level to detect triggering feature(s)

```
\rightarrow Delta function: \delta(W_i, W_i')
```

```
Examples: gen = masc gen = fem
\delta(bohito,bohita) = \{gender\}
\delta(comiamos,comia) = \{person,number\}
\delta(mujer,guitarra) = \{\emptyset\}
```

If δ set is empty, need to postulate a new binary feature ($feat_i = \{+,-\}$)

Error Correction Elicitation

Finding Triggering Features

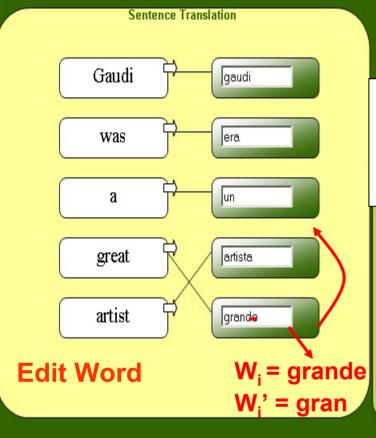
Blame Assignment

Error Correction Elicitation

Finding Triggering Features

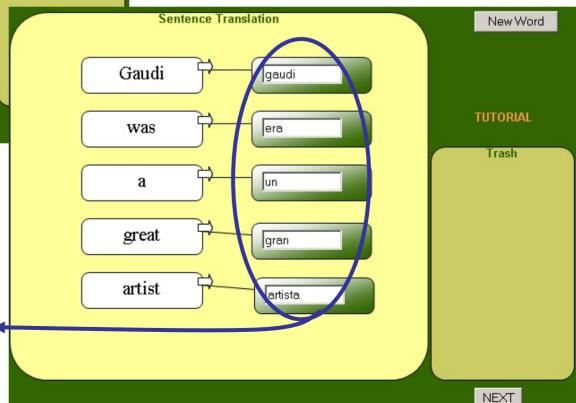
Blame Assignment





1. Error Correction Elicitation

Change Word Order



SL: Gaudí was a great artist

TL: Gaudí era un artista grande

CTL: Gaudí era un gran artista

Error Correction Elicitation

Finding Triggering Feature

Blame Assignment

Rule Refinement

1. Edit: $W_i = grande W_i' = gran$

2. Change Word Order: artista gran → gran artista

Error Correction Elicitation

1. Edit:

 W_i = grande W_i ' = gran

2. Change Word Order:

artista gran → gran artista

Finding Triggering Features

Blame Assignment

Delta function difference at the feature level?

ADJ ADJ

 $[great] \rightarrow [grande]$ $[great] \rightarrow [gran]$

agr num = sg agr num = sg

agr gen = masc agr gen = masc

 δ (grande, gran) = \emptyset

Delta function difference at the feature level?

ADJ

[great] \rightarrow [grande] [great] \rightarrow [gran]

agr num = sg agr num = sg

agr gen = masc agr gen = masc

 δ (grande, gran) = \emptyset

→ need to postulate a new binary feature: feat1

Delta function difference at the feature level?

ADJ

[great] \rightarrow [grande] [great] \rightarrow [gran]

agr num = sg agr num = sg

agr gen = masc agr gen = masc

$$\delta(\text{grande, gran}) = \emptyset$$

→ need to postulate a new binary feature: feat1

[feat1 = +] = [type = pre-nominal] [feat1 = -] = [type = post-nominal]

Delta function difference at the feature level?

ADJ ADJ

 $[great] \rightarrow [grande]$ $[great] \rightarrow [gran]$

agr num = sg agr num = sg

agr gen = masc agr gen = masc

 $\delta(\text{grande, gran}) = \emptyset$

new binary feature: feat1

REFINE

ADJ ADJ

 $[great] \rightarrow [grande]$ $[great] \rightarrow [gran]$

agr num = sg agr num = sg

agr gen = masc agr gen = masc

feat1 = - feat1 = +

Error Correction Elicitation

1. Edit:

 W_i = grande W_i ' = gran

2. Change Word Order: artista gran → gran artista

grande [feat1 = -]

gran [feat1 = +]

Finding Triggering Features

Blame Assignment

Error Correction Elicitation

Finding Triggering Features

1. Edit:

 W_i = grande W_i ' = gran

2. Change Word Order: artista gran → gran artista

grande [feat1 = -]

gran [feat1 = +]

Blame Assignment

3. Blame Assignment

(from transfer and generation tree)

```
tree: <( S,1 (NP,2 (N,5:1 "GAUDI") )
```

```
Grammar (VP,3 (VB,2 (AUX,17:2 "ERA"))

S,1
...
NP,1
(N,4:5 "ARTISTA")
(ADJ,5:4 "GRANDE")))>
```

Error Correction Elicitation

1. Edit:

 W_i = grande W_i ' = gran

2. Change Word Order: artista gran → gran artista

Finding Triggering Features

grande [feat1 = -]

gran [feat1 = +]

Blame Assignment

NP,8

 $N ADJ \rightarrow ADJ N$

Error Correction Elicitation

1. Edit:

 W_i = grande W_i ' = gran

2. Change Word Order: artista gran → gran artista

Finding Triggering Features

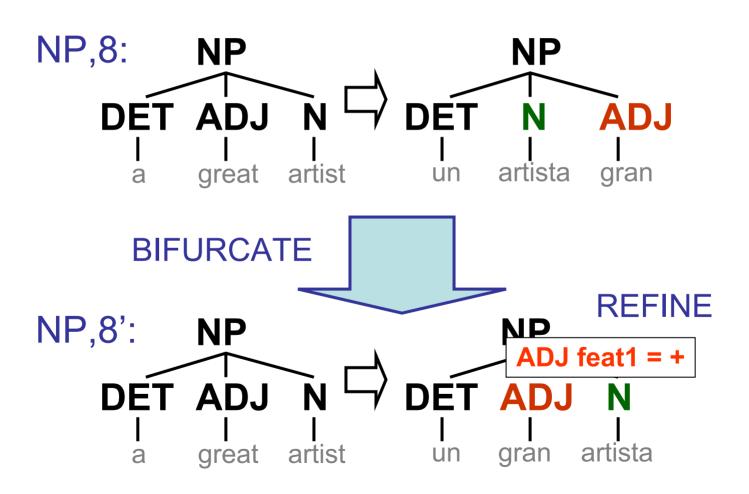
grande [feat1 = -]

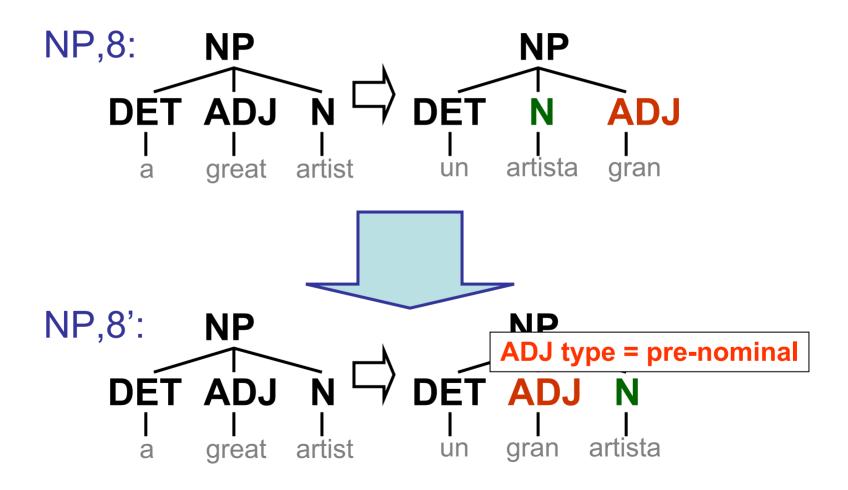
gran [feat1 = +]

Blame Assignment

NP,8

 $NADJ \rightarrow ADJ N$





Error Correction Elicitation

1. Edit:

 $W_i = grande W_i' = gran$

2. Change Word Order: artista gran → gran artista

Finding Triggering Features

grande [feat1 = -]

gran [feat1 = +]

Blame Assignment

NP,8

 $(N ADJ \rightarrow ADJ N)$

Rule Refinement

NP,8 ADJ N → N ADJ

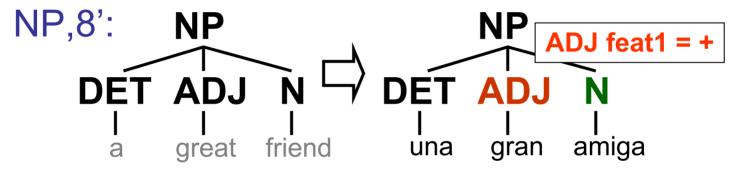
NP,8' ADJ N \rightarrow ADJ N [ADJ feat 1 = +]

Correct Translation Output

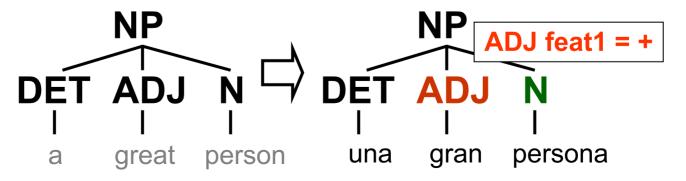
Gaudi era un artista grande Gaudi era un gran artista *Gaudi era un grande artista

Generalization Power: abstract feature (feat_i)

Irina is a great friend → *Irina es una gran amiga* (instead of **Irina es una amiga grande*)



Juan is a great person → Juan es una gran persona (instead of *Juan es una persona grande)





Generalization Power++

```
When triggering feature already exists in the grammar/lexicon (POS, gender number etc.)

| ADJ gender = N gender | N gen
```

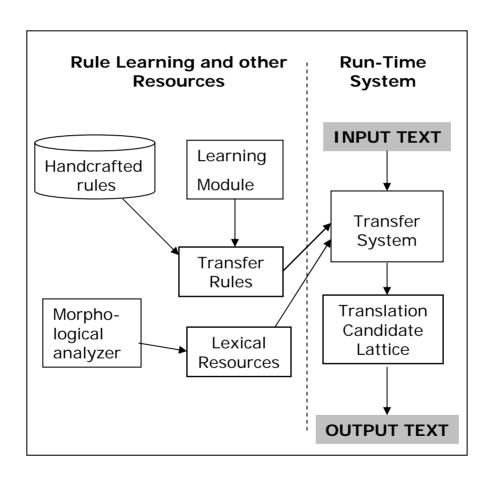
Refinements generalize to all lexical entries that have that feature (gender): gender = fem

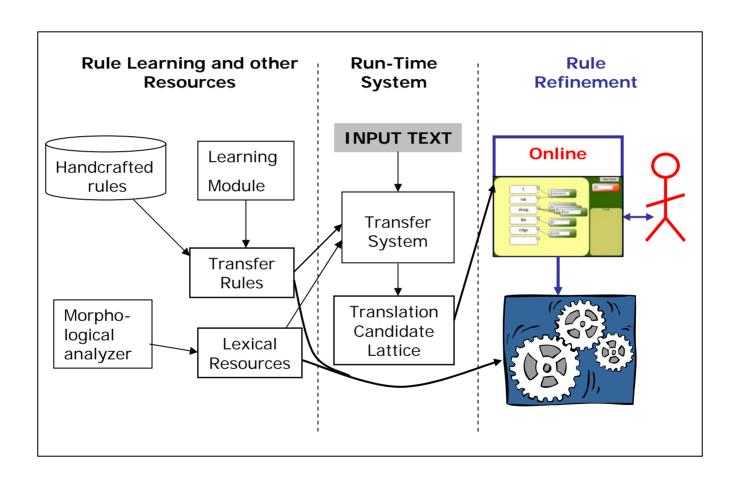
The yellow houses are his → las casas amarillas son suyas (before: *las casas amarillos son suyas)

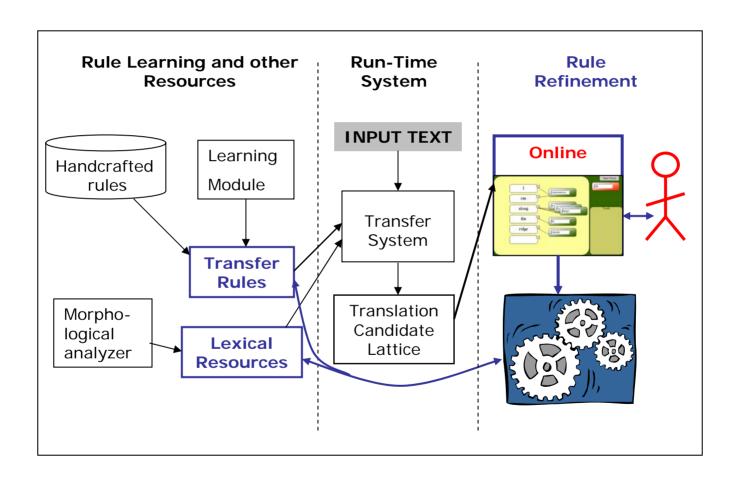
```
gender = fem
```

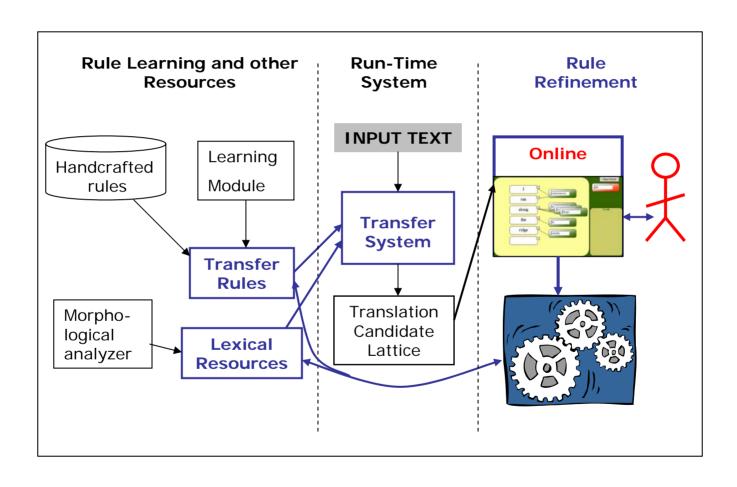
We need to go to a dark cave → tenemos que ir a una cueva oscura

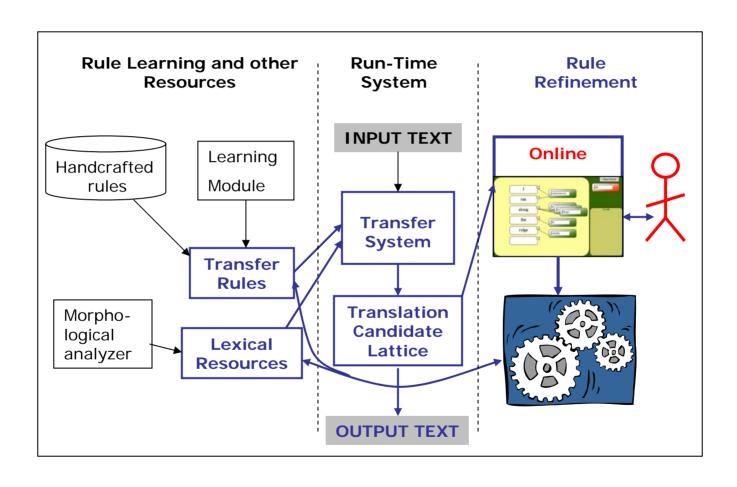
(before: *cueva oscuro)











TCTool can help improve

Rule-based MT (grammar, lexicon, LM)

EBMT (examples, lexicon, alignments)

Statistical MT (lexicon, and alignments)

Work in Progress

- We have identified 10 different types of refinement operations, integrating last 2 into the system.
- Handling incorrect Correction Instances
 - Have multiple users correct the same set of sentences
 - → filter out noise (threshold: 90% users agree)
- User study with multiple users
 - → evaluate improvement after refinements.

Contributions so far



New Framework to improve MT quality: an expandable set of rule refinement operations



An efficient online GUI to display translations and alignments and solicit pinpoint fixes from *non-expert* bilingual users.

Future work

Explore other ways to make the interaction with users more fun



Games with a purpose
 [Von Ahn and Blum, 2004 & 2006]



Future work

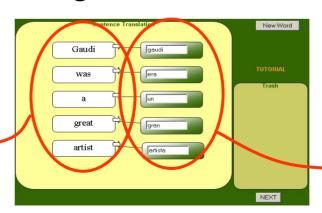
Explore other ways to make the interaction with users more fun



Games with a purpose
 [Von Ahn and Blum, 2004 & 2006]

Language Learning







Gracias!

