

Proceedings of the 14th Conference of the Association for Machine Translation in the Americas October 6 - 9, 2020, 1st Workshop on Post-Editing in Modern-Day Translation

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Presentation Outline

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- 1. Who we are
- 2. Business areas
- 3. Clients
- 4. Introducing QUEST the MT Quality Estimator
- 5. Comparing translation modes (HT, PEMT, QUEST)
- 6. Demo
- 7. Model architecture
- 8. Parting words

Who we are

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LangTec:

- 10 years in business, 15 employees
- We've been working on MTQE since 2018
- QUEST is our second fully functioning QE model and a significant advance over its predecessor



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Clients 1. Semantic Analytics



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Clients

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2. Automated Text Generation



Clients

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3. Computational Linguistics

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Introducing QUEST – MT Quality Estimator

Overview

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- 1. Our in-domain QUEST model is capable of predicting post-editing effort with unprecedented accuracy: only 2% error
- 2. QUEST can be used with any language pair, any machine translation engine and for any domain.
- 3. QUEST can also be used out-of-domain still providing superior prediction accuracy compared with experience-based estimations and other machine learning models
- 4. QUEST's prediction accuracy enables translation service providers to achieve much more competitive pricing and deadline-setting.

In-Domain QUEST Model

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- When historical domain data are available we can also train in-domain models.
- The in-domain baseline beats the out-of-domain baseline.
- The in-domain baseline also beats the out-of-domain TCA.
- The out-of-domain QUEST model beats all baselines and the out-of-domain TCA.
- The in-domain QUEST
 model beats all baselines
 and all TCA models, very
 closely approximating the
 actual post-editing effort.

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Comparing approaches

SEMANTIC TEXT PROCESSING MT + Post-MT + Post-MT + post-Manual Editing (In-Editing + editing + **Translation** Actual domain QUEST (out-**QUEST** (in-(HT) baseline) of-domain) domain) Word Count 22,635 22,635 22,635 22,635 22,635 Predicted Post-0 10,474 9,538 6.443 6,561 Editing Operations Prediction Error 0% 60% 45% 2% 0% 30.07 hrs Total 56.59 hrs. 48.00 hrs. 43.72 person 29.53 hrs. Turnover Effort 7.07 person 6.00 person 3.69 person days 3.76 person hrs. days 5.46 person davs davs days Fraction of 100% 46% 42% 28% 29% words to be processed Per-Word Rate EUR 0.09 EUR 0.05 EUR 0.04 EUR 0.03 EUR 0.03 **Total Cost** EUR 2,037 EUR 679 EUR 1,132 EUR 905 **EUR 679**

Demo

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QUEST LIVE-DEMO

Submit a text and get a machine learning-based prediction of the post-editing effort that will be needed to correct its machine translation.

Currently supported are German texts to be translated into English. .tmx files may also contain English machine translations in addition to the German source segments. Our models perform most reliably in the domain of 'technical manuals'. For this demo, we recommend submitting texts that contain no more than 5000 segments.

| | Select Model Co | nfiguration | | |
|--------------------|--------------------------------|------------------------|------------|--|
| | Select Your Prediction Model f | or Quality Estimation: | | |
| | Deep Learning Model w | ith with 17.8 🗸 | | |
| | Select Language Pair: | | | |
| | DE> EN | ~ | | |
| | | | | |
| Upload File | | | Input Text | |
| Choose a file | | | | |
| (.tmx & .txt only) | | | | |
| | | | | |
| | | | | |

Different PE effort estimations

Model architecture

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QUEST's usage contexts

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- 1. Post-Edit Effort Prediction
- 2. MT-Recommendation Tool
- 3. MT-Profiling Tool
- 4. MT Quality Document Sorting
- 5. MT Quality Threshold Plug-In

Parting words

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- QUEST can be used on-premise or in the cloud and custom models can be built within hours.
- Our machine learning model also scales to any language pair
- QUEST works as solution for leveraging MT for lesserresourced languages
- LangTec can provide a free customized trial QUEST model for interested parties

Thank you!

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