

Machine Translation Assessment



Summary

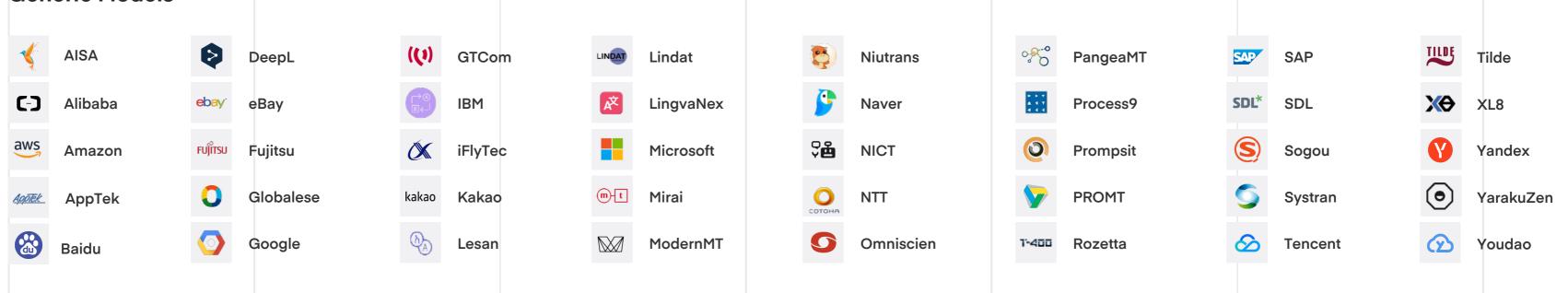
B

- Machine Translation Playground
- Dataset
- Methodology
- Study Results
- Conclusion
- Appendix

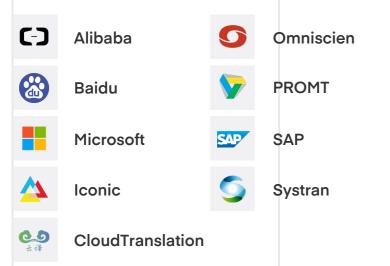
Machine Translation Playground



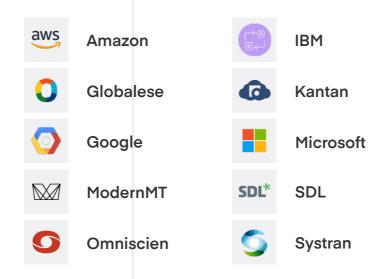
Generic Models



Custom Models



Auto Domain Adaptation



Manual Domain Adaptation

| C-D | Alibaba | 9 | Omniscien | SDL* | SDL |
|---------------|------------------|------|-----------|-------|--------|
| <u>AgōTek</u> | AppTek | 0.00 | PangeaMT | TILDE | Tilde |
| | Baidu | 0 | Prompsit | V | Yandex |
| 6 5 | CloudTranslation | | PROMT | | |
| | Iconic | 5 | Systran | | |

The list shows MT products with API services. All product names and brands are property of their respective owners. All brands, products and services used in this report are for identification purposes only. Use of these brands does not imply endorsement.

Machine Translation Playground



Analyzed in this Study



Google Cloud
Translation API



ModernMT

Translate Text API



Yandex
Translate API



Amazon Translate

The list shows MT products with API services. All product names and brands are property of their respective owners. All brands, products and services used in this report are for identification purposes only. Use of these brands does not imply endorsement.

Dataset



Corpus

Training set is collected from OPUS and TAUS.

Each data record has following aspects;

- Source Text
- Reference Human Translation
- Language Pair
- Industry





Datasets

Language Pair

14 language pairs,~150K sentences for8 industries.

Why we choose this language pairs?

Most of the international brands prepare the content in English as a first direction. That's why we choice English as source language.

The target languages are most spoken languages in the world. 74% of all the world's languages are these pairs. It is a fundamental fact when planning a global expansion strategy.



© Lugath, Inc.

Datasets

B

Industry

8 industries

~ 50 segments per language pair for all industries

| Automotive | | | | | | | | | | | | | | | 10 |
|----------------------|------|------|------|------|------|------|------|------|------|------|------|------|------|------|----|
| Automotive | < 50 | < 50 | 61 | 62 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | 54 | |
| Computer Software | < 50 | 285 | 522 | 107 | 308 | 120 | < 50 | 118 | 208 | 58 | <50 | 200 | 62 | < 50 | |
| Finance | < 50 | 576 | 91 | < 50 | 62 | < 50 | < 50 | < 50 | < 50 | < 50 | 61 | 72 | 61 | 61 | |
| Healthcare | 121 | < 50 | 124 | 86 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | 51 | 51 | 51 | 56 | 50 |
| Consumer Electronics | < 50 | < 50 | 119 | 61 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | 89 | 75 | 52 | |
| Tourism and Arts | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | |
| Computer Hardware | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | |
| Business Services | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | < 50 | 0 |
| | AR | DE | ES | FI | FR | JA | ко | NL | PL | PT | RO | RU | TR | ZH | |

Q3

Methodology

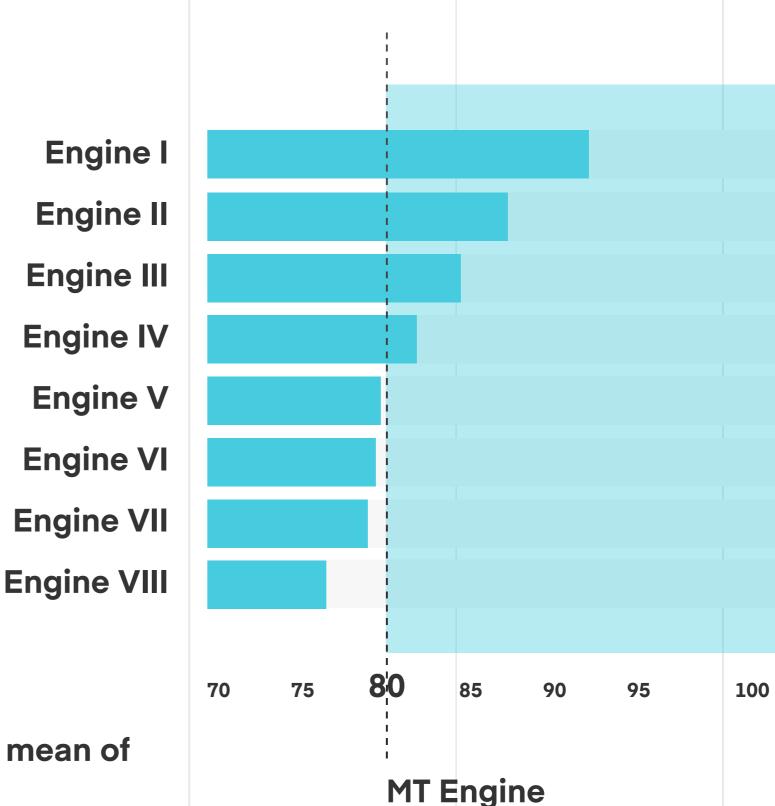
1

We are sending all source texts in training set to MT Engines for translation and collecting translated target sentences.

2

After, we are calculating the similarity scores between target senteces and human refrences that are located in training set.

MT Engine level score is calculated with mean of sentence level scores.



MT ENGINE SCORE

Q3

Quality Threshold

Methodology



How the sentence similarity is calculated?

| Lexical |
|-------------------|
| Similarity |
| Algorithms |

BLEU

hLEPOR

These methods mainly focus to the number of **common words between two sentences.** The advantages of the lexical similarity methods are that they **perform well in capturing the translation fluency**. **Low data** need and **low costs** for calculation but **not enough for capturing the adequacy**.

Syntactic Similarity Algorithms

These methods combining lexical similarity methods with POS(part-of-speech) tags. Focusing more to synonyms of common words. Need dictionary to understand the verb, noun, adjective, adverb or preposition. The main advantage is the syntactic methods are that they perform more then lexical similarity algorithms in capturing the translation adequacy. Higher data need then lexical methods and low cost for calculation. Still not enough for capturing the paraphrases. Sentence length also affect the score.

Semantic Similarity Algorithms

These methods **combining syntactic similarity** algorithms **with machine learning**. People can make sentences via using different words or words order for same meaning. Focusing to capture these scenarios. **High data** need and **high cost** for calculation.

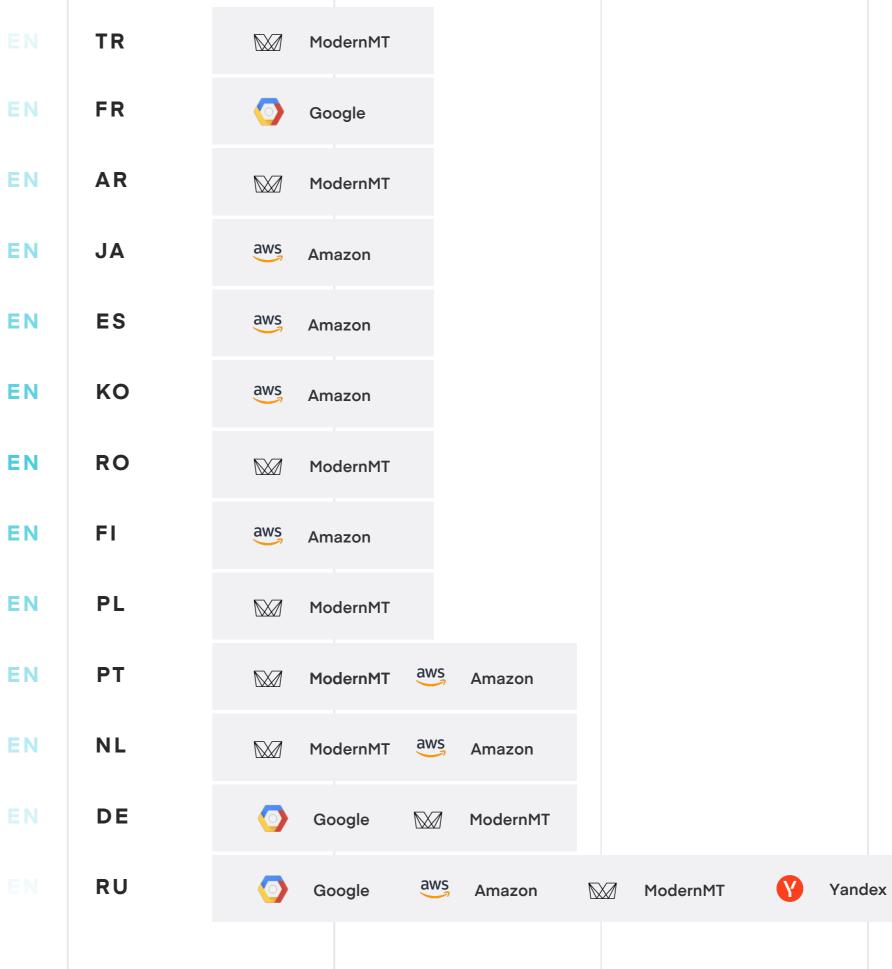
BERTScore

Most of the algorithms have trade-offs. We are using combined strategy to evaluate the sentence similarity for detecting the best performing MT engine.

Study Results

Best MT Engine Per Language Pair

Test data includes all industries that we mentioned before in this study for each language pair.



BEST MT ENGINES

Study Results

Best MT Engine Per Industry





Amazon

Translate



Google Cloud

Translation API



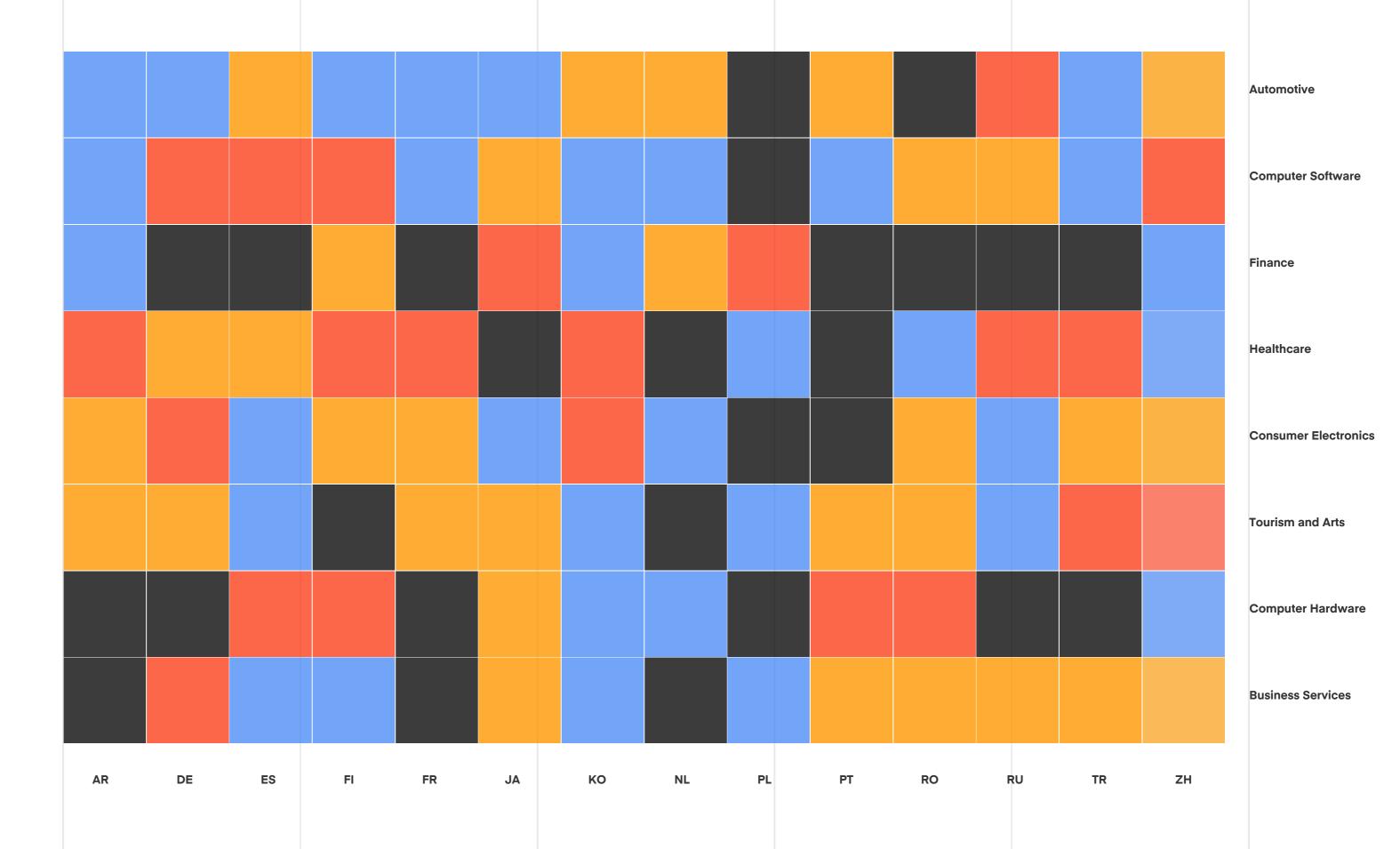
ModernMT

Translate Text API



Yandex

Translate API

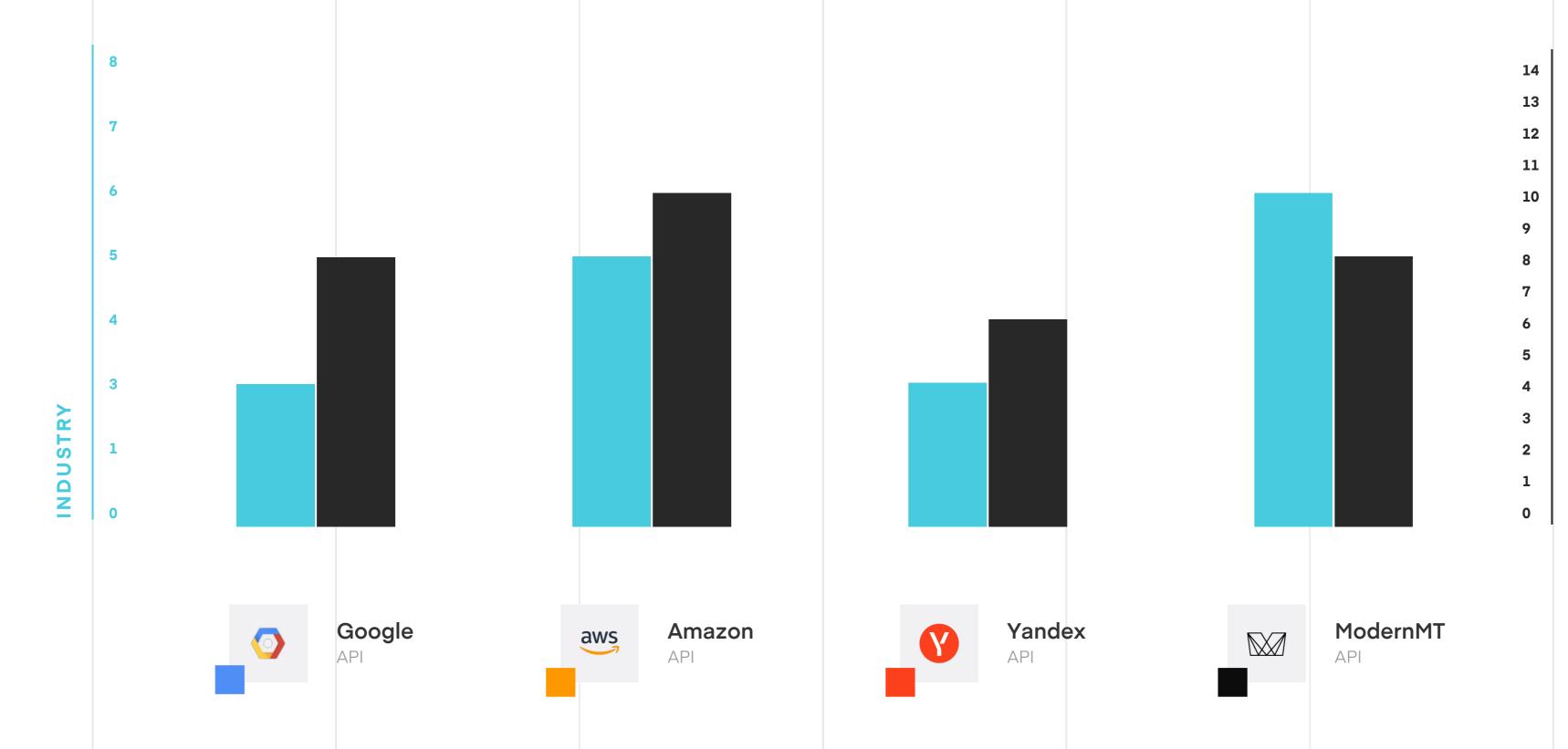


© Lugath, Inc.

Q3 2021

Study Results

Top Performing MT Engines



MT Engine Coverage according to industry or language pair.

The success score thresold considered as 80 (in range of 0-100) to filter the best performed MT engine.

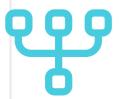
Q3 2021

Appendix

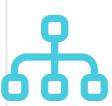
B

Information

Although no machine translation engine has been able to produce the best translation results for all industry and language pairs, Lugath delivers the best-known machine translation experience.



Access to All MT Services



SDK and API Support for Digital Platforms



Documents & Files Translation



Software Localization **Bundles Management**



Intelligent Forwarding & Fallback



Automatic
Translation Memory

for more information, info@lugath.com