First International Conference on Research in Multilingualism: Innovation and New Challenges Universidad de Oviedo, 14-16/03/2018

Google Translate and DeepL: Breaking Taboos in Translator Training

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- 1. Motivation
- Theoretical Framework of Machine Translation (MT)
- 3. Goals and Methodology
- 4. Discussion of Results
- 5. Conclusions and Future Prospects

- To use or not to use MT in translator training?
- Role of machine translation (MT): What has changed?
- Research focus:
  - Potential and limits of MT and students` handling
  - Didactic use

## Theoretical Framework of MT

RBMT: Rule Based Machine Translation (PB)SMT: (Phrase Based) Statistical Machine Translation EBMT: Example Based Machine Translation NMT: Neural Machine Translation

## Goals and Methodology

Potential and limits of MT and students` handling for post-editing:

- Identify errors
- Classify error tendencies

Didactic use:

- Knowledge about grammatical features to be emphasized
- Improve metacognitive competence

Goals and Methodology

## **Post-editing:**

Defined by ISO 18587:2017 as to "edit and correct machine translation output".

## Here:

"Light post-editing": to correct only as far as "to obtain a merely comprehensible text without any attempt to produce a product comparable to a product obtained by human translation".

Goals and Methodology

## Error categories:

- Pragmatics
- Grammar
- Lexicon/ terminology
- Equivalence
- Spelling
- Style
- Culture

Goals and Methodology

## **Observational study:**

- with 32 35 students from the 4th year of Translation Studies (German as a second language after approximately 225 hours)
- MT translation of five texts (126 313 words) from the financial section of a non-specialized publication from Spanish into German + light post-editing
- MT tool: Google Translate and DeepL

Goals and Methodology

- 45 90 minutes for MT + light post-editing
- access to online and paperback resources
- use of own translations, previously corrected in class

## Goals and Methodology

Google T.	Error Category (Number and Percentage) DESEMPLEADOS 52+ (Google Translate)										
	Prag Gr		Equ Spell		Lex/ Sty Term		Cult	Error rate	Total		
1	7	17	8	1	19	3	1				
	12,5 %	30,36 %	14,29 %	1,79 %	33,93 %	5,36 %	1,79 %				

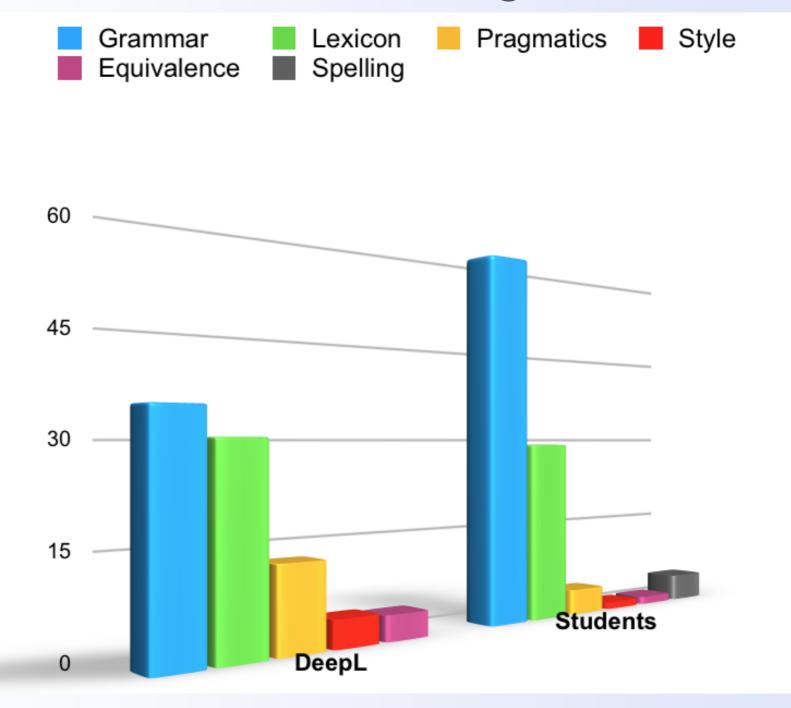
DeepL	Error Category (Number and Percentage)										
	DESEMPLEADOS + 52 AÑOS (DeepL)										
	Prag Gr Equ Spell Lex/ Sty Cuit Error Tota Term										
	3	8	4		7	1					
	13,04 %	34,78 %	17,39 %		30,43 %	4,35 %					

Discussion of Results

 Google Translate performance is below that of DeepL

## Discussion of Results

## 'Desempleados 52+': Error Categories



## Discussion of Results

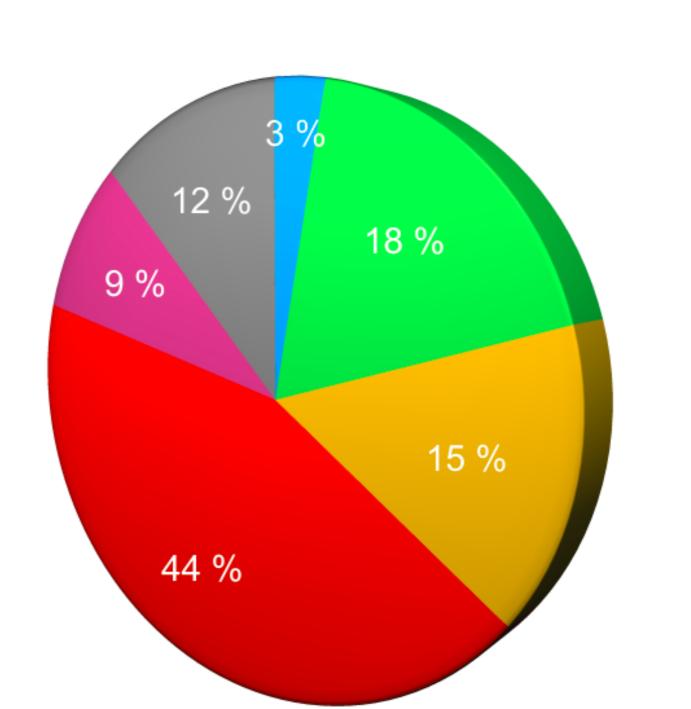
DeepL	Error Category (Number and Percentage)										
	DESEMPLEADOS + 52 AÑOS (DeepL)										
	Prag	Prag Gr Equ Spell Lex/ Sty Cult Error Tota Term Term									
	3	8	4		7	1					
	13,04 %	34,78 %	17,39 %		30,43 %	4,35 %					

Number of Students	Endreategory (Number and Fercentage)									Media of Errors
	Prag	Gr	Equ	Spell	Lex/ Term	Sty	Cult	Error rate	Total	
34	33	475	10	35	232	9			794	
	4,16 %	59,8 %	1,26 %	4,4 %	29,21 %	1,13 %				

## Discussion of Results

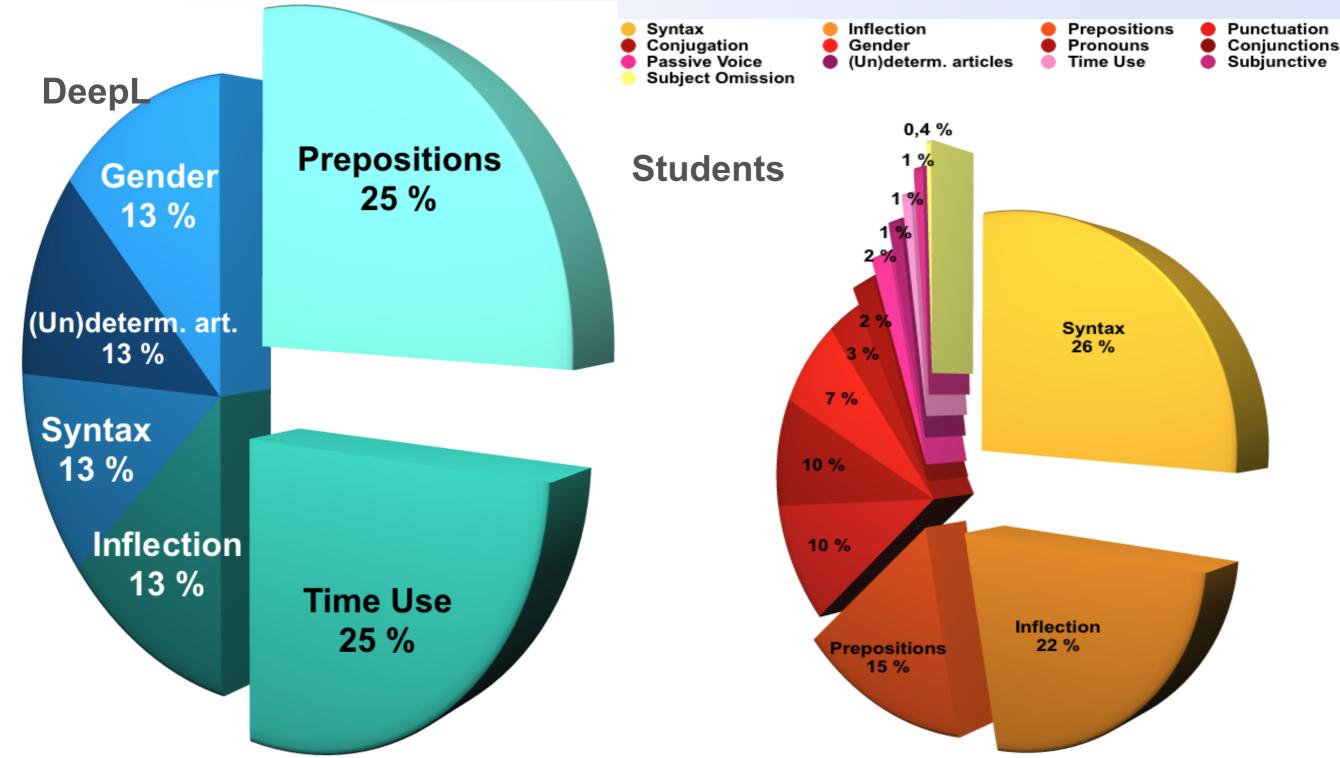
#### 🔵 0 - 6 🔵 7 - 13 😑 14 - 20 🛑 21 - 27 🛑 28 - 34 🌑 35 <

'Desempleados 52 Students' error dis within the group



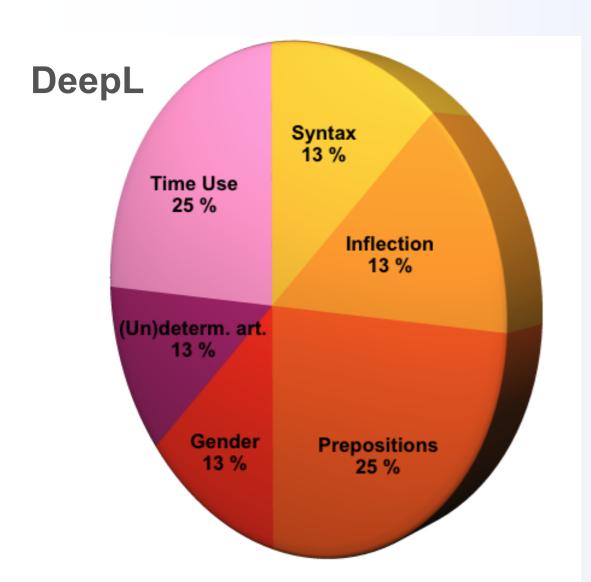
## Discussion of Results

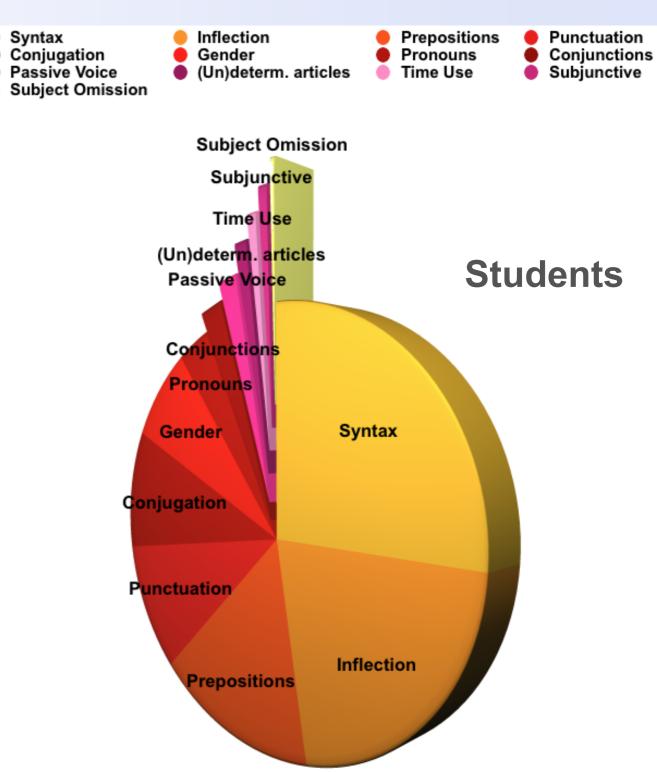
#### Distribution of grammatical errors in terms of error frequency:



## Discussion of Results

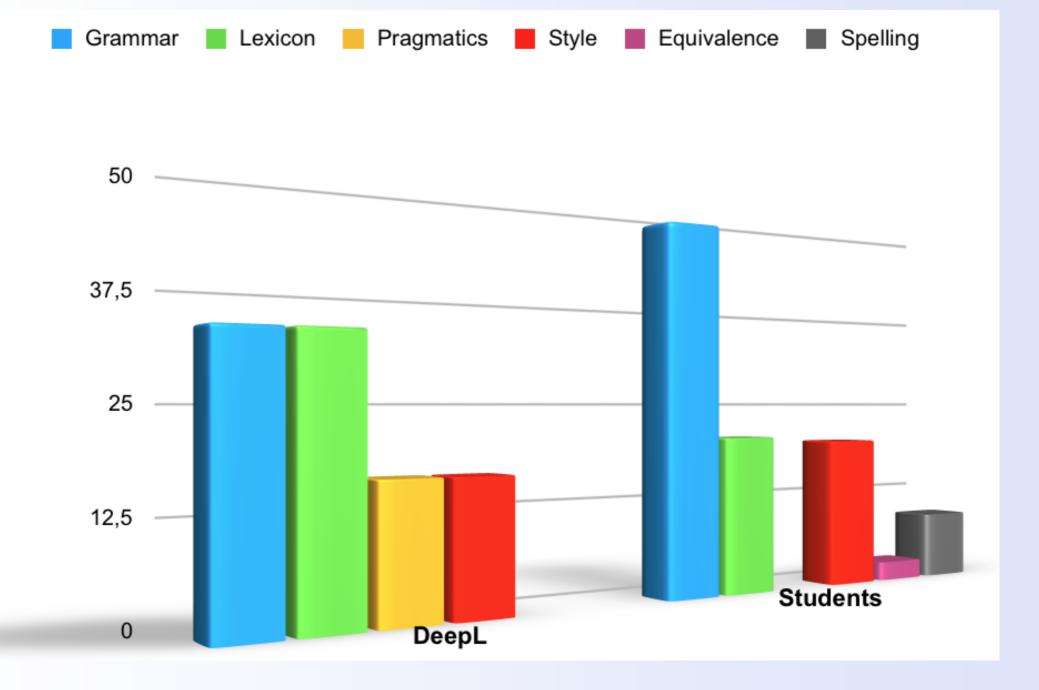
Distribution of grammatical errors in terms of error category:





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## 'Inflación subyacente': Error Categories



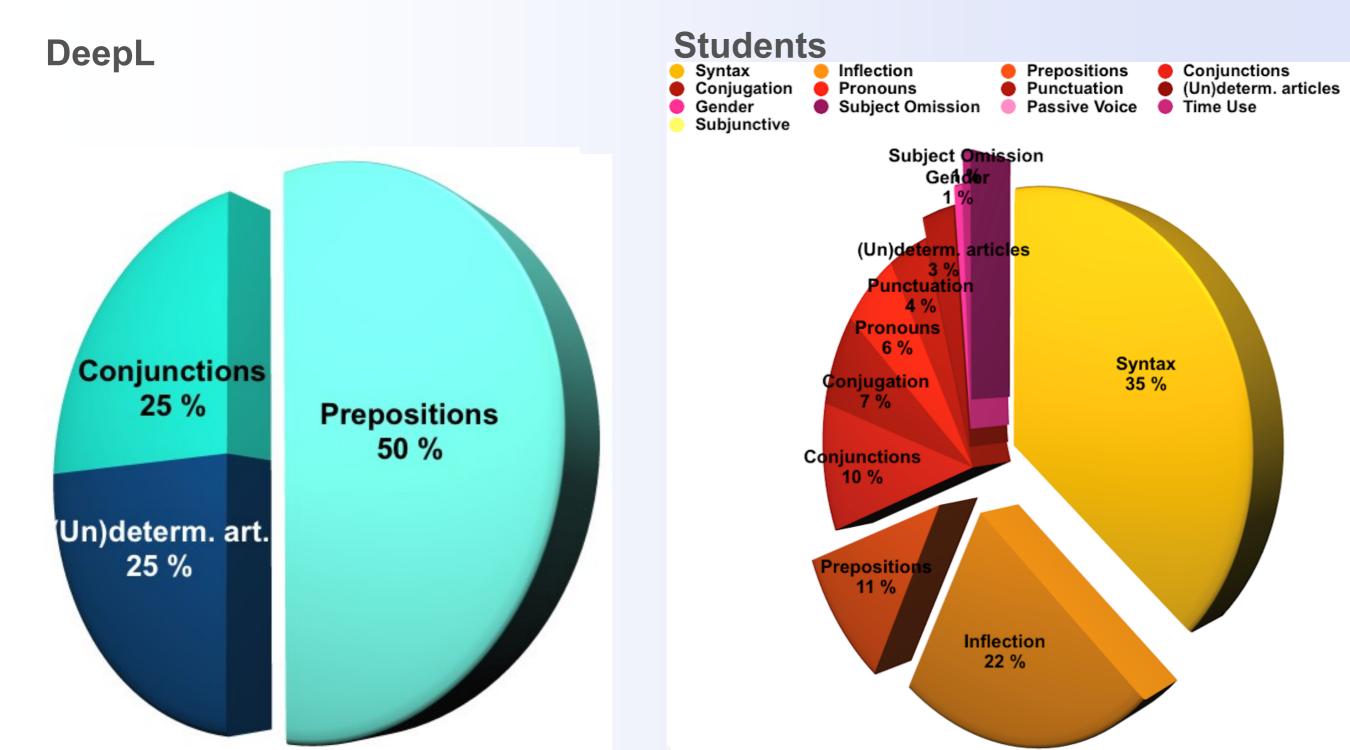
## **Discussion of Results**

	INFLACIÓN SUBYACENTE (DEEPL)										
	Prag	Gr	Equ	Spell	Lex/ Term	Sty	Cult	Error rate	Total		
1	2	4			4	2					
	16,7 %	33,4 %			33,4 %	16,7 %					

DeepL.	Error Category (Number and Percentage) INFLACIÓN SUBYACENTE (STUDENTS)									Media of Errors
	Prag	Gr	Equ	Spell	Lex/ Term	Sty	Cult	Error rate	Total	
31	0	136	7	25	58	56	0		282	
		48,2 %	2,5 %	8,9 %	20,6 %	19,9 %				

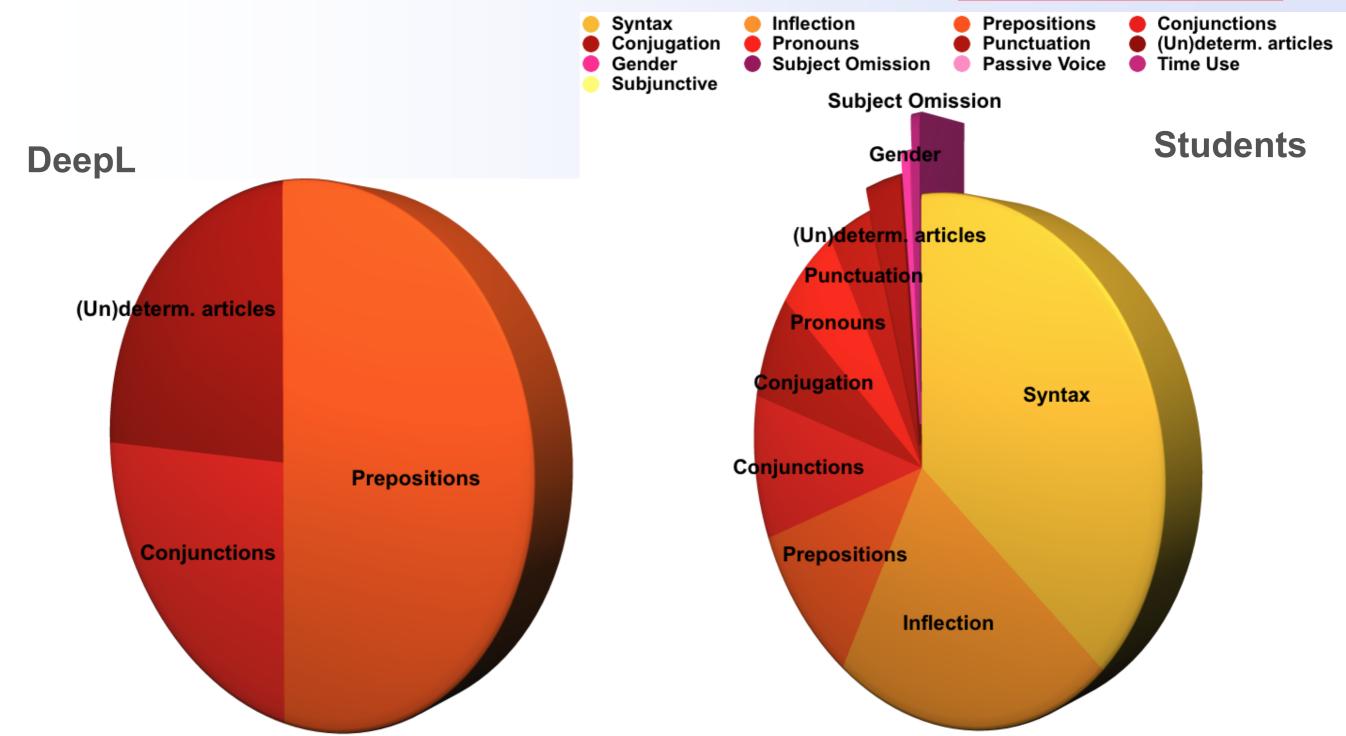
## Discussion of Results

Distribution of grammatical errors in terms of error frequency:



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### Distribution of grammatical errors in terms of error category:



## **Conclusions and Future Prospects**

- Post-editing machine translated texts can be useful if
  - target language is mastered
  - the quality of NMT-System is acceptable

## **Conclusions and Future Prospects**

- MT allows to detect which grammatical aspect needs to be deepened => didactic use
- The error tendencies we found in our research in 2006 were confirmed.
- Students must learn about possibilities and limits of MT, and how they work.
- Context is crucial for correct translation, and this is currently not being considered in MT.

## **Conclusions and Future Prospects**

 Professional translator profile is changing: Syllabus in translator training at universities should include more MT and PE

## **Conclusions and Future Prospects**

## **Future Prospects**

- Studies on comparing translation quality of same target texts with and without MT
- MT into the mother language
- Examples of how MT can be used to promote metacognition among translation trainees and to strengthen their grammar skills

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Some of the research presented was conducted in cooperation with:



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