Computer Aided Translation a User Study

Philipp Koehn

29 September 2009





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- user is tolerant of inferior quality
- focus of majority of research (GALE program, etc.)



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- high demands for quality
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FOCUS



Goal: Helping Human Translators

If you can't beat them, join them.



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• How can machine translation help human translators?



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If you can't beat them, join them.

- How can machine translation help human translators?
- First question: What do translators do?

Setup



- 10 students at the University of Edinburgh
 - half native French speakers
 - half native English speakers with advanced French

Setup

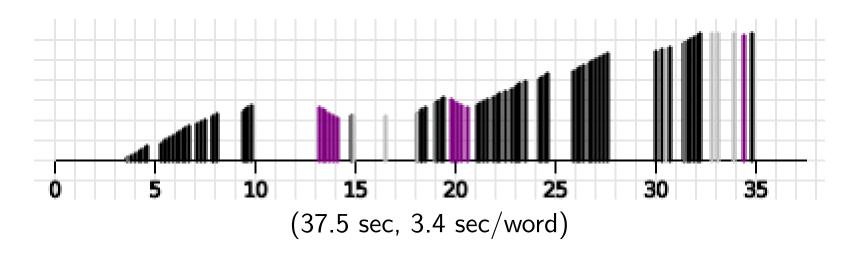


- 10 students at the University of Edinburgh
 - half native French speakers
 - half native English speakers with advanced French
- Each student translated
 - news stories
 - French-English
 - about 40 sentences
 - easy task: familiar content, no specialized terminology
- Keystroke log

Keystroke Log



Input: Au premier semestre, l'avionneur a livr 97 avions.Output: The manufacturer has delivered 97 planes during the first half.



black: keystroke, purple: deletion, grey: cursor move height: length of sentence



- We can observe
 - slow typing



- We can observe
 - slow typing
 - fast typing



- We can observe
 - slow typing
 - fast typing
 - pauses



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 - pauses
- Pauses
 - beginning pause: reading the input sentence
 - final pause: reviewing the translation



- We can observe
 - slow typing
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 - pauses
- Pauses
 - beginning pause: reading the input sentence
 - final pause: reviewing the translation
 - short pauses (2-6 seconds): hesitation
 - medium pauses (6-60 seconds): problem solving
 - big pauses (>60 seconds): serious problem



		Pauses					
User	total	initial	final	short	medium	big	keystroke
L2a	3.3s	0.1s	0.1s	0.2s	1.0s	0.1s	1.8s
L2b	7.7s	1.3s	0.1s	0.3s	1.8s	1.9s	2.3s
L2c	3.9s	0.2s	0.2s	0.3s	0.7s	_	2.5s
L2d	2.8s	0.2s	0.0s	0.2s	0.4s	0.1s	1.8s
L2e	5.2s	0.3s	0.0s	0.3s	1.9s	0.5s	2.2s
L1a	5.7s	0.5s	0.1s	0.3s	1.8s	0.7s	2.2s
L1b	3.2s	0.1s	0.1s	0.2s	0.4s	0.1s	2.2s
L1c	5.8s	0.3s	0.2s	0.5s	1.5s	0.3s	3.1s
L1d	3.4s	0.7s	0.1s	0.3s	0.6s	-	1.8s
L1e	2.8s	0.3s	0.2s	0.2s	0.3s	0.1s	1.9s

L2 = native French, L1 = native English



		not muc	h time	Pauses	5		
Use	r total	initial	final	short	medium	big	keystroke
L2a	3.3s	0.1s	0.1s	0.2s	1.0s	0.1s	1.8s
L2b	7.7s	1.3s	0.1s	0.3s	1.8s	1.9s	2.3s
L2c	3.9s	0.2s	0.2s	0.3s	0.7s	-	2.5s
L2d	2.8s	0.2s	0.0s	0.2s	0.4s	0.1s	1.8s
L2e	5.2s	0.3s	0.0s	0.3s	1.9s	0.5s	2.2s
L1a	5.7s	0.5s	0.1s	0.3s	1.8s	0.7s	2.2s
L1b	3.2s	0.1s	0.1s	0.2s	0.4s	0.1s	2.2s
L1c	5.8s	0.3s	0.2s	0.5s	1.5s	0.3s	3.1s
L1d	3.4s	0.7s	0.1s	0.3s	0.6s	-	1.8s
L1e	2.8s	0.3s	0.2s	0.2s	0.3s	0.1s	1.9s

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		not much time		Pauses	5	similar	
User	total	initial	final	short	medium	big	keystroke
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L2b	7.7s	1.3s	0.1s	0.3s	1.8s	1.9s	2.3s
L2c	3.9s	0.2s	0.2s	0.3s	0.7s	-	2.5s
L2d	2.8s	0.2s	0.0s	0.2s	0.4s	0.1s	1.8s
L2e	5.2s	0.3s	0.0s	0.3s	1.9s	0.5s	2.2s
L1a	5.7s	0.5s	0.1s	0.3s	1.8s	0.7s	2.2s
L1b	3.2s	0.1s	0.1s	0.2s	0.4s	0.1s	2.2s
L1c	5.8s	0.3s	0.2s	0.5s	1.5s	0.3s	3.1s
L1d	3.4s	0.7s	0.1s	0.3s	0.6s	-	1.8s
L1e	2.8s	0.3s	0.2s	0.2s	0.3s	0.1s	1.9s

L2 = native French, L1 = native English



		not muc	h time	Pauses	differen	ces	similar
User	total	initial	final	short	medium	big	keystroke
L2a	3.3s	0.1s	0.1s	0.2s	1.0s	0.1s	1.8s
L2b	7.7s	1.3s	0.1s	0.3s	1.8s	1.9s	2.3s
L2c	3.9s	0.2s	0.2s	0.3s	0.7s	-	2.5s
L2d	2.8s	0.2s	0.0s	0.2s	0.4s	0.1s	1.8s
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Pauses Reconsidered



- Our classification of pauses is arbitrary (2-6sec, 6-60sec, >60sec)
- Extreme view: all you see is pauses
 - keystrokes take no observable time
 - all you see is pauses between action points

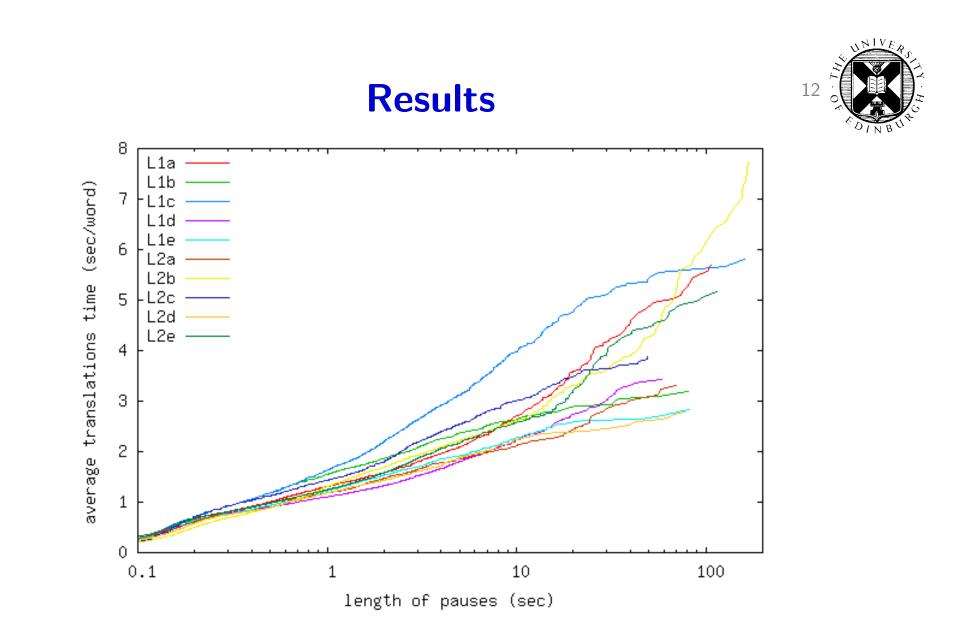
Pauses Reconsidered



- Our classification of pauses is arbitrary (2-6sec, 6-60sec, >60sec)
- Extreme view: all you see is pauses
 - keystrokes take no observable time
 - all you see is pauses between action points
- Visualizing range of pauses:

time t spent in pauses $p \in P$ up to a certain length l

$$sum(t) = \frac{1}{Z} \sum_{p \in P, l(p) \le t} l(p)$$



Related Work: Tools used by Translators¹³

- Translators often use standard text editors and additional tools
- Bilingual dictionary
- Spell checker, grammar checker
- Monolingual concordancer
- Terminology database
- Web search to establish and verify meaning of terms

Translation Memory



• Source:

This feature is available for free in the QX 3400.

• Fuzzy match in translation memory:

This feature is available for free in the QX 3200. Diese Funktion ist kostenlos im Modell QX 3200 verfügbar.

• Translator inspects the fuzzy match and uses it in her translation.



Bilingual Concordancer

xamples	Windkraft (noun, feminine) (also: Windenergie)	wind power (noun)	V
	Zum Vergleich: Windkraft schafft fast sieben Mal	By way of comparison, wind power generates almost	
	mehr. ⇒ German: www.goethe.de/wis/umw/thm/ntr/de92305.htm	seven times as much. ⇒ English: www.goethe.de/wis/umw/thm/ntr/en92305.htm	
	Einführung von Windcube, einer neuen Generation von Wind Lidar für Windkraft. E→ German: www.husumwindenergy.com/index.php?LhowUid]=1177	Introducing Windcube, a new generation of wind Lidar for wind power. G> English: www.husumwindenergy.com/index.php?LhowUid]=1177	
	Windkraft ist eine etablierte, wettbewerbsfähige Technologie mit hoher Zuverlässigkeit ▷→ German: www.powergeneration.siemens.de/abouns- services/	Wind power is an established, competitive technology with high reliability □→ English: www.powergeneration.siemens.com/abons- services/	
amples	Windkraft (noun, feminine) (also: Windenergie)	wind energy (noun)	V
	Je mehr aber klimapolitische Sonntagsreden von der Politik auch in Taten umgesetzt werden, desto höher steigt dieser Preis und desto wettbewerbsfähiger	But as the focus of the climate change issue shifts increasingly from policy to action, this price will increase and cleaner energy sources like wind will become more compatible	
	werden saubere Energien wie die Windkraft. ⊡→ German: emagazine.credit-suisse.com/app /art4382 <=DE	become more competitive. □→ English: emagazine.credit-suisse.com/app /art4382 <=en	
-	German: emagazine.credit-suisse.com/app	G→ English: emagazine.credit-suisse.com/app	

show translations in context (www.linguee.com)

Our Types of Assistance



- Sentence completion
 - tool suggests how to complete the translation
 - one phrase at a time



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- Translation options
 - most likely translations for each word and phrase
 - ordered and color-highlighted by probability



Our Types of Assistance

- Sentence completion
 - tool suggests how to complete the translation
 - one phrase at a time
- Translation options
 - most likely translations for each word and phrase
 - ordered and color-highlighted by probability
- Postediting machine translation
 - start with machine translation output
 - user edits, tool shows changes

Technical Notes



- Online at http://www.caitra.org/
- User uploads source text, translates one sentence at a time
- Implementation
 - AJAX Web 2.0 using Ruby on Rails, mySQL
 - Back end: Moses machine translation system



Predicting Sentence Completion

[1] Paul Newman le magnifique >>

Paul			
enter	Newman	×	

• Tool makes a suggestion how to continue (in red)



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- Tool makes a suggestion how to continue (in red)
- User can accept it (by pressing TAB), or type in her own translation



Predicting Sentence Completion

[1] Paul Newman le magnifique >>

Paul			
enter	Newman	×	

- Tool makes a suggestion how to continue (in red)
- User can accept it (by pressing TAB), or type in her own translation
- Same idea as TransType, with minor modifications
 - show only short text chunks, not full sentence completion
 - show only one suggestion, not alternatives



How does it work?

• Uses search graph of SMT decoding

How does it work?



- Uses search graph of SMT decoding
- Matches partial user translation against search graph, by optimizing
 - 1. minimal string edit distance between path in graph and user translation
 - 2. best full path probability, including best completion to end

How does it work?



- Uses search graph of SMT decoding
- Matches partial user translation against search graph, by optimizing
 - 1. minimal string edit distance between path in graph and user translation
 - 2. best full path probability, including best completion to end
- Technical notes
 - search graph is pre-computed and stored in database
 - matching is done server-side, typically takes less than 1 second
 - completion path is returned to client (web brower)



Translation Options

Paul	Newman	le	magnifique
Paul	Newman	the	wonderful
Mr	Newman,	the	magnificent
Mr Paul	Newman here	th	e wonderful
as Paul	Committee		beautiful
another	Newman , who speaks		magnificent
with Paul		tł	ne splendid
, Paul		th	ne excellent
of Paul		tł	ne beautiful
work of Paul			it
the words of Paul			great

- For each word and phrases: suggested translations
- Ranked (and color-highlighted) by probability
- $\bullet~$ User may click on suggestion \rightarrow appended to text box



• Uses phrase translation table of SMT system



Translation Options - How does it work?²¹

- Uses phrase translation table of SMT system
- Translation score: future cost estimate
 - conditional probabilities $\phi(\bar{e}|\bar{f}),\,\phi(\bar{f}|\bar{e})$
 - lexical probabilities ${\rm lex}(\bar{e}|\bar{f}),~{\rm lex}(\bar{f}|\bar{e})$
 - word count feature
 - language model estimate



Translation Options - How does it work?²¹

- Uses phrase translation table of SMT system
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 - conditional probabilities $\phi(\bar{e}|\bar{f}),\,\phi(\bar{f}|\bar{e})$
 - lexical probabilities $lex(\bar{e}|\bar{f})$, $lex(\bar{f}|\bar{e})$
 - word count feature
 - language model estimate
- Ranking of shorter vs. longer phrases by including outside future cost estimate

Translation Tool translate - Mozilla Firefox 🗕 🗖										
<u>F</u> ile <u>E</u> dit <u>V</u> iew Hi <u>s</u> tory <u>B</u> ookmarks <u>T</u> ools <u>H</u> elp	<u>ن</u>									
 	æ/563 ▼ ▶ 💽 Google 🔍									
🗁 Status 🖆 Wiki 🖂 Mail MgMail 🗀 EdU 🗀 News										
Translation Tool pkoehn logout										
Sentence 2 of 20 [1] [2] [4] [6] [8] [11] [13] [16] [19]										
 [1] Spitzen von Hamburger CDU und Grünen öffnen Weg zu Koalitionsverhandlungen [2] Das erste schwarz-grüne Bündnis auf Landesebene rückt näher. Die Spitzen von CDU und Grünen in Hamburg halten ihre Differenzen für überwindbar. [3] In einer Sondierungsrunde beschlossen sie, in den Parteigremien über den Start von Koalitionsverhandlungen zu beraten. [4] Hamburg - Sechs Stunden sprachen sie miteinander. [5] Dann verkündeten CDU-Chef Michael Freytag und Grünen-Chefin Anja Hajduk, das Trennende zwischen den Parteien sei überbrückbar. 										
<< [2] Das erste schwarz-grüne Bündnis auf Landesebene rückt näher: Die S halten ihre Differenzen für überwindbar. >>	itzen von CDU und Grünen in Hamburg									
enter the first										
das erste schwarz @-@ grüne Bündnis auf Landesebene rückt	näher : die Spitzen									
the first black @-@ green alliance in favour of is appro										
the first black @-@ green the alliance in favour appro										
for the first black Green Alliance on national we are co	ning to . at the top									

cooperation in

the Greens NATO seek to

Belarus approaches

we closer

the top

the this

in black and white @-@ green

this

the first of the black



Postediting Machine Translation

<< [2] L'inoubliable interprète de "Butch Cassidy et le Kid" est mort des suites d'un cancer, à l'âge de 83 ans, dans sa maison du Connecticut. >> The unforgettable interpreter actor of "Butch Cassidy and the Sundance Kid " died as a result of cancer 7 at the age of 83 years 7 in his house in Connecticut . (9 edits)

The unforgettable actor of "Butch <u>Cassidy</u> and the <u>Sundance</u> Kid" died as a result of cancer at the age of 83 in his house in Connecticut.

- Textbox is initially filled with machine translation
- User edits translation
- String edit distance to machine translation is shown (blue background)

Evaluation



- Recall setup
 - 10 students, half native French, half native English
 - each student translated French-English news stories
 - about 40 sentences for each condition of assistance

Evaluation



- Recall setup
 - 10 students, half native French, half native English
 - each student translated French-English news stories
 - about 40 sentences for each condition of assistance
- Five different conditions
 - unassisted
 - prediction (sentence completion)
 - options
 - predictions and options
 - post-editing

Quality



- We want faster translators, but not worse
- Assessment of translation quality
 - show translations to bilingual judges, with source
 - judgment: fully correct? yes/no

Indicate whether each user's input represents a fully fluent and meaning-equivalent translation of the source. The source is shown with context, the actual sentence is bold.

Quality



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- Average score: 50% correct lower than expected
 - judges seemed to be too harsh
 - when given several translations, tendency to judge half as bad



Example of Quality Judgments

Src.	Sans se démonter, il s'est montré concis et précis.	
MT	Without dismantle, it has been concise and accurate.	
1/3	Without fail, he has been concise and accurate. (Predict	ion+Options, L1a)
<mark>4/0</mark>	Without getting flustered, he showed himself to be concise and precise	. (Unassisted, L1b)
<mark>4/0</mark>	Without falling apart, he has shown himself to be concise and accura	te. (Postedit, L1c)
1/ <mark>3</mark>	Unswayable, he has shown himself to be concise and to the point.	(Options, L1d)
<mark>0/4</mark>	Without showing off, he showed himself to be concise and precise.	(Prediction, L1e)
1/ <mark>3</mark>	Without dismantling himself, he presented himself consistent and pre-	ecise.
	(Predict	ion+Options, L2a)
2/2	He showed himself concise and precise.	(Unassisted, L2b)
3/1	Nothing daunted, he has been concise and accurate.	(Postedit, L2c)
3/1	Without losing face, he remained focused and specific.	(Options, L2d)
3/1	Without becoming flustered, he showed himself concise and precise.	(Prediction, L2e)



Faster and Better

Assistance	Speed	Quality
Unassisted	4.4s/word	47% correct
Postedit	2.7s (-1.7s)	55% (+8%)
Options	3.7s (-0.7s)	51% (+4%)
Prediction	3.2s (-1.2s)	54% (+7%)
Prediction+Options	3.3s (-1.1s)	53% (+6%)

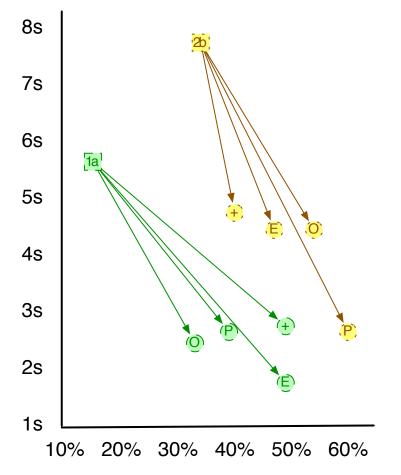


Faster and Better, Mostly

User	Unassisted	Pos	tedit	Ор	tions	Prec	liction	Predict	on+Options
L2a	3.3sec/word	1.2s	-2.2s	2.3s	-1.0s	1.1s	-2.2s	2.4s	-0.9s
	23% correct	39%	+16%)	45%	+22%	30%	+7%)	44%	+21%
L2b	7.7sec/word	4.5s	-3.2s)	4.5s	-3.3s	2.7s	-5.1s	4.8 s	-3.0s
	35% correct	48%	+13%	55%	+20%	61%	+26%	41%	+6%
L2c	3.9sec/word	1.9s	-2.0s	3.8s	-0.1s	3.1s	-0.8s	2.5s	-1.4s
	50% correct	61%	+11%	54%	+4%	64%	+14%	61%	+11%
L2d	2.8sec/word	2.0s	-0.7s	2.9s	(+0.1s)	2.4s	(-0.4s)	1.8s	-1.0s
	38% correct	46%	+8%	59%	(+21%)	37%	(-1%)	45%	+7%
L2e	5.2sec/word	3.9s	-1.3s	4.9s	(-0.2s)	3.5s	-1.7s	4.6s	(-0.5s)
	58% correct	64%	+6%	56%	(-2%)	62%	+4%	56%	(-2%)
L1a	5.7sec/word	1.8s	-3.9s	2.5s	-3.2s	2.7s	-3.0s	2.8s	-2.9s
	16% correct	50%	+34%	34%	+18%	40%	+24%	50%	+34%
L1b	3.2sec/word	2.8s	(-0.4s)	3.5s	+0.3s	6.0s	+2.8s	4.6s	+1.4s
	64% correct	56%	(-8%)	60%	-4%	61%	-3%	57%	-7%
L1c	5.8sec/word	2.9s	-3.0s	4.6s	(-1.2s)	4.1s	-1.7s	2.7s	-3.1s
	52% correct	53%	+1%	37%	(-15%)	59%	+7%	53%	+1%
L1d	3.4sec/word	3.1s	(-0.3s)	4.3s	(+0.9s)	3.8s	(+0.4s)	3.7s	(+0.3s)
	49% correct	49%	(+0%)	51%	(+2%)	53%	(+4%)	58%	(+9%)
L1e	2.8sec/word	2.6s	-0.2s	3.5s	+0.7s	2.8s	(-0.0s)	3.0s	+0.2s
	68% correct	79%	+11%	59%	-9%	64%	(-4%)	66%	-2%
avg.	4.4sec/word	2.7s	-1.7s	3.7s	-0.7s	3.2s	-1.2s	3.3s	-1.1s
	47% correct	55%	+8%	51%	+4%	54%	+7%	53%	+6%

Slow Users 1: Faster and Better

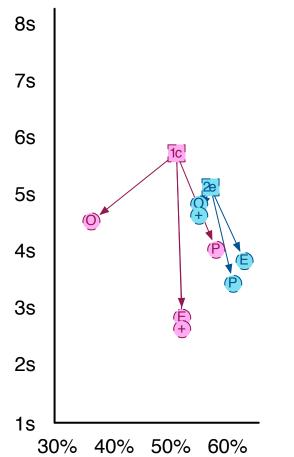




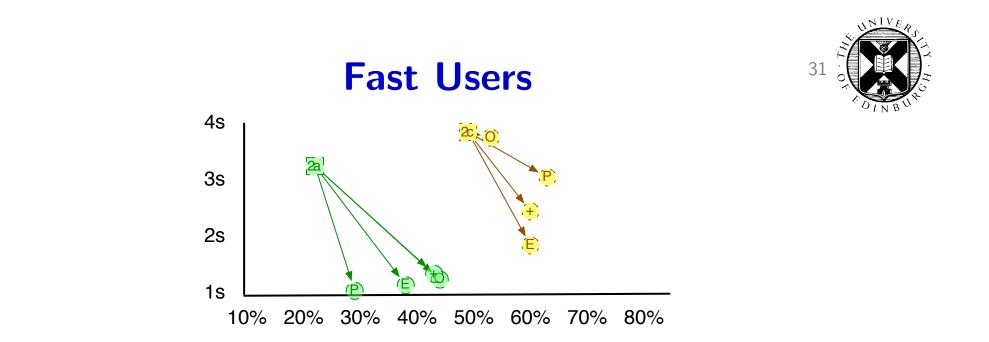
- Unassisted
 - more than 5 seconds per input word
 - very bad (35%, 16%)
- With assistance
 - much faster and better
 - reaching roughly average performance

Slow Users 2: Only Faster

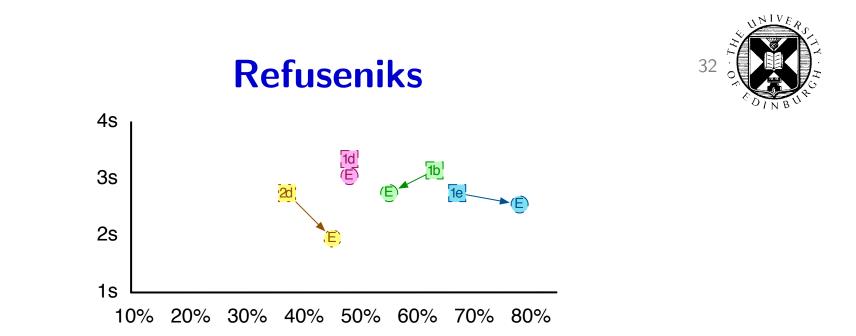




- Unassisted
 - more than 5 seconds per input word
 - average quality
- With assistance
 - faster and but not better



- Unassisted
 - fast: 3-4 seconds per input word
 - L2a is very bad (23%), L2c is average (50%)
- With assistance
 - faster and better
 - L2a closer to average (30-45%), L2c becomes very good (54-61%)

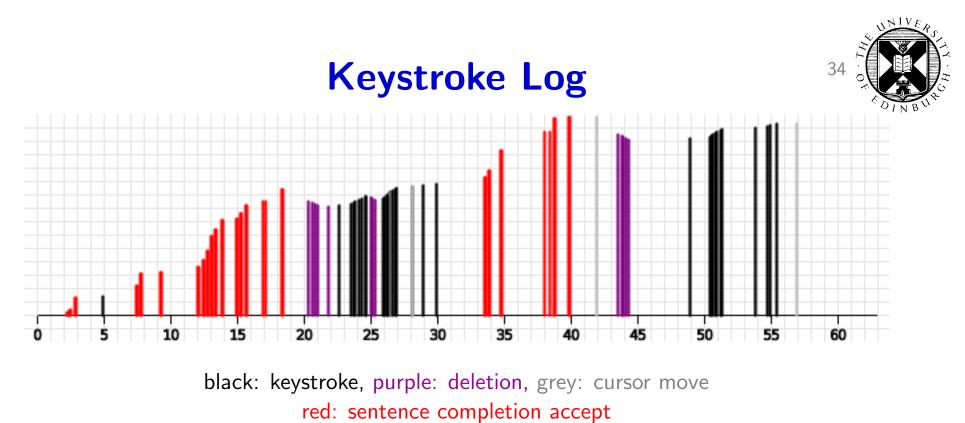


- Use the assistance sparingly or not at all, and see generally no gains
- The two best translators are in this group
- Postediting
 - mixed on quality (2 better, 1 worse, 1 same), but all faster
 - best translator (L1e, 68%) becomes much better (record 79%)

Further Analysis



- How does the assistance change translator behaviour?
- How do translators utilize assistance?
- How is the translation produced?



orange: click on translation option

Analysis: Segment into periods of activity: typing, tabbing, clicking, pauses

one second before and after a keystroke is part of typing interval



User: L2b	total	init-p	end-p	short-p	mid-p	big-p	key	click	tab
Unassisted	7.7s	1.3s	0.1s	0.3s	1.8s	1.9s	2.3s	-	-
Postedit	4.5s	1.5s	0.4s	0.1s	1.0s	0.4s	1.1s	-	-
Options	4.5s	0.6s	0.1s	0.4s	0.9s	0.7s	1.5s	0.4s	-
Prediction	2.7s	0.3s	0.3s	0.2s	0.7s	0.1s	0.6s	-	0.4s
Prediction+Options	4.8s	0.6s	0.4s	0.4s	1.3s	0.5s	0.9s	0.5s	0.2s



User: L2b	total	init-p	end-p	short-p	mid-p	big-p	key	click	tab
Unassisted	7.7s	1.3s	0.1s	0.3s	1.8s	1.9s	2.3s	-	-
Postedit	4.5s	1.5s	0.4s	0.1s	1.0s	0.4s	1.1s	-	-
Options	4.5s	0.6s	0.1s	0.4s	0.9s	0.7s	1.5s	0.4s	-
Prediction	2.7s	0.3s	0.3s	0.2s	0.7s	0.1s	0.6s	-	0.4s
Prediction+Options	4.8s	0.6s	0.4s	0.4s	1.3s	0.5s	0.9s	0.5s	0.2s

Slighly less time spent on typing



User: L2b	total	init-p	end-p	short-p	mid-p	big-p	key	click	tab
Unassisted	7.7s	1.3s	0.1s	0.3s	1.8s	1.9s	2.3s	-	-
Postedit	4.5s	1.5s	0.4s	0.1s	1.0s	0.4s	1.1s	-	-
Options	4.5s	0.6s	0.1s	0.4s	0.9s	0.7s	1.5s	0.4s	-
Prediction	2.7s	0.3s	0.3s	0.2s	0.7s	0.1s	0.6s	-	0.4s
Prediction+Options	4.8s	0.6s	0.4s	0.4s	1.3s	0.5s	0.9s	0.5s	0.2s

Less pausing Slighly less time spent on typing



User: L2b	total	init-p	end-p	short-p	mid-p	big-p	key	click	tab
Unassisted	7.7s	1.3s	0.1s	0.3s	1.8s	1.9s	2.3s	-	-
Postedit	4.5s	1.5s	0.4s	0.1s	1.0s	0.4s	1.1s	-	-
Options	4.5s	0.6s	0.1s	0.4s	0.9s	0.7s	1.5s	0.4s	-
Prediction	2.7s	0.3s	0.3s	0.2s	0.7s	0.1s	0.6s	-	0.4s
Prediction+Options	4.8s	0.6s	0.4s	0.4s	1.3s	0.5s	0.9s	0.5s	0.2s

Less pausing

Especially
less time
in big
pauses

Slighly less time spent on typing



User: L1e	total	init-p	end-p	short-p	mid-p	big-p	key	click	tab
Unassisted	2.8s	0.3s	0.2s	0.2s	0.3s	0.1s	1.9s	-	-
Postedit	2.6s	0.4s	0.3s	0.2s	1.0s	0.1s	0.7s	-	-
Options	3.5s	0.1s	0.3s	0.4s	0.6s	0.2s	1.7s	0.1s	-
Prediction	2.8s	0.1s	0.3s	0.3s	0.3s	-	1.4s	-	0.3s
Prediction+Options	3.0s	0.1s	0.3s	0.2s	0.5s	-	1.9s	-	-



User: L1e	total	init-p	end-p	short-p	mid-p	big-p	key	click	tab
Unassisted	2.8s	0.3s	0.2s	0.2s	0.3s	0.1s	1.9s	-	-
Postedit	2.6s	0.4s	0.3s	0.2s	1.0s	0.1s	0.7s	-	-
Options	3.5s	0.1s	0.3s	0.4s	0.6s	0.2s	1.7s	0.1s	-
Prediction	2.8s	0.1s	0.3s	0.3s	0.3s	-	1.4s	-	0.3s
Prediction+Options	3.0s	0.1s	0.3s	0.2s	0.5s	-	1.9s	-	-

Little time spent on assistance



User: L1e	total	init-p	end-p	short-p	mid-p	big-p	key	click	tab
Unassisted	2.8s	0.3s	0.2s	0.2s	0.3s	0.1s	1.9s	-	-
Postedit	2.6s	0.4s	0.3s	0.2s	1.0s	0.1s	0.7s	-	-
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Prediction	2.8s	0.1s	0.3s	0.3s	0.3s	-	1.4s	-	0.3s
Prediction+Options	3.0s	0.1s	0.3s	0.2s	0.5s	-	1.9s	_	-

Does not use both assistances, little overall change Little time spent on assistance



User: L1e	total	init-p	end-p	short-p	mid-p	big-p	key	click	tab
Unassisted	2.8s	0.3s	0.2s	0.2s	0.3s	0.1s	1.9s	-	-
Postedit	2.6s	0.4s	0.3s	0.2s	1.0s	0.1s	0.7s	-	-
Options	3.5s	0.1s	0.3s	0.4s	0.6s	0.2s	1.7s	0.1s	-
Prediction	2.8s	0.1s	0.3s	0.3s	0.3s	-	1.4s	-	0.3s
Prediction+Options	3.0s	0.1s	0.3s	0.2s	0.5s	-	1.9s	-	-

Does not use both assistances, little overall change Postediting: less typing (-1.2s) more medium pauses (+0.7s)

Little time spent on assistance



Origin of Characters: Native French L2b⁴³

User: L2b	key	click	tab	mt
Postedit	18%	-	-	81%
Options	59%	40%	-	-
Prediction	14%	-	85%	_
Prediction+Options	21%	44%	33%	-



Origin of Characters: Native French L2b⁴⁴

User: L2b	key	click	tab	mt
Postedit	18%	-	-	81%
Options	59%	40%	-	-
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Translation comes to large degree from assistance



Origin of Characters: Native English L1e⁴⁵

User: L1e	key	click	tab	mt
Postedit	20%	-	-	79%
Options	77%	22%	-	-
Prediction	61%	_	38%	-
Prediction+Options	100%	_	-	-

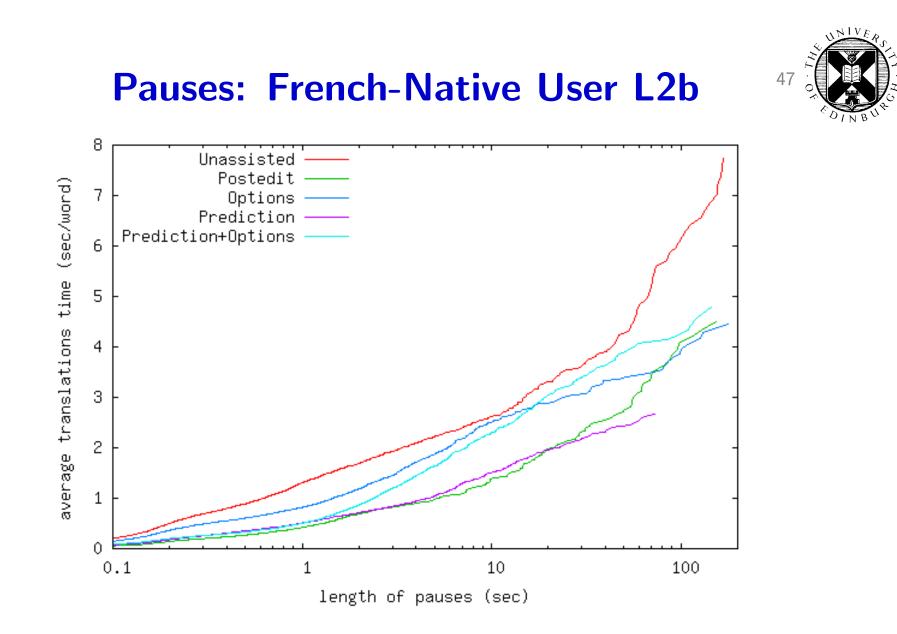


Origin of Characters: Native English L1e⁴⁶

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Postedit	20%	-	-	79%
Options	77%	22%	-	-
Prediction	61%	-	38%	-
Prediction+Options	100%	_	-	_

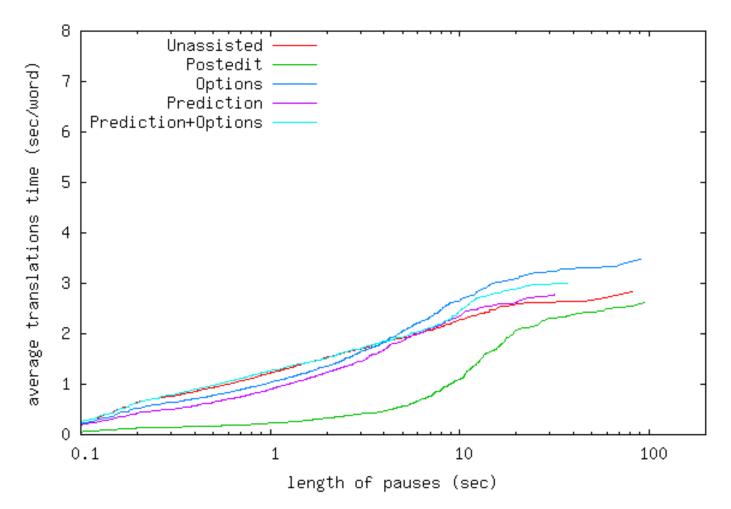
Although hardly any time spent on assistance, fair amount of characters produced by it

Philipp Koehn



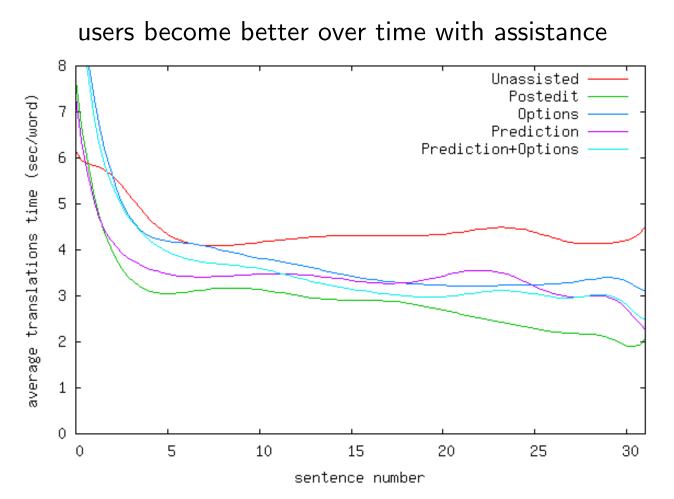


Pauses: English-Native User L1e



Learning Curve





User Feedback



- Q: In which of the five conditions did you think you were most accurate?
 - predictions+options: 5 users
 - options: 2 users
 - prediction: 1 user
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 - prediction+options: 4.6
 - prediction: 3.9
 - options: 3.7
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 - options: 3.7
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• Note: does not match empirical results



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 - average speed improvement from 4.4s/word to 2.7-3.7s/word
 - reduction of big pauses
 - reduction of typing effort in post-editing



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 - even good translators get better with postediting
- Some good translators ignored the assistance
- Fastest and (barely) best with postediting, but did not like it

Outlook: More analysis



• What do translators think about when they are pausing?

Outlook: More analysis



- What do translators think about when they are pausing?
- What are the hard problems?
 - unknown words
 - words without direct translation
 - syntactic re-arrangement

Outlook: More analysis



- What do translators think about when they are pausing?
- What are the hard problems?
 - unknown words
 - words without direct translation
 - syntactic re-arrangement
- What do translators change in post-editing?



- Different types of users
 - experienced professional translators
 - volunteer / amateur
 - no/little knowledge of source language



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 - we have training data: pairs of MT output, corrected sentence
 - \rightarrow supervised binary classification problem (correct/false)
- Visualizing word alignment (not only for post-editing)
- Better prediction model
 - different metrics to find best matching prefix
 - * BLEU, TER, ...
 - * word substitution scored by letter string edit distance
 - prediction as a machine learning multi-class classification problem
 - prediction based on parse forest from tree-based models



Outlook: Additional Types of Assistance⁵⁶

- Bilingual concordancer
 - show translations of words and phrases in context

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Outlook: Additional Types of Assistance⁵⁶

- Bilingual concordancer
 - show translations of words and phrases in context
- Monolingual concordancer / language model
 - show which expressions are more or less common
 - language model to highlight unusual transitions in output

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Outlook: Additional Types of Assistance⁵⁶

- Bilingual concordancer
 - show translations of words and phrases in context
- Monolingual concordancer / language model
 - show which expressions are more or less common
 - language model to highlight unusual transitions in output
- Translation memory
 - finding most similar sentence in parallel corpus
 - show translation, highlight difference, fill in difference with MT





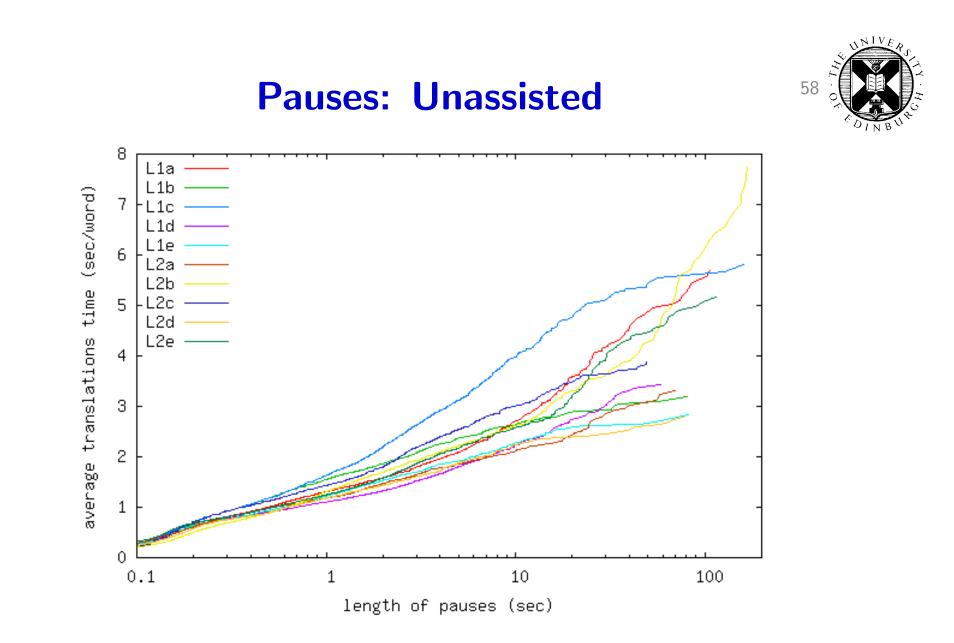
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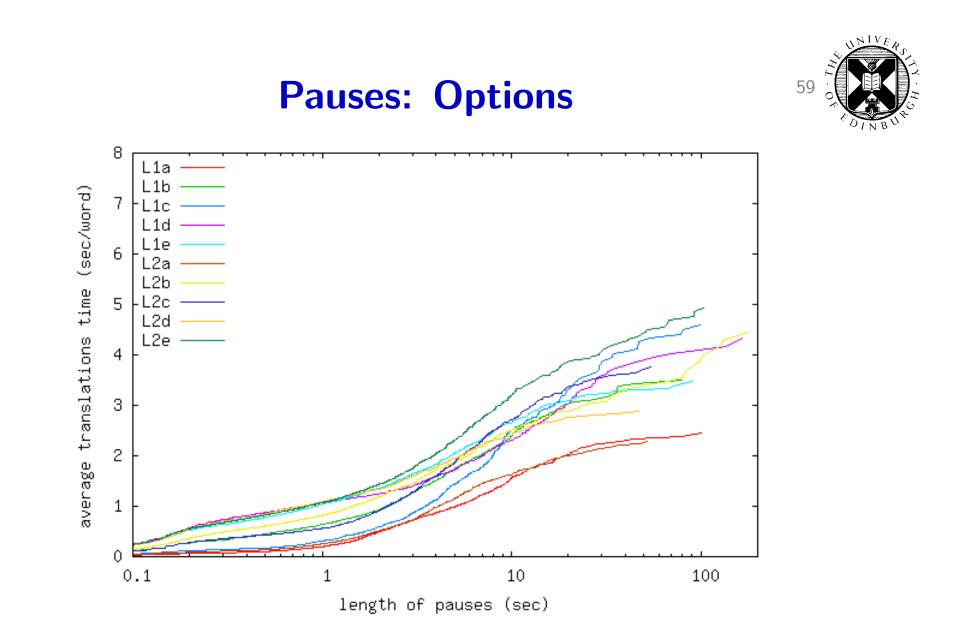
questions?

Philipp Koehn

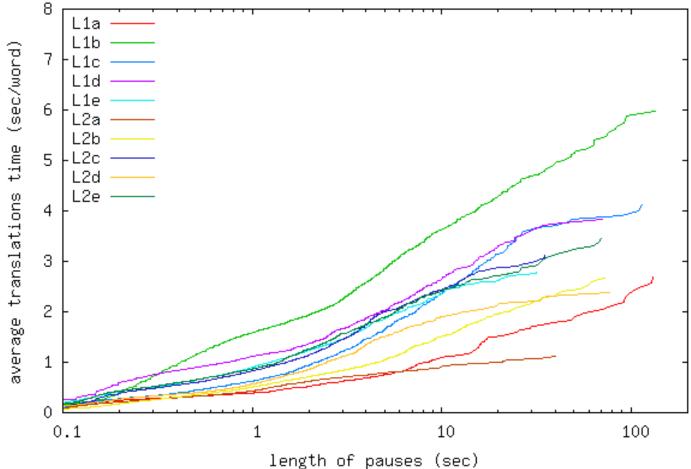
Computer Aided Translation

29 September 2009











Pauses: Postediting

