Vakula: Framework for Large-Scale Parallel Corpus Evaluation

Ensemble Quality Estimation Models Versus Human Assessment

Dmytro Chaplynskyi





Kyrylo Zakharov



The Problem

Why Quality Matters?

- NMT needs large, quality datasets
- English-Ukrainian: <u>158M</u> pairs available on OPUS
- But: BAD QUALITY, duplicates, errors, inappropriate content
- Manual inspection at scale? Impossible



Research Questions

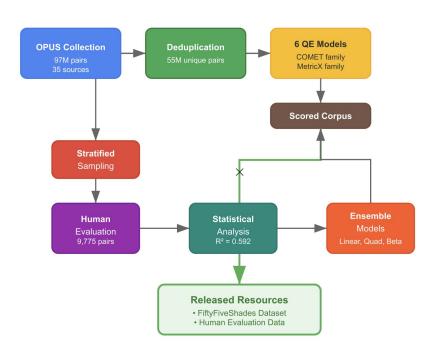
What We Asked

- Can we automatically evaluate large parallel corpora?
- How good are modern QE models vs human evaluation?
- Can ensemble models improve evaluation?
- Is there things <u>beyond</u> QE models to help?

Our Approach

Pipeline Overview

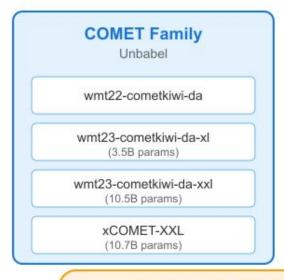
- Collect 97M sentence pairs from OPUS
- Deduplicate → 55M pairs
- Score with 6 QE models
- Human evaluation on 9,775 pairs
- Build ensemble models
- Calculate the correlation
- Rescore entire corpus

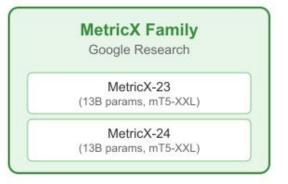


Quality Estimation Models

Six Models Used

Quality Estimation Models Used



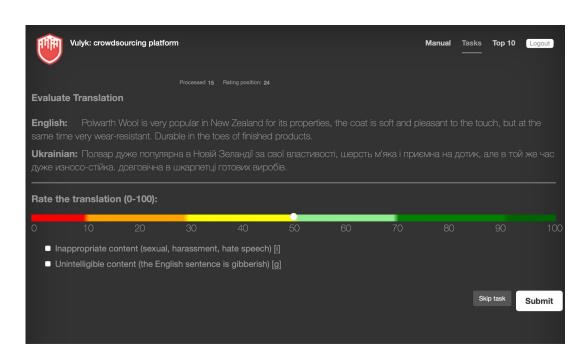


All models: Multilingual + Reference-free mode

Human Evaluation Setup

Crowdsourcing with Students

- 20+ linguistics students
- 0-100 scale
- 9,775 pairs evaluated
- 3+ evaluators per pair
- Flags for inappropriate/garbled content



Human Agreement

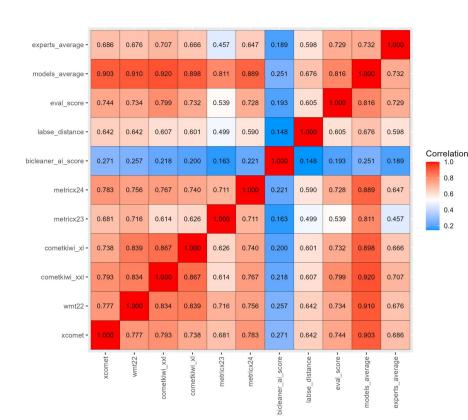
Inter-Annotator Results

- ICC = 0.428 (poor to moderate)
- After percentile transformation: ICC = 0.542
- High variability between evaluators
- Models more consistent than humans

Model Performance

Correlation with Human Judgments

- Models correlate strongly with each other
- Moderate correlation with human scores
- Different model families capture different aspects



Ensemble Models

Three Approaches

- 1. Linear: All 6 models as predictors
- 2. Quadratic: Average score + squared term
- 3. Beta: For bounded [0,1] data

Best: Quadratic model explains ~60% of variance

Surprising Results

Additional Experiments

- Bicleaner-AI: poor correlation (0.19)
- 2. LaBSE similarity: good correlation (0.59)
- 3. LLM-as-Judge (Gemini): correlation 0.76!

Simple methods can work!

What We Release

Open Resources

- Vakula Framework: Download, deduplicate, evaluate OPUS corpora
- 2. FiftyFiveShades Dataset: 55M pairs with all scores
- 3. Human Evaluation Data: 9,775 expert-annotated pairs
- 4. Vulyk Plugin: For crowdsourcing evaluation

https://github.com/lang-uk/vakula
https://huggingface.co/datasets/lang-uk/FiftyFiveShades

Future Work

Next Steps

- Professional translator evaluation
- Train NMT on filtered data
- Ablation studies with different thresholds
- Extend to more language pairs
- More experiments on LLM-as-a-Judge

Conclusions

What We Learned

- QE models work, but relationship is non-linear
- Ensemble improves prediction
- ~9% of web-crawled data is problematic
- Simple methods (LaBSE) can compete with complex models
- LLM-as-a-Judge is good but expensive on such a scale



https://github.com/lang-uk/vakula/

https://huggingface.co/datasets/lang-uk/FiftyFiveShades

chaplinsky.dmitry@gmail.com,

kirillzakharov13@gmail.com

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