Semantix

LT Innovate June 25th

MEMSOURCE

Deriving Business Value from Predicting MT Quality





Outline

- Intro
- Industry trends
- Memsource background on MTQE
- Pilot
- Questions







Who we are

Nils Adie

- Director of Technology, Semantix
- In Localization Technology since 2007
- Interest in Users getting the most out of Tech
- Data driven decisions and process improvement

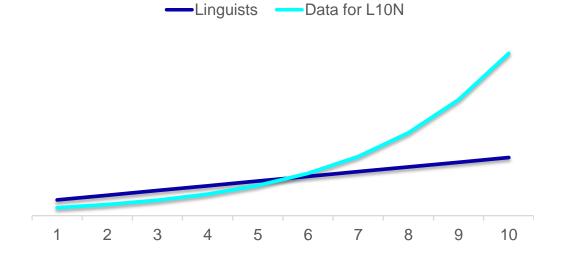
Andrea Tabacchi

- Lead Solution Architect, Memsource
- Technology Enthusiast
- 14 years in Localization
- Working with MT the last 10 years

Industry Trends

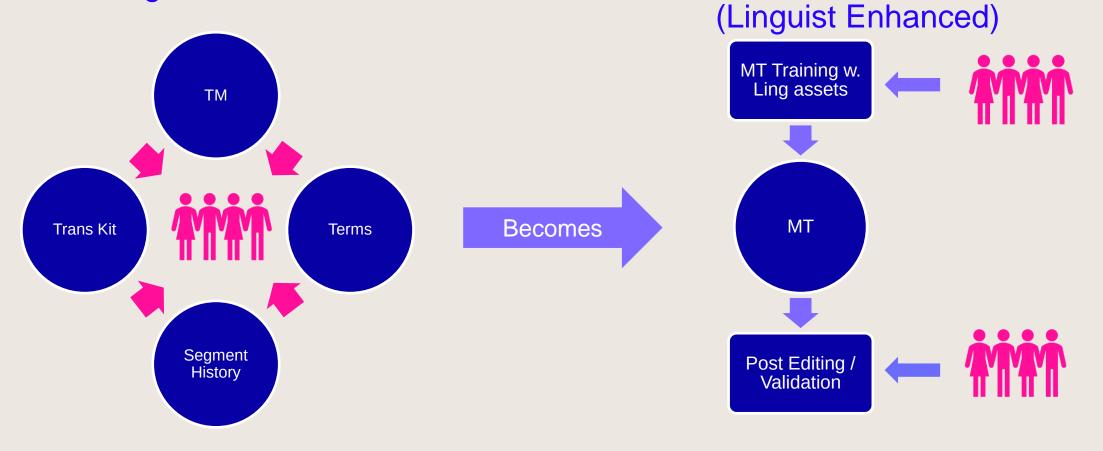
- Exponential growth of Data
- Sustainability
- Artisanal versus Assembly line style localization
- Killing the Linguist Centric Localization Setup
- Machine Translation & Al Evolution





Localization Centre of Gravity

Linguist Centric



MT Centric

MACHINE TRANSLATION EVOLUTION

10 years ago

IT DOESN'T REALLY WORK

Because of lack of training data, technology, quality of source content, post-editors availability

5 years ago

IT MIGHT ACTUALLY WORK

But change is hard (and procrastinating is human)

IT WORKS

How can I get the best out of it?

Now

COMMON OBSTACLES IN MT IMPLEMENTATION

- Complexity of use and predictability of results
- Number of available technologies on the market
- Undefined effort estimation and pricing models

AND HOW WE SOLVED THEM

- Offer connections to as many technologies as possible
- Simplify the selection of the best tech
- Develop ways to estimate the effort (MTQE) prior to PE or verify it afterwards (edit distance report and time-tracking)

MEMSOURCE'S AI JOURNEY

2016 – What's all this hype about AI? Studied, went to conferences

2017 – Can we do something with it? Yes we can, but we need a team (and an infrastructure)

2018 – We released NTs and MTQE and we have now a clear vision: **building an intelligent TMS!**

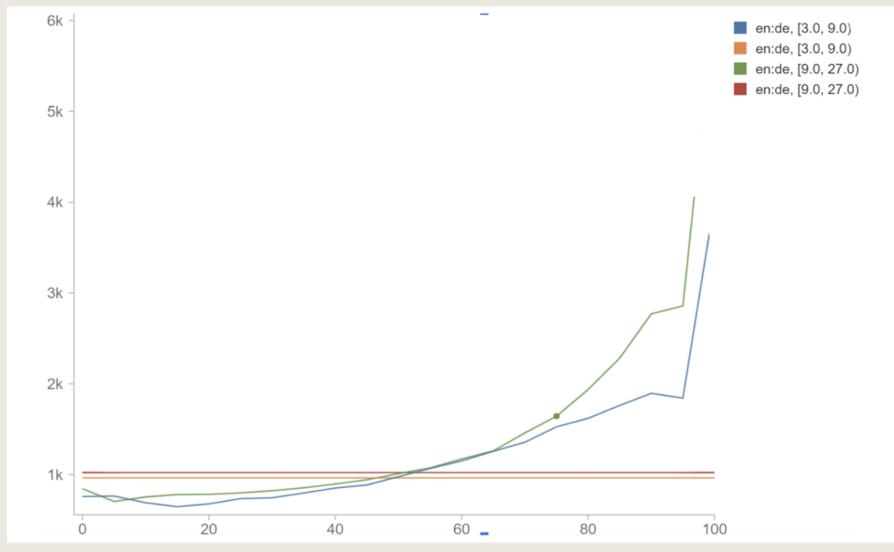
2019 – Can't share right now 😳

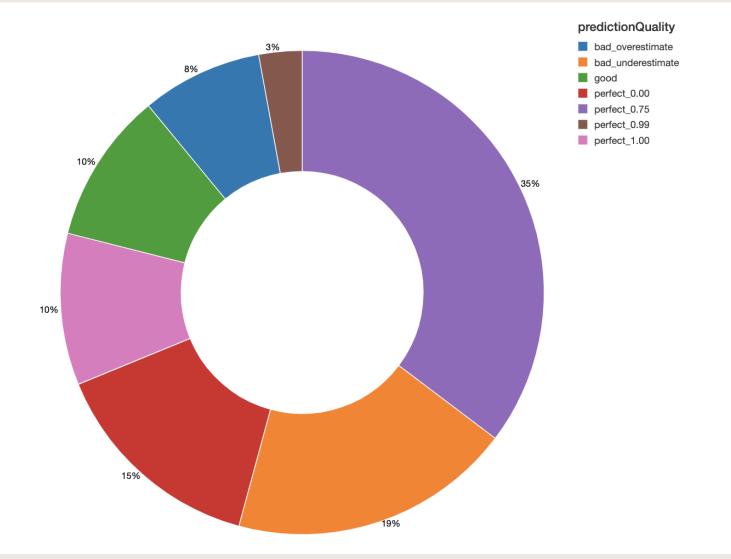
Apply a confidence score to MT outputs

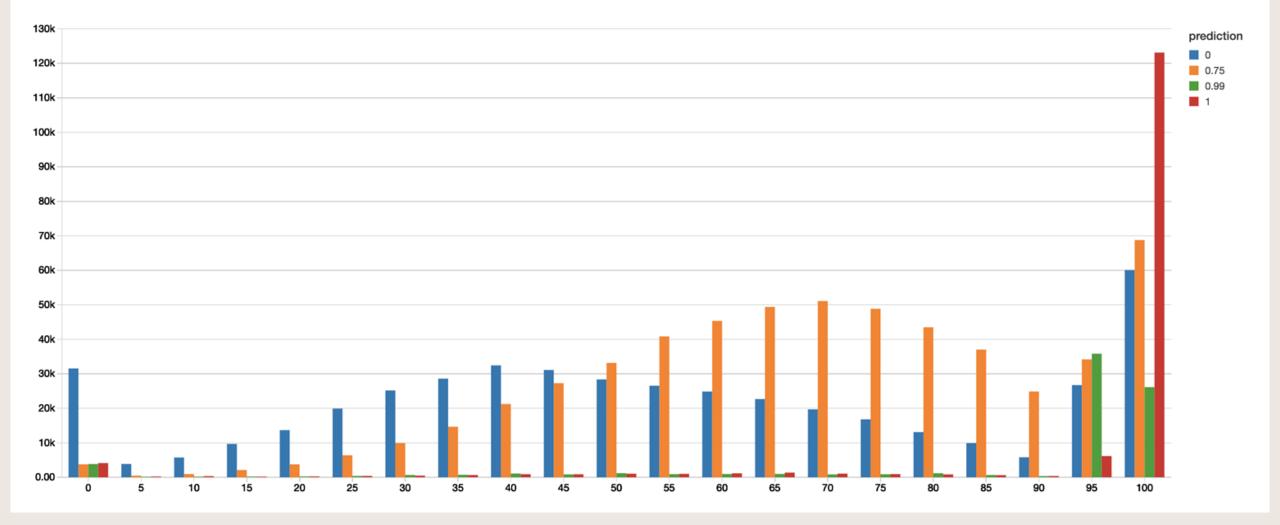
- Based on a proprietary modification of chrF3
- Utilizes deep neural networks
- Trained historical data
- Support for 71 language pairs
- MT engine independent

Goals

- Better project scoping
- Better post-editing







The Pilot

Background

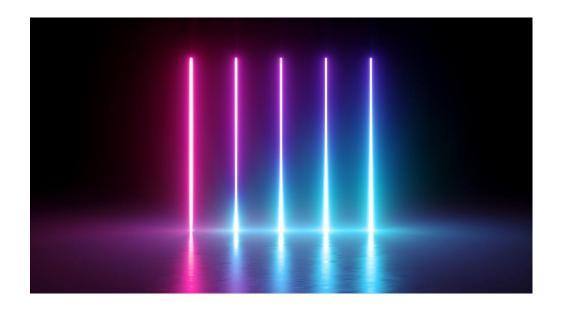
In a study in Q4 2018 we saw a 50% increase in the net word count drop when deploying MTQE. This underpinned the decision to deploy across the board at Semantix. Did we do the right thing?

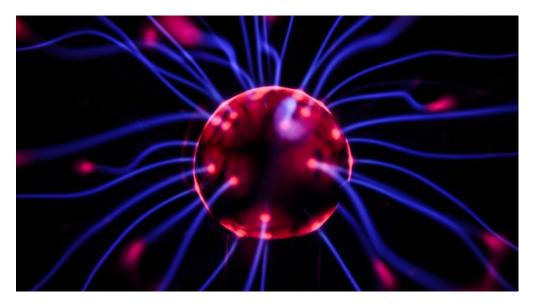
Questions we asked:

- Did the data support our decision?
- What is the MTQE predictive accuracy?
- Can we find a measurable effect on linguist productivity?









The Pilot: Method

On a subset of our Memsource clouds Semantix captured aggregate net wordcount data for Jan-May across mainly Scandinavian language combinations.

For 100 projects:

- we ran a separate analysis for each project to be able to separate MTQE predictions from MT post editing results.
- On segment level we extracted linguist editing time, MT score, TM score, and segment length

The Pilot: Findings

Question:

Did the data support our decision?

Results:

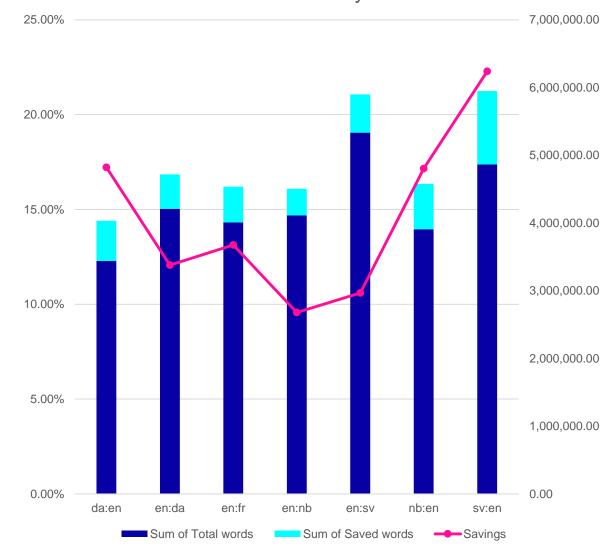
Variations, but average of 15% savings.

Conclusion:

Yes, in line with our predictions from the previous study.

Bonus:

Also see smaller but significant savings in unexpected language pairs.



Saved Word in Post analysis with MTQE

The Pilot: Findings

Question:

What is the MTQE predictive accuracy?

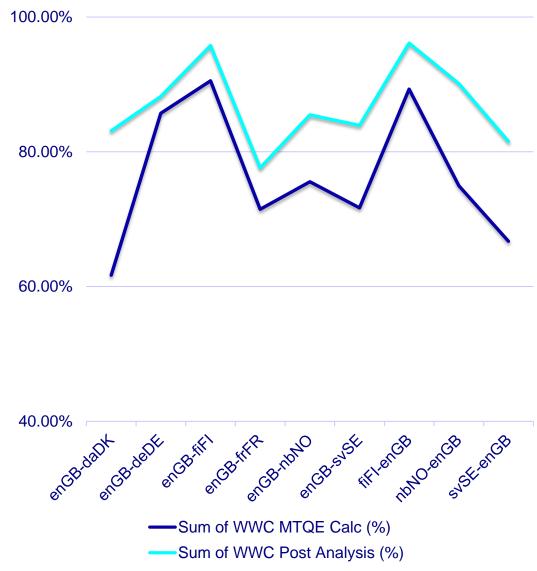
Results:

MTQE savings predictions are overly optimistic, but there is a correlation with the actual outcome.

Conclusion:

We can work with this. By tweaking our net rate schemes we can adjust for the outcomes (and have found a pattern that works).

How outcome compares to prediction



The Pilot: Findings

Question:

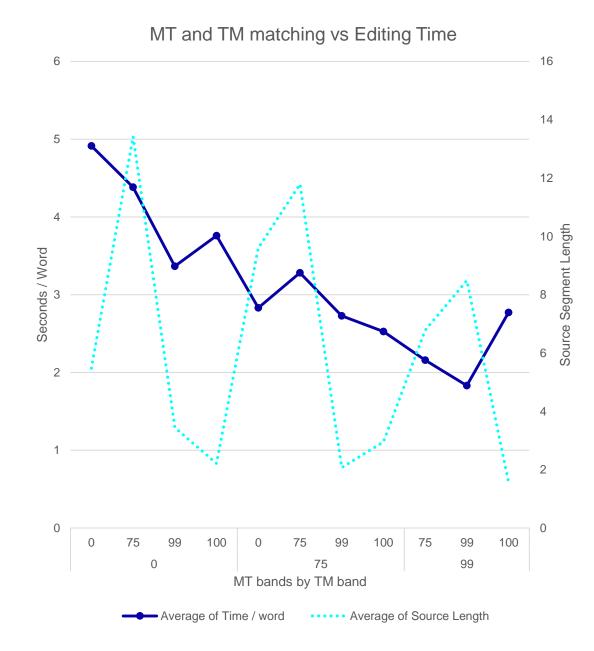
Can we find a measurable effect on linguist productivity?

Assumption:

Linguist relies on highest of MT and TM score

Results:

Largely Inconclusive



Questions?

