



Post-editing for translators — the theory and practice

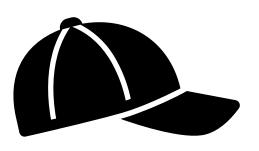
Claudia Wiesinger

Human and Artificial Intelligence in **Trans**lation (HAITrans)

University of Vienna



Two hats







Post-editing machine translation: two sides of the same coin?



slido



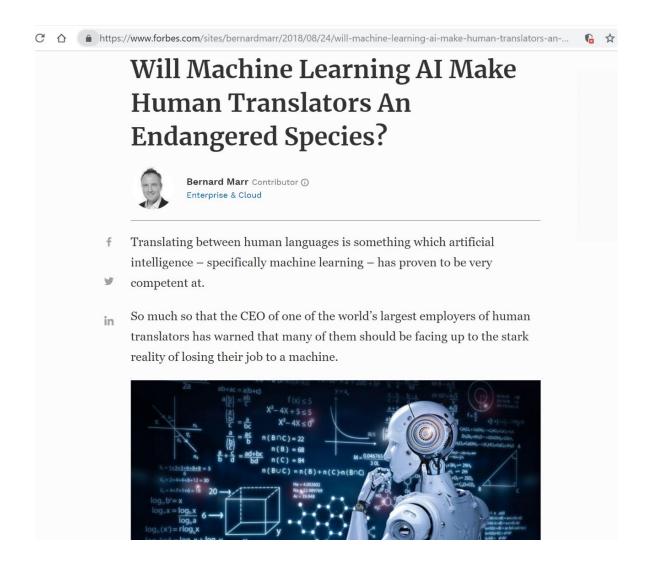
What do you associate with the term "post-editing"?



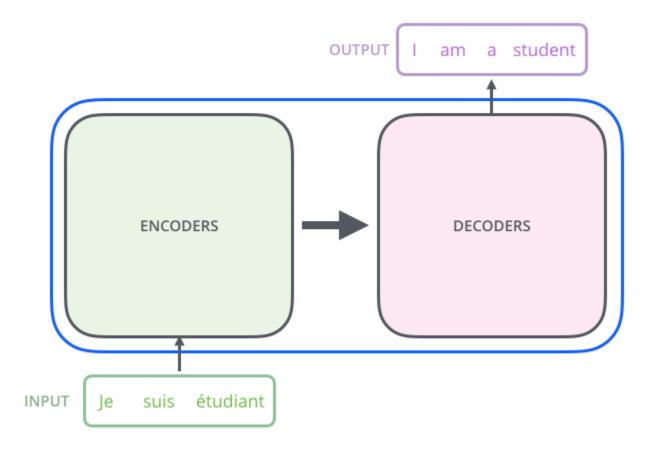
Three common assumptions

1. Machine translation (MT) is or will soon be just as good as human translators











Input

Je

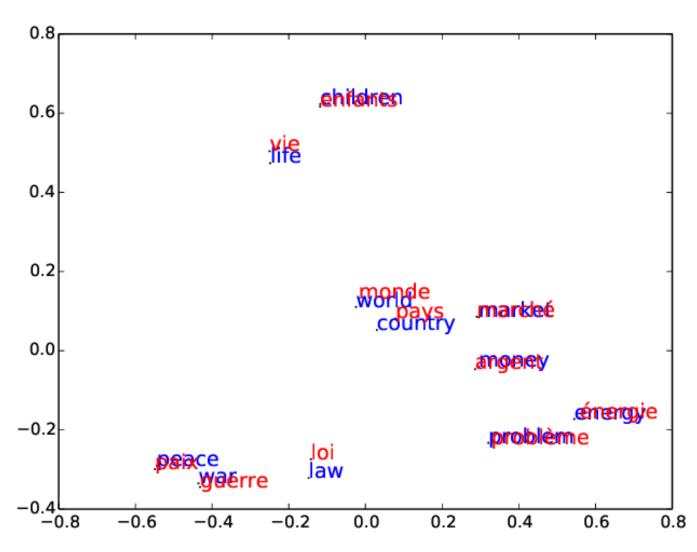
suis

étudiant

0.901	-0.651	-0.194	-0.822
-0.351	0.123	0.435	-0.200
0.081	0.458	-0.400	0.480











Input Sentence

ich glaube aber auch, er ist clever genug um seine Aussagen vage genug zu halten, so dass sie auf verschiedene Art und Weise interpretiert werden können.

Output Word Predictions

Dogt		Alternatives
Best but	(42.10/)	
	(42.1%)	however (25.3%), I (20.4%), yet (1.9%), and (0.8%), nor (0.8%),
I	(80.4%)	also (6.0%), , (4.7%), it (1.2%), in (0.7%), nor (0.5%), he (0.4%),
also	(85.2%)	think (4.2%), do (3.1%), believe (2.9%), , (0.8%), too (0.5%),
believe	(68.4%)	think (28.6%), feel (1.6%), do (0.8%),
he	(90.4%)	that (6.7%), it (2.2%), him (0.2%),
is	(74.7%)	's (24.4%), has (0.3%), was (0.1%),
clever	(99.1%)	smart (0.6%),
enough	(99.9%)	
to	(95.5%)	about (1.2%), for (1.1%), in (1.0%), of (0.3%), around (0.1%),
keep	(69.8%)	maintain (4.5%), hold (4.4%), be (4.2%), have (1.1%), make (1.0%),
his	(86.2%)	its (2.1%), statements (1.5%), what (1.0%), out (0.6%), the (0.6%),
statements	(91.9%)	testimony (1.5%), messages (0.7%), comments (0.6%),
vague	(96.2%)	v@@(1.2%), in (0.6%), ambiguous (0.3%),
enough	(98.9%)	and (0.2%),
so	(51.1%)	, (44.3%) , to (1.2%) , in (0.6%) , and (0.5%) , just (0.2%) , that (0.2%) ,
they	(55.2%)	that (35.3%), it (2.5%), can (1.6%), you (0.8%), we (0.4%), to (0.3%),
can	(93.2%)	may (2.7%), could (1.6%), are (0.8%), will (0.6%), might (0.5%),
be	(98.4%)	have (0.3%), interpret (0.2%), get (0.2%),
interpreted	(99.1%)	interpre@@ (0.1%), constru@@ (0.1%),
in	(96.5%)	on (0.9%) , differently (0.5%) , as (0.3%) , to (0.2%) , for (0.2%) , by (0.1%) ,
different	(41.5%)	a (25.2%), various (22.7%), several (3.6%), ways (2.4%), some (1.7%),
ways	(99.3%)	way (0.2%), manner (0.2%),
•	(99.2%)	(0.2%), , (0.1%),
	(100.0%)	



There's no "best" MT system

MT performance depends on how similar your data is to the data used to train the vendors' models, as well as their algorithms.

1.1 Machine Translation Landscape



Dynamic domain adaptation

aws Amazon

(T) ModernMT

Large Language Models

Vertical Stock Models

C Alibaba

Microsoft

NiuTrans

Omniscien

PROMT

CloudTranslation

🗱 Baidu

Al21 Al21 A\ Anthropic

BigScience

ALAION

RoyalFlush

SYSTRAN

[ղ] Ubiqus

XL8

SAP SAP

Microsoft

M MosaicML

S. Stability

9 of 91

EleutherAl Google

Cohere

Custom terminology support

RWS

SYSTRAN

[ղ] Ubiqus

Yandex

aws Amazon

Baidu

DeepL

Google

IN IBM

Microsoft

Rozetta Rozetta

Static domain adaptation

C Alibaba 490€ AppTek

🔅 Baidu

Globalese

Google

ISM IBM

CloudTranslation

KantanAl Microsoft

PangeaMT

Prompsit

PROMT

Rozetta Rozetta

Omniscien

SYSTRAN

Tilde

RWS

[ŋ] Ubiqus

Yandex

CIOL Translators Day 2024 - Claudia Wiesinger

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The State of Machine Translation 2023

Intento (2023: 9)



Proceed with...





Adequacy vs. fluency in neural MT

"Compared to statistical machine translation, neural machine translation may be **less** adequate, even if **more fluent**. In other words, the translation may diverge from the input in various ways, such as not translating part of the sentence or generating unrelated output words."

(Koehn, 2020: 165)







Things to look out for – Hallucinations

the Ministry of Energy, Natural Resources, Environment and Mines (MiAmbiente), the Forest Conservation Institute (ICF), the Institute of Property (IP), the National Agrarian Institute (INA), the Revenue Administration

Service (SAR), the Customs Revenue Subdirectorate (DARA), the Association of Honduran Municipalities (AMHON), the Honduran Social Security Institute, (IHSS), the Ministry of Labour and Social Security (STSS), the Ministry of Agriculture and Livestock (SAG), the National Directorate of Indigenous and Afro Honduran Peoples (DINAFROH), the Institute for Access to Public Information (IAIP), the Public Prosecutor's Office (MP), the Security Ministry (SSN, representing the National Preventive Police (PNP)), the National Defence Ministry (SEDENA, representing the armed forces of Honduras (FFAA)), the Attorney General's Office (PGR),

Honduras (FFAA)), the Attorney General's Office (PGR) the Central Bank of Honduras (BCH), the National School of Forest Sciences (ESNACIFOR) and other academic institutions concerned with environmental and forestry issues.

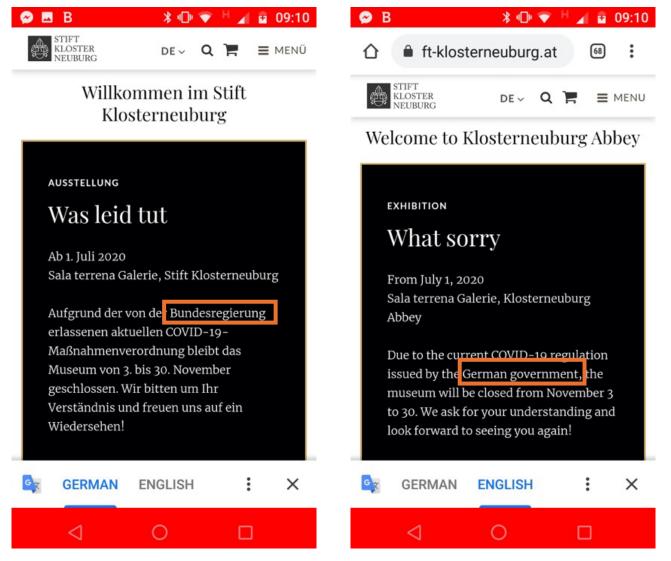
Ministerul Energiei, Resurselor Naturale, Mediului și Minelor (MICF), Institutul pentru Conservarea Naturii (ICF), Institutul National Agrare (IP), Serviciul de administrare fiscală (SAR), Subdirecția de

Administrare Fiscală (AMHON), Ministerul Muncii și Securității Sociale (SAG), Ministerul Muncii și Securității Sociale (SSN), Ministerul Muncii și Securității Sociale (SSN, reprezentant al forțelor armate ale Honduras — FFAA), Biroul Procurorului General (PGR), Central Bank of Honduras (CSIB), Școala Națională de Științe Forestiere (ESNAF), Școala Națională de Științe Forestiere (ESNAF) și alte instituții academice care se ocupă de probleme de

mediu și silvice.

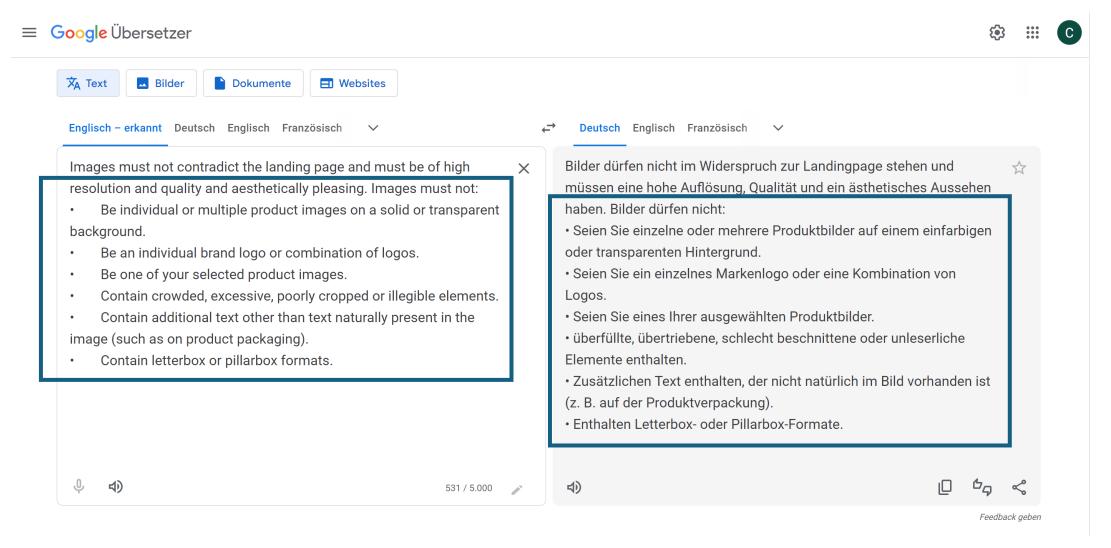
Things to look out for – Distortions





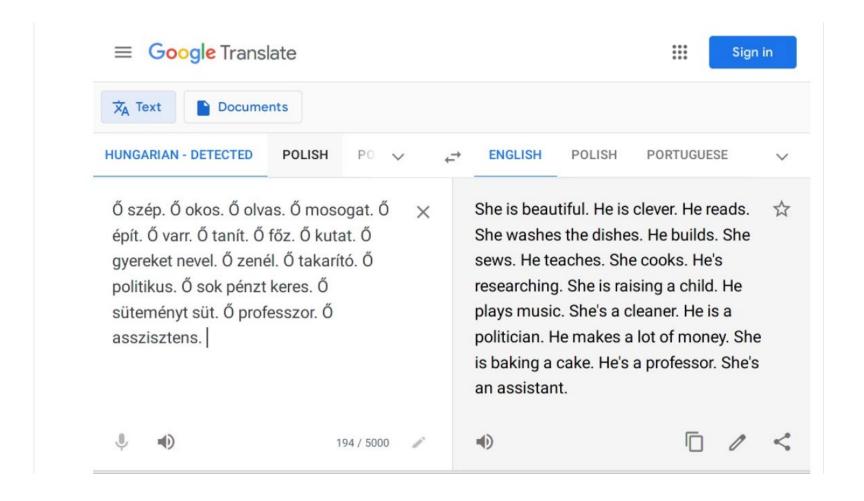
Things to look out for – Loss of coherence





Things to look out for – Gender bias





Things to look out for – Inaccuracies (names)



Iryna Skliar Iryna S<mark>g</mark>liar

James R Elliot James R Elliot<mark>t</mark>

He, H. and Ye, Z. Er, H. and Ya, Z.

Roland W Force Roland W Macht

Roland W Kraft

Roland W Streitkräfte;

Michael P Lesser Michael S. Lesser

Tosepu, Ramadhan Tosepu, RAMADAN

Fredrik Palm Friedrich von Schiller

With many thanks to our students





1674364114		Secs/Word: 06'' PEE: 0%	Segment status APPROVED	
Source		Gefördert wurden rund 7.800 Pkws mit reinem Elektroantrieb, knapp 1.100 Plug-In-Hybride und über 1.500 E-Motorräder und E-Mopeds.	Words:	17
Suggestion				
Translation	•	Around 7.800 have been supported with electric cars, almost 100 Plug-In-Hybride and more than 1,500 E-Motorräder and E-Mopeds.	TTE:	24''
Revision	•	Around 7.800 cars have been supported with electric cars drive, almost 100 Plug-In-Hybride and more than 1,500 the motor rader bikes and the motor rader bikes are the motor rader bikes and the motor rader bikes are the motor rader bikes.	TTE:	01'34''
QA	Human (1)	Terminology and translation consistency: [Neutral]		



eTranslation

The European Commission's Machine Translation system

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Translate online

eTranslation is a cutting-edge neural machine translation service provided by the European Commission. It was officially launched on 15 November 2017, superseding the earlier statistical system, MT@EC.

The <u>web page</u> is free to use for EU institutions, public administrations, universities, EU freelance translators, SMEs, European NGOs and projects financed by the <u>Digital Europe Programme</u>, located in an EU country, Iceland, Norway, Liechtenstein or Ukraine. **Registration is required** (see **Access** below). eTranslation also offers a web service API for machine-to-machine connections from eligible parties.

Product features

- High security all data processed by the system stays within the Commission's firewalls and can't be seen by outsiders
- Translates from and into any <u>official EU language</u> as well as Arabic, Chinese, Icelandic, Japanese, Norwegian, Russian, Turkish and Ukrainian
- Offers several styles of translation ("domains") in the fields of EU Formal Language, General Text. Finance and a few more
- · Supports all common office formats, including pdf
- Can provide tmx or xliff files for professional translators

Raw quality

eTranslation produces raw machine translations. Use it to get the gist of a text or as the starting point for a human-quality translation. If you need a perfectly accurate, high-quality translation, the text still needs to be revised by a skilled professional translator.



Three common assumptions

- 1. Machine translation (MT) is or will soon be just as good as human translators
- 2. Machine translation is the same as translation



Syntag	ctic strategies	Semar	ntic strategies	Pragma	atic strategies
G1	Literal Translation	S1	Semantically Equivalent	Pr1a	Domestication
G2	Loan	S2	Antonymy	Pr1b	Foreignization
G3	Word structure change	S3	Hyponymy	Pr2a	Explicitation
G4	Unit shift	S4	Converses	Pr2b	Implicitation
G5	Phrase structure change	S5	Abstraction change	Pr3a	Information addition
G6a	Clause structure change	S 6	Distribution change	Pr3b	Information omission
G6b	Clause structure change	S7	Emphasis change	Pr4	Interpersonal change
G7	Sentence structure change	S8	Paraphrase	Pr5	Illocutionary change
G9	Modality shift	S9	Trope change	Pr6	Coherence change
G10	Scheme change	S10	Cohesion change	Pr9	Transediting
G11	Addition			Pr11	No change
G12	Omission			Pr12	Domain adaptation
G13	Punctuation change			Pr13	External Information adaptation

(Chesterman 1997/2016)



Three common assumptions

- 1. Machine translation (MT) is or will soon be just as good as human translators
- 2. Machine translation is the same as translation
- 3. Post-editing is a universally understood concept & set of practices



Post-editing is...

...the editing and correcting machine translation output (ISO 18587:2017)

- Full post-editing or light post-editing?
- How is post-editing different from revision?





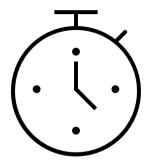




Post-editing skills





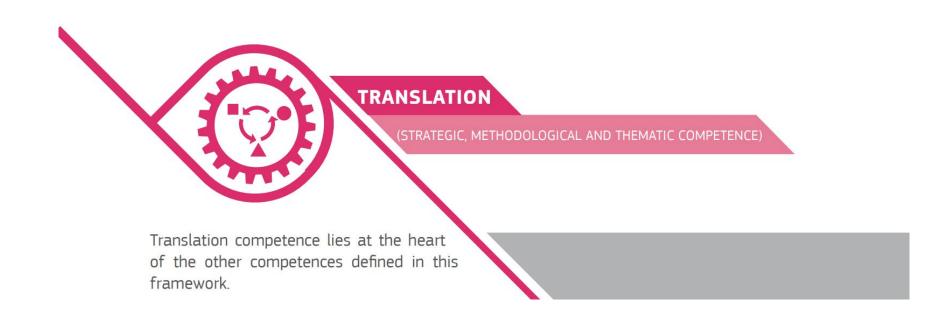


Reasonable speed



Guidelines





14 Post-edit MT output using style guides and terminology glossaries to maintain quality standards in MT-enhanced translation projects



A disclaimer







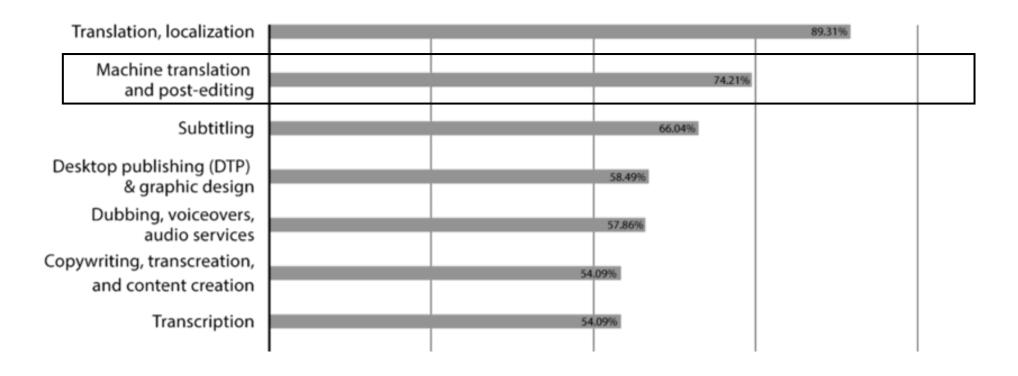
The priming effect and post-editese

"we have found that [post-edited texts] tend to be **simpler** and more normalised and to have **a higher degree of interference** from the source text than [human translations]. This seems to be caused because these characteristics are already present in the MT outputs that are the starting point of the [post-edited texts]"

(Toral, 2019: 280)







% of language service providers offering this service



And yet...

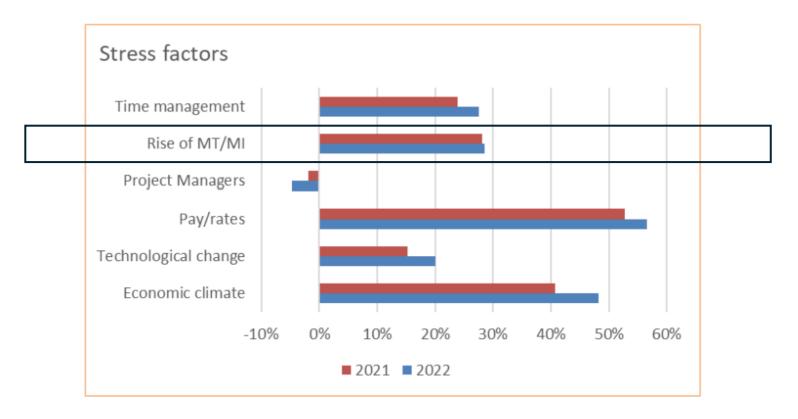


Figure 37 - Stress factors independent professionals



And yet...

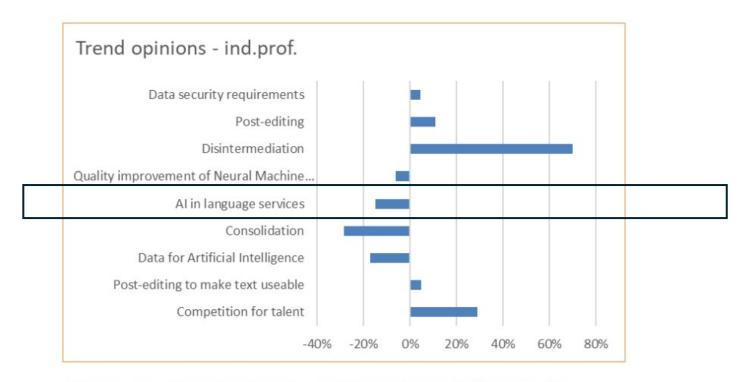
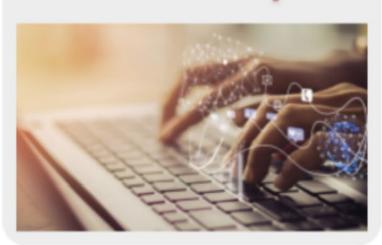


Figure 36 - Trend opinions - independent professionals



Workshop



MT post-editing skills workshop (English as the target language)

Friday 26 April 2024 Delivered by Prof. Dragos Ciobanu and HAITrans 9am - 12pm UK time

Click here to find out more

Workshop



MT post-editing skills workshop (English as the source language)

Friday 26 April 2024
Delivered by Prof. Dragos Ciobanu and HAITrans
1pm - 4pm UK time

Click here to find out more



https://www.ciol.org.uk/online-workshops-focus-sessions



By way of conclusion

- MT has got big backers and is constantly marketed as "the future of translation"
- There are a lot of misconceptions when it comes to MT
- MT can be genuinely useful (and occasionally catastrophic)
- Post-editing has its own ISO standard (ISO 18587:2017)
- MT/AI is on the rise, but perceived as a threat by many language professionals

Passing the mic to you





