

Challenging the state-of-the-art Machine Translation Metrics from a Linguistic Perspective

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Abstract

We employ a linguistically motivated challenge set in order to evaluate the state-of-the-art machine translation metrics submitted to the Metrics Shared Task of the 8th Conference for Machine Translation. The challenge set includes about 21,000 items extracted from 155 machine translation systems for three language directions (German \leftrightarrow English, English \rightarrow Russian), covering more than 100 linguistically-motivated phenomena organized in 14 categories. The metrics that have the best performance with regard to our linguistically motivated analysis are the COMETOID22-WMT23 (a trained metric based on distillation) for German-English and METRICX-23-C (based on a fine-tuned mT5 encoder-decoder language model) for English-German and English-Russian. Some of the most difficult phenomena are *passive voice* for German-English, *named entities*, *terminology* and *measurement units* for English-German and *focus particles*, *adverbial clause* and *stripping* for English-Russian.

1 Introduction

Most NLP evaluation has relied for years on testing the system performance on randomly picked test sets and producing a single generic score. Yet, machine learned systems learn to make abstractions and due to these, phenomena who are on the long tail of the training and test data may be overlooked hidden behind a very high generic score. Additionally, generic scores are often helpful to show relative improvement and reflect overall quality, but cannot explain the performance in a comprehensive way.

For example, old-style machine translation (MT) metrics measuring lexical overlap would equally penalize the omission of an article and the omission of the particle forming the negation in a sentence, although negation is more crucial for its meaning. While the evaluation of so obvious errors has been

addressed by the trained MT metrics, their evaluation relies on correlations with human judgments on randomly picked test-sets. In this case, a single correlation score may not be able to explain the strengths and weaknesses of the metrics with regard to the functioning of language.

Motivated by these considerations, we employ a multifold test set with linguistically-motivated challenges that will allow us to understand the metric performance from a linguistic perspective. These challenges are organized in smaller sets, one set per phenomenon, whereas the phenomena are organized in broader categories. By measuring the ability of the metrics to detect the errors in these challenge sets, we can get scores that indicate different aspects of linguistic performance.

This paper describes the application of such a challenge set on the evaluation of the MT metrics submitted at the relevant shared task of the 8th Conference of Machine Translation (Freitag et al., 2023). The rest of the paper is structured as following: Section 2 describes related work, and section 3 describes the way the challenges were selected. In Section 4 the results are presented and described, first from the perspective of metric comparison and then focusing on the performance for particular linguistically-motivated categories and phenomena per language direction. Some conclusions are given in Section 5.

2 Related work

There has been a growing interest for more fine-grained evaluation of Natural Language Processing (NLP) tools, as shown by the increasing number of publications many of whom have received distinctions (Ribeiro et al., 2020; Avelino et al., 2022; Campolungo et al., 2022). Concerning machine translation (MT), initial efforts were made in the 1990s with the introduction of test suites (King and Falkedal, 1990), and these efforts have been revitalized in light of recent advancements in the

field (Guillou and Hardmeier, 2016). To the best of our knowledge, the first endeavours related to the use of challenge sets in a meta-level in order to evaluate MT metrics were applied to Quality Estimation metrics (Avramidis et al., 2018), based on the first version of our linguistically-motivated test suite (Macketanz et al., 2018). The analysis was broadened to cover a broader range of MT metrics, including reference-based ones, as appeared in the Findings paper of the Metrics shared task of the 6th Conference on Machine Translation (Freitag et al., 2021), which was based on a later version of our test suite on German-English (Avramidis et al., 2019, 2020; Macketanz et al., 2021, 2022a), a resource also employed in this paper.

With the occasion of the first challenge set sub-task for the metrics shared task of the 7th Conference on Machine Translation (Freitag et al., 2022), a few more challenge sets emerged. ACES (Amrhein et al., 2022) for example, focuses on 68 accuracy errors. Similarly, Alves et al. (2022) evaluate the robustness of MT metrics by generating translations with critical errors. In a more linguistic direction, Chen et al. (2022) examine the capability of the metrics to correlate synonyms in different areas and to discern catastrophic errors at both word- and sentence-levels.

Our submission at that sub-task (Avramidis and Macketanz, 2022) augmented the preliminary analysis appearing at Freitag et al. (2021) by adding the language direction of English-German and presenting a more fine-grained analysis, not only in the category level but also on the phenomenon level. This year’s submission, explained on our paper, includes that same challenge set as last year, whereas English-Russian has been added as an additional language direction.

3 Method

3.1 Test suite for MT systems

Here, we are going to explain how we created the pool of MT sentences that were used for the challenge set. The selection was based on a linguistically-motivated test suite (Macketanz et al., 2022a)¹. The test suite contains a set of source sentences focusing on particular phenomena, each of them accompanied by some rules or regular expressions that can detect which translations would be accepted for these source sentences. This allows a

¹<https://github.com/DFKI-NLP/mt-testsuite>

semi-automatic evaluation when new translations are provided, whereas a human annotator resolves cases not covered by the rules.

For this experiment, we employed the test suite on three language directions: German-English (Avramidis et al., 2020), English-German (Macketanz et al., 2021) and English-Russian (Macketanz et al., 2022b). The German-English side consists of 5,539 German test sentences covering 107 linguistically motivated phenomena, the English-German side consists of 4,782 English test sentences covering 126 phenomena, and the English-Russian side consists of 1,225 English test sentences covering 64 phenomena. All language directions are organized in 14 categories, which nevertheless differ among the directions.

The above described test suite has been used to evaluate the outputs of 116 German-English, 29 English-German systems and 10 English-Russian systems submitted at the translation task of the Conference of Machine Translation (WMT). German-English outputs were collected from systems submitted in the years 2018-2021, English-German outputs in the years 2020-2021 and English-Russian in 2022.

3.2 Challenge set for MT metrics

The sentences selected with the help of the test suite are consequently used to create the challenge set. The source sentences and the system outputs have to be organized in contrastive pairs of correct/incorrect translations and a reference. In order to achieve this, for every source sentence from the test suite selection we create a challenge item including:

- one correct translation to be used as a reference translation,
- another correct translation to be used as the first translation candidate
- one incorrect translation to be used as the contrastive translation candidate

The two candidate translations and the reference consist one challenge item. Since source and translations were collected as a result of testing for a particular phenomenon, the same phenomenon will be what the challenge item will test.

Given that we may have many correct and wrong translations for the same source, the reference and the translations of the challenge items result from random combinations of correct and wrong translations from the collected WMT outputs. Therefore,

the same source sentence may appear many times.

As a result, we get a challenge set with 10,402 items for German-English, 8,945 items for English-German and 1,727 items for English-Russian.

3.3 Evaluation of metrics

For each challenge item, the two machine translation (MT) outputs, are provided to the metrics as separate MT hypotheses. Which output is correct, and which is incorrect, is hidden from the metrics. These hypotheses are then evaluated against the previously mentioned reference and/or the source. An item is deemed correctly scored when the metric assigns a higher score to the correct MT output compared to the incorrect one. Following this, the statistics below are computed:

- i) **Accuracy per Phenomenon:** the ratio of all correctly-scored challenge items per phenomenon to the total number of challenge items for that particular phenomenon.
- ii) **Accuracy per Category:** the ratio of all correctly-scored challenge items per category to the total number of challenge items for that category, after consolidating the underlying phenomena of that category into a single set.

Significance tests are performed to compare the highest metric accuracy for each phenomenon with all other metric accuracies for the same phenomenon. This is a one-tailed Z-test, conducted with a significance level of $\alpha = 0.95$. Metrics with accuracies that are not significantly worse than the highest accuracy are considered to share the top position for that phenomenon. A similar approach is used to identify the best accuracies per category, after aggregating the challenge items from the underlying phenomena within each category.

Metric categories We conduct this significance testing in two stages: first, for all metrics involved in the shared task, and then separately for each of the three metric categories (baseline, Quality Estimation (QE) as a metric, reference-based). Systems that are significantly superior per phenomenon across all metrics are highlighted with a gray background, while those that are significantly superior per metric category are denoted in boldface.

Averaging Lastly, we provide three types of averaging scores:

- i) **Micro-average:** This approach considers all items equally, aggregating all test items to compute the average percentages.
- ii) **Category macro-average:** Here, all categories are treated equally, with the percentages being computed independently for each category and then averaged.
- iii) **Phenomenon macro-average:** This average treats all phenomena equally, with the percentages being computed independently for each phenomenon and then averaged.

4 Results

The results are displayed in detail in Tables 1, 2 and 3 for the category level and in Tables 4, 5 and 6 for the phenomenon level, for the three language directions respectively.

4.1 Metric performance analysis

Here we are observing the statistics with a focus on comparing the performance of various metrics on the challenge set.

German-English The accuracies of the metrics, as measured for several categories in German-English, can be seen in Table 1. The best performing metric for German-English is COMETOID22-WMT23 (Gowda et al., 2023), which, wins significantly based on both the micro-average (83%) and the macro-average (87%). This metric is a distilled QE model that has been trained on COMET (Rei et al., 2020) scores of WMT outputs, including the ones of WMT23. For this reason, we include it into the reference-aware metrics. We notice that its performance among the other metrics is impressive. It is the first metric in 6 categories and among them the only one who wins at *Verb tense/aspect/mood* and *function words*, achieving 93% and 91% accuracy respectively.

Another two reference-based baseline metrics, COMET and PRISMREF (Thompson and Post, 2020a,b) share the first position when the category macro-average is considered (82%). None of the other reference-aware metrics submitted this year managed to compete with the metrics with the highest accuracy mentioned above.

The lowest performing metric is the reference-less random baseline RANDOM-SYSNAME, provided by the organizers (44%), followed by XL-SIMQE (55-58%; Mukherjee and Shrivastava,

2023) and MATESE (57-58%; Perrella et al., 2022).

When considering the metric performance with regard to particular categories, one can see, again this year, that different metrics win in different combinations of categories. Here, only COMETOID22-WMT23 as mentioned above, wins 6 metrics, followed by PRISMREF and METRIC-23-C, which win 4 categories. 17 metrics do not win any category, ranging in accuracies around 75%.

English-German The accuracies of the metrics, as measured for several categories in English-German, can be seen in Table 2. The best performing metric in English-German is METRICX-23-C, which is in the first significance cluster based on both the micro-average (81%) and the category macro-average (84%). This metric uses the mT5 encoder-decoder language model, which is fine-tuned using direct assessment data, MQM (Lommel et al., 2014) data and synthetic data. The categories to which its success may be mostly attributed are the *multi-word expressions* (MWE; with 92%) and the non-verbal agreement (95%).

Another three metrics share the first position, when the micro average is considered, namely the QE version of the latter, *MetricX-23-QE-c* and also *mbr-metricx-qe* (Naskar et al., 2023) and XCOMET-Ensemble. It is impressive that QE methods manage to reach high accuracy without access to reference content.

When looking at the worse-performing metrics, MATESE here performs worse than the baseline (36-38%), followed by PARTOKENGRAM_F (55-56%; Dreano et al., 2023b).

In English-German it is even harder to say which metrics perform well in multiple categories, as only one of them, XCOMET-QE-ENSEMBLE, achieves the highest performance in 3 categories (*function words, non-verbal agreement and subordination*). The rest of the metrics show a good performance in 2 categories or fewer.

English-Russian The accuracies of the metrics, as measured for several categories in English-German, can be seen in Table 3. For this language pair, variants of the MetricX achieve significantly higher accuracies than all the other metrics. In particular, METRICX-23-C achieves the highest accuracy based on both micro-average and category macro-average, whereas METRICX-23-B and METRICX-23-QE-C achieve a slightly

lower macro-average, which is nevertheless not significantly worse than the one of the former. MATESE is again by far the lowest performing metric (32/34%), lower than the random baseline. We may assume that this metric has not been optimized for this language direction.

4.2 Linguistically motivated analysis

In this section, we are focusing on the results for particular phenomena or categories.

4.2.1 German-English

Category-level The overall average accuracy of all metrics with regard to the linguistically motivated categories is at 76% for German-English, which is two percentage points lower than last year’s average. It is still a fact, that the metrics in average fail to predict properly the scores for one out of four challenge items that we provided. Luckily, there has been noticeable accuracy for some categories, for example METRICX reference-based variants achieved an accuracy of 96% for *false friends*, whereas *negation* errors have been scored correctly with a 98.5

The worse performing category is *Verb valency*, where the best metrics achieved only 66% accuracy, and the rest of the metrics averaged to a mere 56%. In this category one can observe the lowest accuracy, given by an LLM-based metric, EMBED_LLAMA (Dreano et al., 2023a) with 41%.

Phenomenon-level The best accuracy for this language pair (Table 4) is achieved this year at several variations of verb tenses, i.e. *Transitive - future II, Modal negated - present, Reflexive - preterite subjunctive II* and *Intransitive - pluperfect* which get more than 85% in average.

The lowest accuracy of all metrics in average is given for *passive voice*, where the highest accuracy achieved by several metrics is only 54.5%. Errors related to *commas, domain-specific terms* and *locations* have also been scored with a less than 65% accuracy.

4.2.2 English-German

Category-level The overall average accuracy of all metrics with regard to the linguistically motivated categories is at 71-73% for English-German. The category where all metrics perform better in average is *negation* (83%), where 11 of the metrics achieve more than 90% accuracy. Negation is closely followed by *function words Non-verbal agreement* (80%).

The worse performing category in average is *named entities and terminology* (58,8%), where most metrics' accuracies are close to 50%, except for BLEURT (Yan et al., 2023) (80.3%). The rest of the categories lie in rather mediocre accuracies, between 58.8% and 80%.

Phenomenon-level The English-German phenomena, where metrics perform best in average (Table 5) are the *transitive conditional II simple, gerunds, contact clause* and the *intransitive present perfect simple*, achieving more than 85% of accuracy. The phenomena which incur the lowest average accuracies are the *transitive present progressive, measuring units, modals* and *intransitive - future II progressive* with less than 50% accuracy. The former and the latter were observed as the most difficult phenomena to score also last year.

4.3 English-Russian

This analysis for English-Russian occurs for the first time this year, based on the MT outputs collected at last year's shared task. For this purpose the test instances are much fewer than the other language pairs and therefore the numbers are not very conclusive. Therefore, categories and phenomena that have only a handful of samples will not be included in our analysis, although they appear in the tables.

Category-level Here, the average accuracy over all metrics is much lower than the other language directions, reaching only 66%, only 20% above the random baseline. The best performing category is *ambiguity* (86,3%), more than 13% better than the following categories. The worst performing categories are *function words* and *punctuation*, with less than 55%. The rest of the categories range in accuracies between 53 and 73%.

Phenomenon-level The good performance of the *ambiguity* category is also confirmed in the table on the phenomenon level (Table 6), as in Russian this is the only phenomenon of this category, as opposed to other language pairs where we also have examples of structural ambiguity. The most difficult phenomena to score appear to be the *focus particles, adverbial clause* and *stripping* with less than 50% average accuracy, in many cases lower than the random baseline.

5 Conclusion

In this paper we analysed the performance of several state-of-the-art metrics with regard to particular linguistically-motivated phenomena for three language pairs, German-English, English-German and for the first time, English-Russian. The analysis gave a multitude of observations, regarding both the performance of the metrics and the corresponding linguistic observations.

The metrics demonstrating the best performance in average are COMETOID22-WMT23 for the German-English language pair, and METRICX-23-C for both the English-German and English-Russian language pairs. Quality estimation methods have impressively good performance in several phenomena. Some metrics that report usage of LLMs (EMBED_LLAMA) have not scored very high in overall, indicating that more work is required in this direction.

Among the various linguistic phenomena, we could identify some of the particularly challenging ones. In German-English, metrics have difficulties scoring the *passive voice* properly. In English-German *named entities and terminology* as well as specific *measurement units* pose the most difficulties. In English-Russian translation, translations with *focus particles, adverbial clause, and stripping* phenomena emerge as particularly complex challenges.

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| ling. category | baselines | | | | | | | | | | | | | | QE as a metric | | | | | | | | | | | | | | ref. based metrics | | | | | | | | | | | | | |
|-------------------------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|-----------|--------------|---------------|-----------|-----------|-------------|-----------------|-----------------|---------------|----------------|--------------------|-----------|-----------------|-----------------|-----------|-----------|-----------|---------------|-------|--------------|--------------|--------------------|-----------------|-----------|------------|-----------|-----------------|-----------|-------------|----------------|-------------|-----|----|--|--|
| | BERTscore | BLEU | BLEURT-20 | COMET | YSI-1 | chrF | prtmRef | spBLEU | CometKiwi-XL | CometKiwi-XXL | CometKiwi | GEMBA-MQM | MS-COMET-QE | MetricX-23-QE-b | MetricX-23-QE-c | MetricX-23-QE | Random-synname | XCOMET-QE-Ensemble | XLsimQE | cometoid22-wm11 | cometoid22-wm12 | prtmSrc2 | KG-BERT | MEE4 | MEE4_sisb_xlm | MaTSE | MetricX-23-b | MetricX-23-c | MetricX-23 | XCOMET-Ensemble | XCOMET-XL | XCOMET-XXL | XLsim | cometoid22-wm13 | eBLEU | embed_llama | partokengram_F | tokengram_F | avg | | | |
| Ambiguity | 87 | 71 | 89 | 90 | 90 | 81 | 91 | 79 | 78 | 90 | 85 | 77 | 82 | 78 | 86 | 80 | 46 | 87 | 61 | 86 | 85 | 66 | 85 | 83 | 88 | 74 | 87 | 90 | 90 | 88 | 89 | 88 | 92 | 84 | 91 | 74 | 56 | 48 | 80 | 81 | | |
| Composition | 90 | 66 | 88 | 87 | 88 | 74 | 86 | 71 | 83 | 87 | 80 | 81 | 75 | 85 | 90 | 81 | 44 | 86 | 40 | 81 | 81 | 70 | 80 | 85 | 86 | 69 | 85 | 90 | 85 | 87 | 84 | 83 | 79 | 87 | 74 | 65 | 61 | 75 | 79 | | | |
| Coordination & ellipsis | 316 | 81 | 74 | 81 | 78 | 83 | 78 | 82 | 74 | 83 | 75 | 84 | 69 | 77 | 77 | 78 | 45 | 79 | 59 | 75 | 73 | 79 | 84 | 81 | 77 | 49 | 81 | 82 | 81 | 81 | 79 | 70 | 77 | 81 | 73 | 68 | 58 | 78 | 75 | | | |
| False friends | 90 | 64 | 91 | 92 | 88 | 76 | 84 | 69 | 82 | 82 | 82 | 77 | 72 | 89 | 90 | 88 | 47 | 80 | 51 | 87 | 83 | 86 | 82 | 87 | 80 | 73 | 96 | 96 | 84 | 84 | 74 | 74 | 93 | 72 | 68 | 51 | 72 | 80 | | | | |
| Function word | 586 | 82 | 73 | 83 | 83 | 82 | 73 | 84 | 73 | 84 | 87 | 76 | 77 | 68 | 80 | 86 | 47 | 86 | 60 | 84 | 87 | 81 | 76 | 79 | 76 | 57 | 87 | 83 | 86 | 87 | 83 | 72 | 80 | 91 | 79 | 66 | 55 | 75 | 78 | | | |
| LDD & interrogatives | 1014 | 84 | 74 | 85 | 87 | 84 | 76 | 86 | 74 | 80 | 81 | 84 | 72 | 81 | 83 | 83 | 45 | 80 | 61 | 81 | 79 | 72 | 84 | 81 | 76 | 53 | 82 | 83 | 82 | 84 | 82 | 72 | 77 | 87 | 76 | 68 | 54 | 77 | 77 | | | |
| MWE | 610 | 82 | 72 | 84 | 84 | 85 | 76 | 86 | 73 | 74 | 83 | 77 | 73 | 59 | 80 | 87 | 45 | 83 | 53 | 72 | 73 | 65 | 76 | 80 | 81 | 67 | 84 | 86 | 87 | 84 | 84 | 80 | 75 | 85 | 75 | 66 | 54 | 76 | 76 | | | |
| Named entity & termin. | 861 | 73 | 64 | 69 | 74 | 76 | 69 | 73 | 72 | 65 | 67 | 66 | 61 | 62 | 65 | 70 | 66 | 44 | 67 | 59 | 66 | 67 | 69 | 66 | 72 | 70 | 58 | 72 | 74 | 72 | 72 | 75 | 66 | 69 | 71 | 64 | 57 | 52 | 70 | 67 | | |
| Negation | 76 | 79 | 78 | 91 | 93 | 84 | 82 | 84 | 76 | 91 | 91 | 93 | 75 | 84 | 91 | 95 | 86 | 50 | 93 | 55 | 91 | 89 | 92 | 93 | 86 | 84 | 51 | 91 | 95 | 92 | 91 | 79 | 76 | 91 | 75 | 59 | 45 | 82 | 82 | | | |
| Non-verbal agreement | 419 | 78 | 78 | 85 | 84 | 78 | 78 | 79 | 79 | 78 | 80 | 77 | 67 | 78 | 78 | 81 | 79 | 43 | 80 | 64 | 77 | 79 | 74 | 76 | 78 | 75 | 59 | 81 | 84 | 81 | 77 | 72 | 70 | 81 | 76 | 60 | 74 | 77 | 76 | | | |
| Punctuation | 293 | 75 | 74 | 69 | 69 | 70 | 65 | 80 | 76 | 62 | 62 | 63 | 59 | 59 | 61 | 60 | 39 | 65 | 67 | 66 | 66 | 86 | 63 | 76 | 67 | 48 | 62 | 55 | 62 | 66 | 66 | 44 | 69 | 67 | 65 | 66 | 67 | 67 | 65 | | | |
| Subordination | 679 | 75 | 70 | 76 | 78 | 75 | 70 | 78 | 69 | 79 | 78 | 78 | 64 | 65 | 77 | 77 | 78 | 43 | 79 | 53 | 75 | 75 | 68 | 78 | 77 | 75 | 50 | 77 | 78 | 77 | 79 | 75 | 64 | 70 | 79 | 73 | 62 | 71 | 71 | 72 | | |
| Verb tense/aspect/mood | 4697 | 88 | 69 | 85 | 88 | 89 | 76 | 90 | 71 | 83 | 84 | 85 | 84 | 81 | 85 | 87 | 81 | 43 | 82 | 60 | 74 | 75 | 80 | 85 | 84 | 84 | 56 | 82 | 87 | 83 | 85 | 83 | 78 | 78 | 93 | 76 | 62 | 77 | 77 | 79 | | |
| Verb valency | 211 | 64 | 54 | 63 | 64 | 64 | 55 | 65 | 52 | 59 | 54 | 59 | 45 | 51 | 60 | 58 | 55 | 39 | 57 | 33 | 57 | 48 | 59 | 62 | 56 | 44 | 62 | 63 | 61 | 62 | 61 | 53 | 61 | 66 | 55 | 41 | 56 | 56 | 56 | | | |
| macro avg. | 10402 | 81 | 70 | 81 | 82 | 81 | 73 | 82 | 72 | 77 | 79 | 78 | 70 | 78 | 81 | 77 | 44 | 79 | 55 | 76 | 77 | 74 | 78 | 79 | 77 | 58 | 81 | 82 | 81 | 81 | 79 | 71 | 74 | 83 | 72 | 62 | 59 | 74 | 74 | | | |
| micro avg. | 10402 | 83 | 70 | 82 | 84 | 84 | 75 | 85 | 72 | 79 | 80 | 80 | 75 | 74 | 80 | 83 | 78 | 44 | 80 | 58 | 75 | 76 | 75 | 80 | 81 | 79 | 57 | 81 | 83 | 81 | 82 | 81 | 74 | 76 | 87 | 74 | 63 | 67 | 75 | 76 | | |

Table 1: Accuracy of the metrics (%) with regard to the 14 linguistically motivated categories for German-English. The significantly best systems per phenomenon over all metrics are indicated with a gray background, whereas the significantly best systems per metrics category are indicated with boldface.

| ling. category | # | baselines | | | | | | | | | | | | QE as a metric | | | | | | | | | | | | ref. based metrics | | | | | | | | | | | | | | | | | | | | |
|-------------------------|------|-----------|------|-----------|-------|-------|------|----------|--------|-------------------|-------------|--------------|----------|----------------|---------|-------------|-----------------|-----------------|---------------|----------------|--------------------|---------|----------------|----------------|----------------|------------------------|-----------------------|-----------|----------------|------|--------------|--------|--------------|--------------|------------|-----------------|-----------|------------|-------|-------|-------------|--------------|----------------|-------------|-----|----|
| | | BERTscore | BLEU | BLEURT-20 | COMET | Ys1-1 | chrf | prismRef | spBLEU | Calibri-COMET2-QE | CometKwi-XL | CometKwi-XXL | CometKwi | GEMBA-MQM | KG-BERT | MS-COMET-QE | MetricX-23-QE-b | MetricX-23-QE-c | MetricX-23-QE | Random-sysname | XCOMET-QE-Ensemble | XLSimQE | cometoid2-wm11 | cometoid2-wm12 | cometoid2-wm13 | mbr-metricx-qe0p2p1-qe | mbr-metricx-qexv1p-qe | prismSrc2 | Calibri-COMET2 | MEB4 | MEF4_sbs_xlm | MATESe | MetricX-23-b | MetricX-23-c | MetricX-23 | XCOMET-Ensemble | XCOMET-XL | XCOMET-XXL | XLSim | eBLEU | embed_llama | mftRegressor | partokengram_F | tokengram_F | avg | |
| Ambiguity | 146 | 84 | 71 | 90 | 84 | 86 | 88 | 97 | 88 | 42 | 77 | 60 | 42 | 77 | 42 | 39 | 75 | 85 | 82 | 46 | 56 | 38 | 59 | 58 | 87 | 86 | 86 | 95 | 14 | 84 | 90 | 92 | 50 | 88 | 95 | 88 | 82 | 88 | 62 | 93 | 70 | 62 | 81 | 62 | 88 | 73 |
| Coordination & ellipsis | 836 | 69 | 61 | 80 | 78 | 69 | 61 | 74 | 63 | 73 | 74 | 75 | 74 | 60 | 74 | 71 | 78 | 79 | 80 | 48 | 77 | 38 | 75 | 74 | 76 | 83 | 80 | 58 | 75 | 62 | 63 | 33 | 76 | 81 | 72 | 79 | 81 | 64 | 75 | 61 | 56 | 72 | 49 | 62 | 69 | |
| False friends | 225 | 65 | 63 | 69 | 74 | 71 | 71 | 67 | 65 | 71 | 61 | 75 | 71 | 65 | 71 | 73 | 64 | 80 | 66 | 46 | 76 | 82 | 64 | 52 | 66 | 76 | 63 | 68 | 74 | 83 | 87 | 30 | 60 | 82 | 62 | 65 | 59 | 53 | 76 | 66 | 55 | 56 | 52 | 68 | 67 | |
| Function word | 200 | 90 | 78 | 79 | 90 | 82 | 74 | 93 | 74 | 92 | 94 | 92 | 90 | 82 | 90 | 78 | 86 | 72 | 85 | 36 | 95 | 62 | 94 | 94 | 94 | 90 | 67 | 84 | 86 | 82 | 76 | 60 | 86 | 86 | 84 | 94 | 86 | 64 | 76 | 70 | 52 | 76 | 60 | 76 | 80 | |
| MWE | 829 | 79 | 72 | 87 | 90 | 86 | 77 | 85 | 74 | 74 | 86 | 79 | 73 | 66 | 73 | 71 | 88 | 89 | 85 | 45 | 83 | 37 | 78 | 78 | 90 | 88 | 89 | 47 | 89 | 76 | 78 | 18 | 86 | 92 | 85 | 90 | 93 | 76 | 81 | 69 | 69 | 72 | 54 | 78 | 76 | |
| Named entity & termin. | 1272 | 57 | 54 | 66 | 64 | 62 | 61 | 67 | 62 | 54 | 51 | 49 | 55 | 56 | 54 | 55 | 61 | 65 | 58 | 44 | 54 | 46 | 60 | 54 | 68 | 62 | 80 | 56 | 62 | 60 | 63 | 30 | 72 | 74 | 71 | 62 | 60 | 60 | 71 | 60 | 48 | 54 | 49 | 60 | 59 | |
| Negation | 174 | 87 | 83 | 89 | 92 | 84 | 84 | 86 | 84 | 90 | 90 | 84 | 89 | 76 | 89 | 92 | 83 | 88 | 78 | 47 | 84 | 40 | 90 | 90 | 96 | 89 | 91 | 90 | 91 | 91 | 58 | 85 | 91 | 80 | 84 | 85 | 79 | 80 | 86 | 67 | 88 | 52 | 86 | 83 | | |
| Non-verbal agreement | 372 | 74 | 72 | 81 | 88 | 78 | 70 | 83 | 75 | 82 | 90 | 84 | 81 | 83 | 81 | 78 | 95 | 93 | 91 | 45 | 95 | 43 | 93 | 91 | 93 | 82 | 94 | 71 | 87 | 76 | 69 | 57 | 90 | 95 | 87 | 94 | 91 | 88 | 73 | 72 | 65 | 84 | 54 | 69 | 80 | |
| Punctuation | 336 | 69 | 73 | 74 | 70 | 66 | 72 | 77 | 69 | 82 | 72 | 76 | 82 | 60 | 82 | 65 | 73 | 66 | 73 | 42 | 60 | 44 | 81 | 82 | 81 | 75 | 70 | 75 | 65 | 74 | 72 | 28 | 74 | 70 | 67 | 65 | 65 | 46 | 68 | 74 | 49 | 57 | 44 | 73 | 68 | |
| Subordination | 994 | 78 | 74 | 81 | 84 | 78 | 75 | 86 | 74 | 89 | 88 | 89 | 76 | 89 | 80 | 83 | 85 | 83 | 86 | 83 | 84 | 76 | 83 | 79 | 83 | 86 | 83 | 84 | 76 | 83 | 79 | 78 | 16 | 83 | 84 | 83 | 89 | 85 | 79 | 75 | 71 | 75 | 56 | 76 | 78 | |
| Verb tense/aspect/mood | 3081 | 68 | 62 | 70 | 70 | 69 | 69 | 67 | 64 | 71 | 72 | 75 | 75 | 71 | 75 | 61 | 82 | 84 | 85 | 43 | 83 | 52 | 62 | 65 | 75 | 71 | 77 | 61 | 70 | 70 | 74 | 46 | 77 | 77 | 78 | 83 | 76 | 71 | 69 | 72 | 55 | 71 | 62 | 69 | 70 | |
| Verb valency | 480 | 73 | 64 | 82 | 79 | 74 | 70 | 79 | 69 | 77 | 81 | 85 | 77 | 62 | 77 | 71 | 86 | 80 | 86 | 42 | 87 | 52 | 79 | 80 | 77 | 87 | 88 | 66 | 77 | 70 | 36 | 84 | 81 | 85 | 85 | 80 | 75 | 79 | 66 | 69 | 65 | 70 | 70 | 74 | | |
| macro avg. | 8945 | 74 | 69 | 79 | 80 | 76 | 73 | 80 | 72 | 75 | 78 | 77 | 75 | 70 | 75 | 70 | 80 | 80 | 79 | 44 | 78 | 48 | 76 | 75 | 82 | 81 | 81 | 64 | 79 | 76 | 76 | 38 | 80 | 84 | 79 | 81 | 79 | 68 | 76 | 70 | 60 | 71 | 55 | 73 | 73 | |
| micro avg. | 8945 | 70 | 65 | 75 | 76 | 72 | 69 | 74 | 68 | 73 | 74 | 75 | 74 | 68 | 74 | 67 | 79 | 81 | 80 | 44 | 78 | 47 | 71 | 71 | 78 | 77 | 81 | 62 | 74 | 71 | 73 | 36 | 79 | 81 | 78 | 81 | 77 | 70 | 73 | 69 | 59 | 69 | 56 | 70 | 71 | |

Table 2: Accuracy of the metrics (%) with regard to the 12 linguistically motivated categories for English-German

| ling. category | baselines | | | | | | | | | | | | QE as a metric | | | | | | | | | | | | ref. based metrics | | | | | | | | | | | | | |
|-------------------------|------------|-----------|------------|------------|------------|-----------|-----------|-----------|--------------|---------------|------------|------------|----------------|-------------|-----------------|-----------------|---------------|----------------|--------------------|-----------|-----------------|-----------------|-----------------|-----------|--------------------|--------------|--------------|------------|-----------------|------------|------------|------------|-----------|-------------|----------------|-------------|-----|----|
| | BERTscore | BLEU | BLEURT-20 | COMET | YiS-1 | chrf | prismRet | spBLEU | CometKiwI-XL | CometKiwI-XXL | CometKiwI | GEMBA-MQM | KG-BERT | MS-COMET-QE | MetricX-23-QE-b | MetricX-23-QE-c | MetricX-23-QE | Random-sysname | XCOMET-QE-Ensemble | XLsimQE | cometoid22-wm11 | cometoid22-wm12 | cometoid22-wm13 | prismSrc2 | MATSE | MetricX-23-b | MetricX-23-c | MetricX-23 | XCOMET-Ensemble | XCOMET-XL | XCOMET-XXL | XLsim | eBLEU | embed_llama | partokengram_F | tokengram_F | avg | |
| Ambiguity | 80 | 67 | 100 | 97 | 88 | 67 | 83 | 73 | 100 | 94 | 100 | 100 | 100 | 95 | 100 | 100 | 100 | 39 | 100 | 30 | 100 | 100 | 100 | 56 | 98 | 100 | 100 | 100 | 100 | 100 | 100 | 88 | 77 | 44 | 59 | 70 | 86 | |
| Coordination & ellipsis | 84 | 73 | 68 | 73 | 76 | 67 | 77 | 69 | 73 | 64 | 68 | 38 | 68 | 76 | 62 | 79 | 64 | 45 | 64 | 41 | 70 | 67 | 69 | 55 | 21 | 61 | 69 | 56 | 67 | 64 | 50 | 74 | 80 | 60 | 57 | 67 | 64 | |
| False friends | 100 | 17 | 100 | 100 | 100 | 17 | 83 | 50 | 0 | 0 | 100 | 0 | 0 | 0 | 0 | 100 | 17 | 17 | 0 | 17 | 33 | 33 | 100 | 0 | 17 | 100 | 100 | 100 | 100 | 33 | 33 | 100 | 83 | 100 | 50 | 83 | 33 | 50 |
| Function word | 59 | 69 | 69 | 59 | 53 | 38 | 48 | 72 | 53 | 55 | 50 | 51 | 50 | 36 | 66 | 55 | 48 | 44 | 72 | 39 | 73 | 64 | 54 | 28 | 9 | 63 | 57 | 60 | 60 | 60 | 60 | 22 | 49 | 61 | 52 | 53 | 38 | 53 |
| MWE | 122 | 70 | 64 | 79 | 80 | 74 | 64 | 79 | 72 | 70 | 77 | 57 | 77 | 77 | 75 | 88 | 75 | 40 | 67 | 35 | 75 | 70 | 77 | 23 | 57 | 87 | 91 | 85 | 70 | 73 | 69 | 75 | 66 | 59 | 59 | 64 | 69 | |
| Named entity & termin. | 243 | 79 | 69 | 95 | 91 | 79 | 76 | 85 | 84 | 62 | 59 | 54 | 59 | 69 | 63 | 73 | 73 | 49 | 64 | 53 | 85 | 86 | 95 | 31 | 51 | 89 | 86 | 86 | 91 | 91 | 71 | 84 | 75 | 65 | 58 | 75 | 73 | |
| Negation | 34 | 74 | 74 | 79 | 68 | 74 | 85 | 74 | 68 | 85 | 85 | 91 | 85 | 41 | 71 | 100 | 9 | 59 | 91 | 9 | 62 | 29 | 59 | 91 | 0 | 94 | 71 | 94 | 79 | 71 | 74 | 65 | 76 | 76 | 85 | 85 | 69 | |
| Non-verbal agreement | 61 | 69 | 56 | 82 | 77 | 67 | 61 | 57 | 84 | 100 | 54 | 72 | 54 | 74 | 72 | 98 | 66 | 41 | 77 | 26 | 95 | 100 | 85 | 62 | 43 | 92 | 92 | 85 | 79 | 31 | 64 | 61 | 66 | 69 | 69 | 70 | | |
| Punctuation | 121 | 42 | 46 | 53 | 68 | 86 | 41 | 73 | 66 | 29 | 65 | 27 | 65 | 87 | 55 | 66 | 43 | 41 | 61 | 44 | 66 | 67 | 36 | 56 | 1 | 60 | 71 | 52 | 66 | 57 | 12 | 63 | 66 | 84 | 45 | 45 | 54 | |
| Subordination | 499 | 62 | 62 | 61 | 70 | 67 | 66 | 63 | 72 | 66 | 66 | 41 | 66 | 75 | 65 | 66 | 61 | 45 | 77 | 32 | 68 | 61 | 65 | 37 | 18 | 60 | 84 | 58 | 80 | 70 | 52 | 63 | 54 | 57 | 65 | 65 | 61 | |
| Verb tense/aspect/mood | 135 | 79 | 62 | 76 | 85 | 76 | 84 | 81 | 79 | 79 | 79 | 73 | 61 | 73 | 64 | 67 | 65 | 66 | 39 | 74 | 60 | 70 | 59 | 85 | 38 | 50 | 94 | 90 | 85 | 92 | 74 | 79 | 70 | 73 | 82 | 82 | 73 | |
| Verb valency | 121 | 70 | 67 | 70 | 75 | 76 | 67 | 68 | 71 | 88 | 79 | 66 | 79 | 71 | 76 | 85 | 76 | 36 | 76 | 26 | 63 | 61 | 65 | 25 | 45 | 79 | 81 | 78 | 72 | 70 | 65 | 84 | 76 | 64 | 69 | 69 | 68 | |
| macro avg. | 1727 | 72 | 60 | 78 | 79 | 76 | 61 | 71 | 66 | 69 | 65 | 63 | 65 | 64 | 64 | 81 | 58 | 41 | 69 | 34 | 72 | 66 | 74 | 42 | 34 | 81 | 83 | 79 | 74 | 72 | 60 | 73 | 72 | 62 | 65 | 64 | 66 | |
| micro avg. | 1727 | 69 | 64 | 72 | 76 | 73 | 65 | 70 | 67 | 74 | 67 | 68 | 51 | 68 | 71 | 74 | 64 | 44 | 72 | 39 | 73 | 68 | 71 | 40 | 32 | 73 | 82 | 71 | 77 | 74 | 55 | 71 | 67 | 62 | 63 | 66 | 66 | |

Table 3: Accuracy of the metrics (%) with regard to the linguistically motivated categories for English-Russian

| # | baselines | | | | | | | | | | QE as a metric | | | | | | | | | | ref. based metrics | | | | | | | | | | | | | | | | | | | |
|-----|-----------|------|--------|-------|-------|------|----------|--------|-------------|--------------|----------------|-----------|-------------|-----------------|-----------------|---------------|----------------|--------------------|---------|-----------------|--------------------|-----------|---------|------|--------------|--------|--------------|--------------|------------|-----------------|-----------|------------|-------|-----------------|-------|-------------|----------------|-------------|-----|----|
| | BERTscore | BLEU | BLUR20 | COMET | YIS-1 | chrF | prismRef | spBLEU | CometKwi-XL | CometKwi-XXL | CometKwi | GEMBA-MQM | MS-COMET-QE | MetricX-23-QE-b | MetricX-23-QE-c | MetricX-23-QE | Random-synname | XCOMET-QE-Ensemble | XLsimQE | cometoid22-wm11 | cometoid22-wm12 | prismSrc2 | KG-BERT | MEE4 | MEE4_snb_xlm | MATeSE | MetricX-23-b | MetricX-23-c | MetricX-23 | XCOMET-Ensemble | XCOMET-XL | XCOMET-XXL | XLsim | cometoid22-wm13 | eBLEU | embed_llama | partokengram_F | tokengram_F | avg | |
| 129 | 91 | 75 | 95 | 97 | 89 | 88 | 95 | 83 | 81 | 97 | 88 | 79 | 82 | 81 | 91 | 84 | 50 | 91 | 51 | 84 | 87 | 63 | 88 | 84 | 97 | 83 | 91 | 95 | 95 | 93 | 96 | 94 | 98 | 78 | 73 | 49 | 86 | 58 | | |
| 129 | 85 | 69 | 84 | 84 | 90 | 71 | 88 | 76 | 75 | 84 | 82 | 79 | 82 | 76 | 82 | 70 | 43 | 84 | 68 | 87 | 84 | 69 | 82 | 82 | 82 | 86 | 83 | 85 | 83 | 86 | 89 | 76 | 86 | 71 | 44 | 47 | 75 | 77 | | |
| 129 | 91 | 64 | 93 | 90 | 92 | 71 | 80 | 85 | 80 | 85 | 79 | 79 | 80 | 81 | 91 | 80 | 43 | 84 | 29 | 74 | 74 | 64 | 77 | 91 | 90 | 70 | 87 | 95 | 88 | 86 | 89 | 80 | 86 | 71 | 61 | 57 | 72 | 78 | | |
| 123 | 89 | 67 | 82 | 85 | 77 | 81 | 72 | 87 | 89 | 84 | 82 | 80 | 80 | 89 | 89 | 83 | 46 | 87 | 50 | 87 | 87 | 76 | 84 | 78 | 82 | 67 | 82 | 85 | 83 | 86 | 81 | 76 | 77 | 89 | 77 | 69 | 66 | 79 | 79 | |
| 61 | 86 | 69 | 88 | 88 | 84 | 73 | 90 | 67 | 96 | 96 | 92 | 75 | 78 | 88 | 94 | 90 | 49 | 94 | 51 | 82 | 86 | 76 | 84 | 79 | 69 | 45 | 96 | 90 | 94 | 94 | 86 | 73 | 94 | 80 | 65 | 57 | 73 | 82 | | |
| 67 | 81 | 69 | 72 | 69 | 84 | 73 | 84 | 69 | 75 | 57 | 84 | 76 | 61 | 75 | 58 | 69 | 52 | 66 | 42 | 55 | 52 | 76 | 84 | 79 | 68 | 45 | 81 | 78 | 73 | 73 | 72 | 64 | 70 | 75 | 72 | 70 | 55 | 73 | 69 | |
| 128 | 80 | 81 | 80 | 79 | 81 | 84 | 81 | 80 | 82 | 74 | 85 | 69 | 68 | 77 | 77 | 78 | 45 | 79 | 73 | 79 | 77 | 78 | 85 | 82 | 77 | 49 | 79 | 82 | 80 | 80 | 69 | 84 | 77 | 73 | 69 | 59 | 84 | 76 | 76 | |
| 70 | 80 | 71 | 84 | 76 | 84 | 74 | 77 | 73 | 84 | 79 | 74 | 59 | 64 | 73 | 81 | 80 | 37 | 80 | 57 | 81 | 77 | 83 | 74 | 77 | 81 | 40 | 74 | 81 | 79 | 79 | 74 | 64 | 73 | 71 | 66 | 59 | 76 | 73 | | |
| 90 | 80 | 64 | 91 | 92 | 88 | 76 | 84 | 69 | 82 | 82 | 72 | 72 | 72 | 89 | 90 | 88 | 47 | 80 | 51 | 87 | 83 | 86 | 82 | 87 | 80 | 73 | 96 | 96 | 84 | 84 | 74 | 74 | 93 | 72 | 68 | 51 | 72 | 80 | 78 | |
| 64 | 88 | 78 | 84 | 80 | 88 | 75 | 83 | 78 | 77 | 81 | 84 | 70 | 86 | 78 | 83 | 47 | 80 | 55 | 78 | 81 | 88 | 81 | 88 | 80 | 48 | 86 | 84 | 89 | 81 | 72 | 70 | 81 | 81 | 80 | 73 | 64 | 78 | 78 | | |
| 166 | 84 | 77 | 84 | 89 | 89 | 77 | 90 | 81 | 88 | 92 | 72 | 72 | 69 | 84 | 95 | 86 | 43 | 94 | 55 | 92 | 93 | 69 | 72 | 86 | 85 | 51 | 89 | 89 | 88 | 91 | 89 | 78 | 84 | 94 | 81 | 70 | 54 | 76 | 80 | |
| 356 | 80 | 70 | 82 | 81 | 78 | 71 | 81 | 69 | 83 | 86 | 76 | 78 | 67 | 77 | 83 | 77 | 49 | 83 | 63 | 82 | 85 | 85 | 72 | 74 | 71 | 61 | 86 | 81 | 84 | 86 | 70 | 78 | 92 | 78 | 63 | 54 | 74 | 76 | 76 | |
| 320 | 85 | 78 | 85 | 90 | 86 | 80 | 88 | 79 | 76 | 80 | 88 | 70 | 88 | 85 | 83 | 82 | 49 | 78 | 62 | 81 | 83 | 74 | 88 | 84 | 78 | 59 | 82 | 85 | 82 | 83 | 81 | 78 | 81 | 92 | 81 | 67 | 53 | 82 | 79 | |
| 92 | 77 | 75 | 79 | 80 | 77 | 79 | 78 | 74 | 80 | 75 | 76 | 61 | 78 | 71 | 75 | 74 | 48 | 74 | 50 | 74 | 74 | 49 | 70 | 84 | 78 | 37 | 76 | 75 | 70 | 62 | 74 | 77 | 46 | 68 | 79 | 71 | 67 | 52 | 71 | 70 |
| 87 | 79 | 79 | 72 | 84 | 84 | 77 | 76 | 78 | 80 | 75 | 76 | 54 | 62 | 67 | 66 | 48 | 74 | 50 | 74 | 74 | 69 | 70 | 84 | 78 | 37 | 76 | 75 | 86 | 88 | 86 | 85 | 75 | 78 | 87 | 77 | 67 | 60 | 73 | 80 | |
| 162 | 90 | 73 | 89 | 89 | 88 | 73 | 89 | 71 | 87 | 86 | 85 | 76 | 73 | 84 | 85 | 58 | 85 | 71 | 85 | 81 | 87 | 87 | 85 | 83 | 78 | 56 | 89 | 86 | 88 | 86 | 85 | 75 | 78 | 87 | 77 | 67 | 60 | 73 | 80 | |
| 51 | 84 | 63 | 76 | 76 | 53 | 76 | 61 | 84 | 82 | 84 | 75 | 86 | 83 | 84 | 88 | 43 | 78 | 57 | 84 | 84 | 73 | 84 | 71 | 65 | 47 | 88 | 88 | 80 | 88 | 80 | 59 | 67 | 90 | 63 | 67 | 53 | 57 | 73 | 79 | |
| 144 | 81 | 62 | 87 | 85 | 78 | 90 | 72 | 81 | 89 | 85 | 78 | 83 | 84 | 72 | 91 | 86 | 41 | 87 | 58 | 81 | 81 | 67 | 85 | 75 | 77 | 69 | 83 | 90 | 88 | 88 | 86 | 84 | 79 | 85 | 72 | 61 | 56 | 78 | 79 | |
| 61 | 90 | 70 | 97 | 90 | 89 | 72 | 89 | 70 | 90 | 82 | 97 | 90 | 89 | 97 | 92 | 93 | 41 | 85 | 72 | 90 | 92 | 74 | 97 | 87 | 90 | 59 | 92 | 92 | 90 | 90 | 93 | 82 | 77 | 93 | 75 | 80 | 43 | 75 | 83 | |
| 97 | 78 | 68 | 83 | 87 | 79 | 74 | 85 | 69 | 82 | 84 | 75 | 82 | 71 | 86 | 86 | 45 | 76 | 49 | 78 | 78 | 69 | 75 | 76 | 78 | 37 | 76 | 84 | 91 | 86 | 86 | 86 | 74 | 80 | 83 | 72 | 62 | 51 | 76 | 75 | |
| 190 | 80 | 71 | 86 | 85 | 86 | 76 | 81 | 74 | 77 | 85 | 76 | 64 | 72 | 87 | 82 | 45 | 84 | 55 | 75 | 74 | 67 | 75 | 78 | 76 | 79 | 70 | 84 | 91 | 86 | 86 | 86 | 74 | 80 | 83 | 72 | 62 | 51 | 76 | 75 | |
| 133 | 84 | 72 | 83 | 89 | 88 | 78 | 95 | 72 | 62 | 77 | 59 | 64 | 32 | 73 | 80 | 71 | 41 | 78 | 43 | 58 | 59 | 41 | 59 | 80 | 93 | 69 | 80 | 92 | 86 | 80 | 84 | 88 | 75 | 96 | 81 | 75 | 51 | 77 | 73 | |
| 146 | 82 | 77 | 86 | 82 | 85 | 81 | 88 | 77 | 78 | 85 | 87 | 71 | 67 | 80 | 84 | 84 | 52 | 86 | 50 | 79 | 79 | 72 | 87 | 86 | 78 | 66 | 88 | 84 | 92 | 87 | 79 | 81 | 74 | 81 | 78 | 67 | 57 | 70 | 78 | |
| 141 | 82 | 67 | 80 | 80 | 81 | 70 | 82 | 70 | 78 | 83 | 84 | 72 | 61 | 79 | 92 | 81 | 42 | 83 | 62 | 77 | 79 | 78 | 82 | 74 | 74 | 62 | 86 | 74 | 86 | 82 | 84 | 80 | 70 | 82 | 71 | 63 | 57 | 70 | 75 | |
| 203 | 67 | 54 | 67 | 70 | 67 | 61 | 77 | 61 | 78 | 80 | 70 | 68 | 64 | 68 | 76 | 70 | 43 | 75 | 67 | 66 | 75 | 67 | 70 | 66 | 66 | 62 | 73 | 76 | 74 | 76 | 70 | 70 | 72 | 57 | 54 | 50 | 60 | 68 | | |
| 214 | 73 | 60 | 70 | 69 | 73 | 70 | 64 | 66 | 58 | 63 | 64 | 51 | 63 | 66 | 65 | 62 | 43 | 61 | 62 | 66 | 69 | 49 | 65 | 70 | 62 | 49 | 65 | 70 | 72 | 69 | 76 | 66 | 61 | 69 | 60 | 51 | 50 | 71 | 64 | |
| 181 | 72 | 69 | 70 | 81 | 82 | 69 | 72 | 77 | 54 | 55 | 60 | 55 | 61 | 58 | 66 | 65 | 48 | 56 | 42 | 62 | 74 | 67 | 64 | 75 | 69 | 57 | 65 | 70 | 68 | 63 | 72 | 57 | 70 | 58 | 60 | 52 | 72 | 64 | | |
| 203 | 79 | 69 | 65 | 72 | 80 | 73 | 76 | 84 | 66 | 65 | 63 | 62 | 55 | 64 | 71 | 63 | 42 | 70 | 58 | 62 | 66 | 77 | 64 | 75 | 69 | 61 | 72 | 69 | 68 | 74 | 66 | 63 | 70 | 79 | 61 | 54 | 76 | 68 | | |
| 60 | 77 | 73 | 83 | 88 | 77 | 87 | 75 | 78 | 82 | 83 | 82 | 80 | 82 | 80 | 82 | 75 | 85 | 47 | 83 | 78 | 85 | 83 | 83 | 83 | 83 | 83 | 85 | 83 | 85 | 85 | 88 | 83 | 82 | 83 | 73 | 65 | 48 | 80 | 79 | |
| 76 | 79 | 78 | 91 | 93 | 84 | 82 | 84 | 76 | 91 | 91 | 93 | 75 | 84 | 91 | 95 | 86 | 50 | 93 | 55 | 91 | 89 | 92 | 93 | 86 | 84 | 51 | 91 | 95 | 92 | 95 | 91 | 79 | 76 | 91 | 75 | 59 | 45 | 82 | 82 | |
| 251 | 76 | 71 | 89 | 87 | 77 | 75 | 77 | 72 | 86 | 89 | 81 | 72 | 83 | 86 | 88 | 43 | 87 | 67 | 84 | 85 | 74 | 81 | 74 | 73 | 62 | 87 | 89 | 88 | 87 | 81 | 80 | 64 | 86 | 72 | 57 | 69 | 74 | 78 | | |
| 104 | 88 | 75 | 78 | 81 | 84 | 80 | 90 | 80 | 62 | 65 | 68 | 58 | 65 | 68 | 62 | 67 | 38 | 66 | 59 | 58 | 62 | 69 | 77 | 83 | 83 | 72 | 85 | 83 | 85 | 83 | 82 | 83 | 73 | 65 | 48 | 80 | 79 | | | |
| 64 | 81 | 88 | 84 | 83 | 80 | 78 | 88 | 88 | 73 | 72 | 78 | 58 | 72 | 73 | 77 | 47 | 75 | 64 | 83 | 80 | 89 | 67 | 83 | 78 | 66 | 78 | 83 | 78 | 68 | 75 | 78 | 78 | 77 | 66 | 88 | 78 | 79 | 80 | 76 | |
| 46 | 80 | 76 | 61 | 67 | 83 | 74 | 78 | 83 | 48 | 52 | 61 | 46 | 63 | 46 | 54 | 35 | 59 | 65 | 70 | 67 | 85 | 61 | 78 | 78 | 66 | 78 | 78 | 75 | 78 | 78 | 68 | 67 | 66 | 64 | 70 | 67 | 66 | 66 | 65 | |
| 247 | 74 | 74 | 70 | 69 | 68 | 63 | 81 | 74 | 64 | 64 | 64 | 58 | 61 | 62 | 61 | 40 | 66 | 68 | 65 | 66 | 86 | 64 | 76 | 66 | 48 | 57 | 64 | 67 | 66 | 65 | 64 | 67 | 66 | 65 | 66 | 66 | 66 | 66 | 65 | |
| 87 | 56 | 56 | 75 | 75 | 69 | 56 | 75 | 54 | 82 | 71 | 78 | 54 | 63 | 77 | 69 | 44 | 69 | 51 | 74 | 74 | 68 | 78 | 72 | 71 | 68 | 49 | 68 | 74 | 64 | 74 | 67 | 49 | 68 | 77 | 62 | 66 | 56 | 66 | 66 | |
| 109 | 75 | 83 | 74 | 74 | 69 | 78 | 64 | 80 | 73 | 72 | 49 | 59 | 68 | 71 | 76 | 50 | 74 | 69 | | | | | | | | | | | | | | | | | | | | | | |

| ling. category | ling. phenomenon | # | baselines | | | | | | | | | | | QE as a metric | | | | | | | | | | | ref. based metrics | | | | | | | | | | | | | | | | |
|----------------|----------------------------------|-------|-----------|------|-----------|-------|-------|------|----------|-------|--------------|---------------|-----------|----------------|-------------|-----------------|-----------------|----------------|--------------------|---------|-----------------|-----------------|-----------|---------|--------------------|--------------|--------|--------------|--------------|------------|-----------------|-----------|------------|-------|-----------------|-------|-------------|----------------|-------------|-----|----|
| | | | BERTscore | BLEU | BLEURT-20 | COMET | YSI-1 | chrF | prismRef | sBLEU | CometKiwi-XL | CometKiwi-XXL | CometKiwi | GEMBA-MQM | MS-COMET-QE | MetricX-23-QE-b | MetricX-23-QE-c | Random-synname | XCOMET-QE-Ensemble | XLsimQE | cometoid22-wm11 | cometoid22-wm12 | prismSrc2 | KG-BERT | MEE4 | MEE4_snb_xlm | MATESe | MetricX-23-b | MetricX-23-c | MetricX-23 | XCOMET-Ensemble | XCOMET-XL | XCOMET-XXL | XLsim | cometoid22-wm13 | eBLEU | embed_llama | partokengram_F | tokengram_F | avg | |
| | Reflexive - future II subj. II | 107 | 90 | 67 | 84 | 93 | 92 | 71 | 93 | 70 | 89 | 93 | 91 | 93 | 76 | 84 | 88 | 79 | 41 | 92 | 63 | 84 | 92 | 83 | 91 | 75 | 82 | 55 | 86 | 84 | 88 | 94 | 93 | 82 | 69 | 93 | 69 | 57 | 73 | 73 | 81 |
| | Reflexive - perfect | 188 | 85 | 61 | 83 | 85 | 80 | 60 | 85 | 64 | 82 | 88 | 90 | 80 | 78 | 83 | 90 | 80 | 46 | 88 | 58 | 79 | 84 | 84 | 90 | 76 | 74 | 63 | 84 | 91 | 88 | 86 | 83 | 82 | 69 | 91 | 70 | 57 | 60 | 60 | 77 |
| | Reflexive - pluperfect | 109 | 96 | 74 | 76 | 91 | 94 | 67 | 86 | 71 | 70 | 77 | 80 | 84 | 75 | 66 | 84 | 62 | 43 | 73 | 55 | 72 | 71 | 81 | 80 | 80 | 81 | 55 | 69 | 83 | 69 | 79 | 74 | 70 | 77 | 96 | 72 | 75 | 72 | 72 | 75 |
| | Reflexive - pluperfect subj. II | 90 | 90 | 68 | 87 | 90 | 87 | 74 | 92 | 74 | 78 | 87 | 87 | 92 | 88 | 89 | 83 | 78 | 38 | 86 | 64 | 89 | 84 | 88 | 87 | 84 | 86 | 58 | 79 | 82 | 79 | 84 | 81 | 84 | 76 | 92 | 83 | 69 | 76 | 76 | 81 |
| | Reflexive - present | 125 | 82 | 55 | 90 | 86 | 82 | 72 | 88 | 66 | 79 | 84 | 88 | 81 | 91 | 89 | 92 | 87 | 46 | 85 | 58 | 89 | 86 | 91 | 88 | 81 | 78 | 62 | 87 | 90 | 84 | 92 | 87 | 86 | 69 | 88 | 70 | 58 | 70 | 70 | 80 |
| | Reflexive - preterite | 117 | 92 | 75 | 86 | 85 | 90 | 77 | 92 | 77 | 71 | 68 | 74 | 89 | 83 | 80 | 80 | 73 | 39 | 71 | 61 | 81 | 78 | 74 | 92 | 88 | 81 | 78 | 62 | 90 | 93 | 86 | 77 | 82 | 79 | 71 | 85 | 85 | 79 | 80 | 79 |
| | Reflexive - preterite subj. II | 124 | 94 | 89 | 90 | 93 | 97 | 81 | 96 | 85 | 86 | 81 | 90 | 96 | 88 | 93 | 86 | 93 | 44 | 85 | 55 | 82 | 88 | 90 | 95 | 90 | 77 | 90 | 77 | 98 | 98 | 88 | 96 | 92 | 77 | 97 | 85 | 73 | 83 | 83 | 87 |
| | Transitive - future I | 43 | 98 | 84 | 95 | 95 | 100 | 86 | 98 | 86 | 84 | 93 | 77 | 88 | 91 | 84 | 81 | 79 | 44 | 93 | 51 | 84 | 93 | 77 | 95 | 91 | 60 | 93 | 98 | 93 | 93 | 86 | 86 | 86 | 95 | 65 | 56 | 86 | 86 | 85 | |
| | Transitive - future I subj. II | 37 | 97 | 81 | 86 | 95 | 100 | 89 | 95 | 81 | 95 | 95 | 84 | 81 | 100 | 95 | 97 | 89 | 43 | 92 | 68 | 70 | 95 | 100 | 84 | 89 | 89 | 62 | 86 | 89 | 84 | 89 | 89 | 73 | 84 | 97 | 81 | 54 | 84 | 84 | 85 |
| | Transitive - future II | 33 | 97 | 82 | 94 | 94 | 100 | 91 | 94 | 79 | 91 | 94 | 97 | 91 | 85 | 94 | 97 | 82 | 45 | 94 | 67 | 76 | 85 | 100 | 97 | 85 | 97 | 61 | 97 | 97 | 94 | 88 | 79 | 94 | 97 | 73 | 36 | 94 | 94 | 87 | |
| | Transitive - future II subj. II | 50 | 96 | 62 | 78 | 80 | 86 | 82 | 68 | 74 | 92 | 98 | 70 | 92 | 66 | 96 | 98 | 98 | 56 | 96 | 54 | 90 | 88 | 86 | 70 | 72 | 70 | 50 | 82 | 76 | 82 | 80 | 80 | 80 | 82 | 96 | 68 | 56 | 82 | 82 | 79 |
| | Transitive - perfect | 99 | 88 | 73 | 76 | 91 | 95 | 85 | 96 | 77 | 70 | 77 | 72 | 84 | 82 | 82 | 83 | 72 | 43 | 73 | 60 | 62 | 69 | 86 | 72 | 88 | 94 | 58 | 83 | 84 | 84 | 81 | 78 | 73 | 85 | 96 | 80 | 72 | 86 | 86 | 79 |
| | Transitive - pluperfect | 22 | 82 | 41 | 86 | 86 | 86 | 59 | 95 | 50 | 77 | 91 | 82 | 95 | 95 | 91 | 91 | 77 | 27 | 91 | 45 | 68 | 68 | 91 | 86 | 82 | 64 | 77 | 82 | 82 | 86 | 91 | 91 | 77 | 68 | 91 | 64 | 32 | 55 | 55 | 75 |
| | Transitive - pluperfect subj. II | 39 | 90 | 64 | 92 | 90 | 90 | 72 | 92 | 69 | 79 | 87 | 92 | 95 | 95 | 92 | 90 | 92 | 49 | 82 | 51 | 82 | 85 | 85 | 92 | 97 | 87 | 87 | 97 | 100 | 97 | 87 | 90 | 90 | 85 | 95 | 85 | 59 | 72 | 84 | |
| | Transitive - present | 33 | 73 | 52 | 85 | 91 | 82 | 67 | 91 | 70 | 82 | 88 | 91 | 100 | 91 | 85 | 97 | 82 | 48 | 85 | 70 | 88 | 91 | 70 | 91 | 94 | 88 | 88 | 97 | 94 | 97 | 88 | 85 | 94 | 79 | 97 | 85 | 64 | 76 | 76 | 83 |
| | Transitive - preterite | 57 | 82 | 39 | 84 | 79 | 77 | 56 | 77 | 54 | 84 | 81 | 81 | 63 | 72 | 77 | 77 | 40 | 79 | 58 | 81 | 81 | 74 | 81 | 77 | 70 | 58 | 77 | 79 | 75 | 82 | 75 | 84 | 68 | 84 | 74 | 53 | 58 | 58 | 72 | |
| | Transitive - preterite subj. II | 97 | 81 | 48 | 88 | 88 | 89 | 59 | 88 | 54 | 82 | 87 | 92 | 94 | 84 | 85 | 94 | 77 | 41 | 88 | 65 | 82 | 85 | 79 | 92 | 78 | 77 | 92 | 91 | 95 | 93 | 95 | 92 | 74 | 88 | 64 | 62 | 62 | 62 | 80 | |
| | Case government | 80 | 89 | 74 | 86 | 91 | 85 | 72 | 90 | 70 | 84 | 74 | 84 | 60 | 70 | 82 | 78 | 78 | 57 | 76 | 50 | 79 | 80 | 71 | 84 | 84 | 70 | 57 | 86 | 86 | 86 | 86 | 89 | 74 | 80 | 92 | 70 | 51 | 75 | 75 | 77 |
| | Mediopassive voice | 50 | 94 | 72 | 92 | 90 | 96 | 78 | 94 | 72 | 90 | 84 | 88 | 84 | 86 | 90 | 96 | 88 | 54 | 94 | 40 | 88 | 94 | 72 | 88 | 90 | 88 | 68 | 90 | 90 | 94 | 90 | 90 | 84 | 92 | 94 | 86 | 58 | 80 | 80 | 84 |
| | Passive voice | 33 | 55 | 55 | 55 | 55 | 55 | 55 | 55 | 55 | 39 | 39 | 42 | 18 | 27 | 45 | 39 | 30 | 27 | 39 | 27 | 42 | 39 | 27 | 42 | 55 | 55 | 36 | 52 | 55 | 39 | 48 | 36 | 33 | 55 | 55 | 52 | 55 | 55 | 45 | |
| macro avg. | | 10402 | 84 | 69 | 82 | 84 | 85 | 75 | 86 | 71 | 80 | 81 | 81 | 76 | 75 | 81 | 83 | 79 | 43 | 80 | 57 | 74 | 75 | 76 | 81 | 81 | 80 | 56 | 81 | 84 | 81 | 83 | 80 | 73 | 76 | 87 | 74 | 62 | 69 | 75 | 76 |
| micro avg. | | 10402 | 83 | 70 | 82 | 84 | 84 | 75 | 85 | 72 | 79 | 80 | 80 | 75 | 74 | 80 | 83 | 78 | 44 | 80 | 58 | 75 | 76 | 75 | 80 | 81 | 79 | 57 | 81 | 83 | 81 | 82 | 81 | 74 | 76 | 87 | 74 | 63 | 67 | 75 | 76 |

Table 4: Accuracy of the metrics(%) with regard to the linguistically-motivated phenomena for German-English

| ling. category | ling. phenomenon | # | baselines | | | | | | | | | | | | | | | | | | | | QE as a metric | | | | | | | | | | ref. based metrics | | | | | | | | | | avg |
|----------------------------|-----------------------|-----|-----------|------------|------------|------------|------------|------------|------------|------------|--------------|---------------|------------|------------|------------|-------------|-----------------|-----------------|---------------|----------------|--------------------|------------|-----------------|-----------------|-----------------|------------|------------|--------------|--------------|------------|-----------------|------------|--------------------|------------|------------|-------------|----------------|-------------|------------|----|--|--|-----|
| | | | BERTscore | BLEU | BLEURT-20 | COMET | Ysis-1 | chrf | prismKef | sBLEU | Cometkiwi-XL | Cometkiwi-XXL | Cometkiwi | GEMBA-MQM | KG-BERT | MS-COMET-QE | MetricX-23-QE-b | MetricX-23-QE-c | MetricX-23-QE | Random-synname | XCOMET-QE-Ensemble | XLSimQE | cometoid22-wm12 | cometoid22-wm11 | cometoid22-wm13 | prismKef2 | MATESE | MetricX-23-b | MetricX-23-c | MetricX-23 | XCOMET-Ensemble | XCOMET-XL | XCOMET-XXL | XLSim | eBLEU | embed_llama | partokengram_F | tokengram_F | | | | | |
| Ambiguity | Lexical ambiguity | 66 | 80.67 | 100 | 100 | 97 | 88 | 67 | 83 | 73 | 100 | 94 | 100 | 100 | 100 | 95 | 100 | 100 | 100 | 39 | 100 | 30 | 100 | 100 | 100 | 56 | 98 | 100 | 100 | 100 | 100 | 100 | 100 | 88 | 77 | 44 | 59 | 70 | 86 | | | | |
| | Gapping | 44 | 80.75 | 68 | 77 | 73 | 66 | 77 | 73 | 93 | 86 | 66 | 30 | 66 | 93 | 43 | 63 | 84 | 52 | 84 | 34 | 73 | 68 | 80 | 53 | 14 | 14 | 45 | 75 | 39 | 73 | 70 | 57 | 73 | 86 | 50 | 55 | 75 | 66 | | | | |
| Coordination & ellipsis | Pseudogapping | 45 | 87.78 | 89 | 89 | 91 | 87 | 91 | 80 | 76 | 60 | 40 | 76 | 69 | 69 | 69 | 69 | 56 | 47 | 71 | 53 | 78 | 67 | 78 | 63 | 13 | 13 | 64 | 73 | 49 | 73 | 60 | 44 | 89 | 84 | 78 | 62 | 80 | 70 | | | | |
| | Stuicing | 11 | 64.55 | 82 | 45 | 55 | 64 | 55 | 55 | 18 | 55 | 36 | 64 | 36 | 64 | 64 | 82 | 82 | 45 | 64 | 73 | 55 | 55 | 27 | 27 | 0 | 55 | 27 | 36 | 64 | 55 | 0 | 91 | 91 | 64 | 45 | 64 | 54 | | | | | |
| False friends | Stripping | 27 | 78.63 | 59 | 48 | 63 | 56 | 81 | 52 | 41 | 26 | 26 | 15 | 26 | 52 | 67 | 22 | 22 | 22 | 22 | 32 | 48 | 52 | 56 | 48 | 74 | 4 | 30 | 37 | 41 | 67 | 41 | 67 | 41 | 67 | 41 | 67 | 41 | 67 | 41 | | | |
| | VP-ellipsis | 76 | 91.76 | 57 | 74 | 76 | 59 | 70 | 68 | 72 | 57 | 86 | 47 | 86 | 82 | 63 | 93 | 95 | 43 | 62 | 32 | 74 | 74 | 71 | 46 | 39 | 79 | 82 | 79 | 68 | 62 | 61 | 66 | 76 | 45 | 70 | 61 | 68 | | | | | |
| Function word | False friends | 6 | 100.17 | 100 | 100 | 100 | 100 | 17 | 83 | 50 | 0 | 0 | 0 | 0 | 0 | 0 | 100 | 17 | 17 | 0 | 17 | 33 | 33 | 100 | 0 | 17 | 100 | 100 | 100 | 33 | 33 | 100 | 83 | 100 | 50 | 83 | 33 | 50 | | | | | |
| | Focus particle | 50 | 56.44 | 36 | 30 | 52 | 36 | 56 | 46 | 42 | 24 | 42 | 20 | 42 | 18 | 48 | 44 | 62 | 44 | 56 | 34 | 48 | 26 | 10 | 46 | 16 | 66 | 70 | 64 | 48 | 36 | 20 | 42 | 42 | 52 | 56 | 38 | 42 | | | | | |
| MWE | Question tag | 66 | 62.88 | 94 | 82 | 53 | 39 | 42 | 91 | 62 | 79 | 56 | 74 | 56 | 50 | 80 | 64 | 38 | 44 | 85 | 42 | 92 | 88 | 14 | 3 | 61 | 47 | 58 | 70 | 79 | 24 | 55 | 76 | 52 | 52 | 56 | 38 | 61 | | | | | |
| | Collocation | 79 | 71.68 | 81 | 82 | 76 | 67 | 73 | 72 | 86 | 87 | 90 | 65 | 90 | 90 | 82 | 86 | 85 | 39 | 85 | 34 | 85 | 85 | 34 | 46 | 82 | 86 | 84 | 89 | 84 | 81 | 70 | 72 | 70 | 65 | 66 | 75 | | | | | | |
| Named entity & terminology | Idiom | 31 | 61.45 | 65 | 68 | 58 | 42 | 84 | 42 | 26 | 13 | 39 | 39 | 68 | 58 | 55 | 90 | 45 | 48 | 10 | 16 | 52 | 32 | 52 | 0 | 90 | 100 | 100 | 13 | 35 | 32 | 81 | 45 | 39 | 45 | 45 | 50 | | | | | | |
| | Verbal MWE | 12 | 83.83 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 92 | 50 | 92 | 17 | 83 | 92 | 83 | 25 | 100 | 92 | 75 | 67 | 92 | 8 | 42 | 83 | 100 | 100 | 83 | 100 | 83 | 100 | 83 | 42 | 58 | 100 | 81 | | | | | |
| Negation | Date | 45 | 60.47 | 87 | 69 | 58 | 49 | 80 | 49 | 80 | 78 | 53 | 40 | 53 | 64 | 69 | 56 | 87 | 42 | 76 | 38 | 82 | 82 | 58 | 9 | 84 | 67 | 84 | 82 | 100 | 53 | 73 | 64 | 67 | 53 | 49 | 65 | | | | | | |
| | Domain-specific term | 150 | 82.71 | 97 | 99 | 85 | 79 | 93 | 81 | 93 | 57 | 53 | 65 | 53 | 92 | 61 | 87 | 77 | 51 | 61 | 44 | 81 | 83 | 98 | 8 | 68 | 97 | 100 | 97 | 95 | 99 | 94 | 85 | 81 | 65 | 51 | 79 | 77 | | | | | |
| Non-verbal agreement | Measuring unit | 45 | 87.84 | 100 | 100 | 100 | 100 | 82 | 89 | 62 | 84 | 52 | 83 | 82 | 2 | 69 | 42 | 42 | 47 | 60 | 100 | 100 | 100 | 80 | 33 | 64 | 60 | 47 | 82 | 53 | 11 | 91 | 62 | 62 | 87 | 68 | 87 | | | | | | |
| | Proper name | 3 | 100.67 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 0 | 100 | 0 | 0 | 100 | 67 | 100 | 33 | 100 | 100 | 33 | 67 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 85 | | | | |
| Punctuation | Negation | 34 | 74.74 | 79 | 68 | 74 | 85 | 68 | 74 | 68 | 85 | 85 | 91 | 85 | 41 | 71 | 100 | 9 | 59 | 91 | 62 | 29 | 59 | 91 | 0 | 94 | 71 | 94 | 79 | 71 | 65 | 76 | 76 | 67 | 85 | 85 | 69 | 71 | | | | | |
| | Conference | 57 | 68.54 | 86 | 77 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 67 | 80 | 80 | 64 | 60 | 47 | 82 | 53 | 11 | 91 | 62 | 62 | 87 | 68 | 87 | | | | | |
| Subordination | Genitive | 4 | 75.75 | 25 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 25 | 25 | 25 | 25 | 75 | 50 | 25 | 100 | 25 | 25 | 100 | 50 | 0 | 75 | 75 | 75 | 100 | 25 | 0 | 100 | 75 | 100 | 25 | 0 | 100 | 75 | 100 | | | | |
| | Direct Speech | 121 | 42.46 | 53 | 68 | 86 | 41 | 73 | 55 | 66 | 29 | 65 | 27 | 65 | 87 | 55 | 66 | 43 | 41 | 64 | 66 | 67 | 36 | 56 | 1 | 60 | 71 | 52 | 66 | 57 | 12 | 63 | 66 | 84 | 45 | 45 | 54 | 46 | | | | | |
| Verb tense/aspect/mood | Adverbial clause | 132 | 43.52 | 32 | 47 | 55 | 46 | 48 | 47 | 49 | 74 | 41 | 32 | 41 | 32 | 41 | 32 | 41 | 32 | 41 | 16 | 48 | 16 | 39 | 22 | 5 | 60 | 92 | 44 | 64 | 43 | 80 | 39 | 30 | 39 | 45 | 45 | 46 | | | | | |
| | Cleft sentence | 24 | 71.54 | 92 | 92 | 92 | 92 | 75 | 79 | 54 | 96 | 79 | 75 | 88 | 75 | 79 | 92 | 96 | 100 | 54 | 100 | 46 | 88 | 88 | 67 | 79 | 88 | 92 | 92 | 100 | 92 | 75 | 54 | 75 | 62 | 62 | 80 | 80 | | | | | |
| Verb tense/aspect/mood | Contact clause | 10 | 60.30 | 90 | 100 | 100 | 100 | 100 | 100 | 60 | 100 | 100 | 40 | 100 | 100 | 40 | 100 | 70 | 100 | 70 | 100 | 100 | 100 | 100 | 60 | 0 | 100 | 100 | 100 | 60 | 0 | 80 | 40 | 60 | 60 | 74 | 74 | | | | | | |
| | Infinitive clause | 24 | 71.75 | 67 | 96 | 79 | 96 | 67 | 67 | 96 | 54 | 83 | 62 | 83 | 83 | 71 | 88 | 75 | 38 | 83 | 38 | 75 | 75 | 79 | 79 | 33 | 92 | 100 | 100 | 100 | 96 | 96 | 50 | 71 | 58 | 96 | 96 | 76 | | | | | |
| Verb tense/aspect/mood | Object clause | 70 | 66.80 | 73 | 83 | 67 | 84 | 79 | 83 | 100 | 93 | 100 | 23 | 100 | 84 | 21 | 19 | 21 | 34 | 100 | 37 | 100 | 100 | 50 | 0 | 26 | 80 | 26 | 100 | 84 | 17 | 71 | 57 | 89 | 89 | 67 | 67 | | | | | | |
| | Participle clause | 33 | 70.82 | 67 | 79 | 94 | 76 | 42 | 82 | 42 | 70 | 42 | 9 | 42 | 9 | 42 | 9 | 42 | 9 | 42 | 9 | 42 | 88 | 43 | 35 | 75 | 78 | 76 | 76 | 78 | 63 | 92 | 71 | 65 | 63 | 72 | 63 | 72 | | | | | |
| Verb tense/aspect/mood | Pseudo-cleft sentence | 51 | 84.78 | 86 | 86 | 78 | 65 | 73 | 73 | 65 | 63 | 65 | 69 | 65 | 61 | 75 | 100 | 88 | 39 | 75 | 55 | 94 | 84 | 88 | 43 | 35 | 75 | 78 | 76 | 76 | 78 | 63 | 92 | 71 | 65 | 63 | 72 | 63 | | | | | |
| | Relative clause | 117 | 66.56 | 66 | 71 | 63 | 74 | 66 | 67 | 81 | 41 | 64 | 35 | 64 | 91 | 85 | 78 | 86 | 47 | 83 | 36 | 62 | 74 | 66 | 25 | 32 | 54 | 85 | 68 | 84 | 85 | 32 | 67 | 58 | 60 | 71 | 64 | 64 | | | | | |
| Verb tense/aspect/mood | Subject clause | 38 | 55.53 | 53 | 55 | 61 | 45 | 61 | 45 | 61 | 47 | 84 | 58 | 97 | 82 | 97 | 84 | 61 | 76 | 84 | 47 | 84 | 24 | 84 | 82 | 53 | 63 | 5 | 50 | 68 | 37 | 84 | 87 | 47 | 53 | 45 | 79 | 50 | 62 | | | | |
| | Conditional | 6 | 100.67 | 100 | 100 | 100 | 100 | 83 | 100 | 83 | 100 | 0 | 17 | 83 | 17 | 83 | 0 | 17 | 33 | 17 | 17 | 67 | 0 | 0 | 83 | 50 | 33 | 100 | 100 | 100 | 67 | 67 | 50 | 100 | 50 | 100 | 83 | 62 | | | | | |
| Verb tense/aspect/mood | Direct object | 13 | 85.69 | 77 | 100 | 69 | 92 | 100 | 69 | 92 | 69 | 92 | 62 | 92 | 69 | 62 | 92 | 54 | 69 | 46 | 92 | 77 | 92 | 31 | 31 | 85 | 100 | 100 | 100 | 92 | 92 | 85 | 62 | 69 | 62 | 85 | 85 | 79 | | | | | |
| | Gerund | 34 | 82.74 | 100 | 100 | 100 | 88 | 91 | 85 | 91 | 88 | 91 | 88 | 91 | 91 | 94 | 38 | 100 | 26 | 100 | 100 | 26 | 100 | 100 | 18 | 74 | 91 | 91 | 97 | 91 | 91 | 82 | 79 | 71 | 68 | 82 | 82 | 83 | | | | | |
| Verb tense/aspect/mood | Imperative | 12 | 75.67 | 75 | 75 | 58 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 75 | 50 | 83 | 33 | 17 | 100 | 100 | 100 | 100 | 100 | 92 | 75 | 92 | 75 | 92 | 75 | 75 | 80 | 80 | | | | |
| | Intransitive | 3 | 67.33 | 67 | 67 | 67 | 100 | 67 | 67 | 100 | 67 | 67 | 100 | 67 | 100 | 67 | 100 | 67 | 100 | 67 | 100 | 67 | 67 | 100 | 0 | 0 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 100 | 76 | | | |
| Verb valency | Reflexive | 24 | 71.54 | 62 | 62 | 58 | 88 | 71 | 75 | 100 | 92 | 71 | 33 | 71 | 83 | 71 | 79 | 71 | 50 | 67 | 88 | 88 | 58 | 83 | 50 | 17 | 100 | 100 | 92 | 100 | 96 | 100 | 75 | 83 | 67 | 83 | 74 | 74 | | | | | |
| | Transitive | 43 | 77.56 | 63 | 84 | 79 | 81 | 72 | 65 | 65 | 53 | 49 | 51 | 49 | 37 | 35 | 33 | 28 | 35 | 58 | 67 | 37 | 33 | 72 | 51 | 53 | 77 | 88 | 77 | 90 | 60 | 74 | 70 | 67 | 84 | 84 | 62 | | | | | | |
| Verb valency | Case government | 3 | 100.33 | 0 | 33 | 100 | 33 | 33 | 33 | 67 | 67 | 0 | 0 | 0 | 0 | 0 | 33 | 0 | 33 | 0 | 33 | 33 | 0 | 33 | 33 | 0 | 33 | 0 | 67 | 33 | 67 | 0 | 33 | 67 | 33 | 33 | 33 | 35 | | | | | |
| | Catenative verb | 50 | 78.80 | 74 | 90 | 86 | 70 | 66 | 80 | 94 | 96 | 56 | 96 | 90 | 74 | 90 | 74 | 36 | 78 | 42 | 74 | 66 | 76 | 42 | 58 | 76 | 82 | 74 | 70 | 66 | 90 | 92 | 94 | 74 | 74 | 75 | 75 | | | | | | |
| Verb valency | Impersonal Subject | 10 | 40.50 | 80 | 40 | 70 | 50 | 20 | 40 | 40 | 40 | 90 | 40 | 90 | 40 | 40 | 100 | 100 | 40 | 40 | 0 | 40 | 40 | 10 | 60 | 100 | 100 | 100 | 40 | 40 | 40 | 40 | 40 | 4 | | | | | | | | | |