Macro-Average: Rare Types Are Important Too

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Abstract

While traditional corpus-level evaluation metrics for machine translation (MT) correlate well with fluency, they struggle to reflect adequacy. Model-based MT metrics trained on segment-level human judgments have emerged as an attractive replacement due to strong correlation results. These models, however, require potentially expensive re-training for new domains and languages. Furthermore, their decisions are inherently non-transparent and appear to reflect unwelcome biases. We explore the simple type-based classifier metric, MACROF₁, and study its applicability to MT evaluation. We find that $MACROF_1$ is competitive on direct assessment, and outperforms others in indicating downstream cross-lingual information retrieval task performance. Further, we show that $MACROF_1$ can be used to effectively compare supervised and unsupervised neural machine translation, and reveal significant qualitative differences in the methods' outputs.¹

1 Introduction

Model-based metrics for evaluating machine translation such as BLEURT (Sellam et al., 2020), ESIM (Mathur et al., 2019), and YiSi (Lo, 2019) have recently attracted attention due to their superior correlation with human judgments (Ma et al., 2019). However, BLEU (Papineni et al., 2002) remains the most widely used corpus-level MT metric. It correlates reasonably well with human judgments, and moreover is easy to understand and cheap to calculate, requiring only reference translations in the target language. By contrast, model-based metrics require tuning on thousands of examples of human evaluation for every new target language or domain

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(Sellam et al., 2020). Model-based metric scores are also opaque and can hide undesirable biases, as can be seen in Table 1.

Reference:	You must be a doctor.				
Hypothesis:	m	ust be a doctor.			
	He	-0.735			
	Joe	-0.975			
	Sue	-1.043			
	She	-1.100			
Reference:	It is the gr	eatest country in the world.			
Hypothesis:	is	the greatest country in the world.			
	France	-0.022			
	America	-0.060			
	Russia	-0.161			
	Canada	-0.309			

 Table 1: A demonstration of BLEURT's internal biases; model-free metrics like BLEU would consider each of the errors above to be equally wrong.

The source of model-based metrics' (e.g. BLEURT) correlative superiority over model-free metrics (e.g. BLEU) appears to be the former's ability to focus evaluation on *adequacy*, while the latter are overly focused on *fluency*. BLEU and most other generation metrics consider each output *token* equally. Since natural language is dominated by a few high-count types, an MT model that concentrates on getting its *if* s, *and*s and *buts* right will benefit from BLEU in the long run more than one that gets its *xylophones*, *peripatetics*, and *defenestrates* right. Can we derive a metric with the discriminating power of BLEURT that does not share its bias or expense and is as interpretable as BLEU?

As it turns out, the metric may already exist and be in common use. Information extraction and other areas concerned with classification have long used both *micro averaging*, which treats each token equally, and *macro averaging*, which instead treats each *type* equally, when evaluating. The latter in particular is useful when seeking to avoid results dominated by overly frequent types. In this

¹Tools and analysis are available at https://github. com/thammegowda/007-mt-eval-macro. MT evaluation metrics are at https://github.com/isi-nlp/sacrebleu/tree/ macroavg-naacl21.

work we take a classification-based approach to evaluating machine translation in order to obtain an easy-to-calculate metric that focuses on adequacy as much as BLEURT but does not have the expensive overhead, opacity, or bias of model-based methods.

Our contributions are as follows: We consider MT as a classification task, and thus admit MACROF₁ as a legitimate approach to evaluation (Section 2). We show that $MACROF_1$ is competitive with other popular methods at tracking human judgments in translation (Section 3.2). We offer an additional justification of MACROF₁ as a performance indicator on adequacy-focused downstream tasks such as cross-lingual information retrieval (Section 3.3). Finally, we demonstrate that MACROF₁ is just as good as the expensive BLEURT at discriminating between structurally different MT approaches in a way BLEU cannot, especially regarding the adequacy of generated text, and provide a novel approach to qualitative analysis of the effect of metrics choice on quantitative evaluation (Section 4).

2 NMT as Classification

Neural machine translation (NMT) models are often viewed as pairs of encoder-decoder networks. Viewing NMT as such is useful in practice for implementation; however, such a view is inadequate for theoretical analysis. Gowda and May (2020) provide a high-level view of NMT as two fundamental ML components: an autoregressor and a classifier. Specifically, NMT is viewed as a multi-class classifier that operates on representations from an autoregressor. We may thus consider classifier-based evaluation metrics.

Consider a test corpus, $T = \{(x^{(i)}, h^{(i)}, y^{(i)}) | i = 1, 2, 3...m\}$ where $x^{(i)}, h^{(i)}$, and $y^{(i)}$ are source, system hypothesis, and reference translation, respectively. Let $x = \{x^{(i)} \forall i\}$ and similar for h and y. Let $V_h, V_y, V_{h \cap y}$, and V be the vocabulary of h, the vocabulary of $y, V_h \cap V_y$, and $V_h \cup V_y$, respectively. For each class $c \in V$,

$$PREDS(c) = \sum_{i=1}^{m} C(c, h^{(i)})$$
$$REFS(c) = \sum_{i=1}^{m} C(c, y^{(i)})$$
$$MATCH(c) = \sum_{i=1}^{m} min\{C(c, h^{(i)}), C(c, y^{(i)})\}$$

where C(c,a) counts the number of tokens of type c in sequence a (Papineni et al., 2002). For each class $c \in V_{h\cap y}$, precision (P_c) , recall (R_c) , and F_{β} measure $(F_{\beta;c})$ are computed as follows:²

$$P_{c} = \frac{\text{MATCH}(c)}{\text{PREDS}(c)}; \qquad R_{c} = \frac{\text{MATCH}(c)}{\text{REFS}(c)}$$
$$F_{\beta;c} = (1 + \beta^{2}) \frac{P_{c} \times R_{c}}{\beta^{2} \times P_{c} + R_{c}}$$

The *macro-average* consolidates individual performance by averaging by type, while the *microaverage* averages by token:

$$MACROF_{\beta} = \frac{\sum_{c \in V} F_{\beta;c}}{|V|}$$
$$MICROF_{\beta} = \frac{\sum_{c \in V} f(c) \times F_{\beta;c}}{\sum_{c' \in V} f(c')}$$

where f(c) = REFS(c) + k for smoothing factor k.³ We scale MACROF_{β} and MICROF_{β} values to percentile, similar to BLEU, for the sake of easier readability.

3 Justification for MACROF₁

In the following sections, we verify and justify the utility of MACROF₁ while also offering a comparison with popular alternatives such as MICROF₁, BLEU, CHRF₁, and BLEURT.⁴ We use Kendall's rank correlation coefficient, τ , to compute the association between metrics and human judgments. Correlations with p-values smaller than $\alpha = 0.05$ are considered to be statistically significant.

3.1 Data-to-Text: WebNLG

We use the 2017 WebNLG Challenge dataset (Gardent et al., 2017; Shimorina, 2018)⁵ to analyze the differences between micro- and macro- averaging. WebNLG is a task of generating English text for sets of triples extracted from DBPedia. Human annotations are available for a sample of 223 records each from nine NLG systems. The human

²We consider $F_{\beta;c}$ for $c \notin V_{h\cap y}$ to be 0.

³We use k = 1. When $k \to \infty$, MICROF_{β} \to MACROF_{β}.

⁴BLEU and CHRF₁ scores reported in this work are computed with SACREBLEU; see the Appendix for details. BLEURT scores are from the *base* model (Sellam et al., 2020). We consider two varieties of averaging to obtain a corpus-level metric from the segment-level BLEURT: mean and median of segment-level scores per corpus.

⁵https://gitlab.com/webnlg/ webnlg-human-evaluation

Name	Fluency & Grammar	Semantics
BLEU	×.444	×.500
CHRF ₁	×.278	.778
MACROF ₁	×.222	.722
MICROF ₁	×.333	.611
BLEURTmean	×.444	.833
BLEURTmedian	.611	.667

Table 2: WebNLG data-to-text task: Kendall's τ between system-level MT metric scores and human judgments. Fluency and grammar are correlated identically by all metrics. Values that are *not* significant at $\alpha = 0.05$ are indicated by \times .

judgments provided have three linguistic aspects fluency, grammar, and semantics⁶—which enable us to perform a fine grained analysis of our metrics. We compute Kendall's τ between metrics and human judgments, which are reported in Table 2.

As seen in Table 2, the metrics exhibit much variance in agreements with human judgments. For instance, BLEURTmedian is the best indicator of fluency and grammar, however BLEURTmean is best on semantics. BLEURT, being a model-based measure that is directly trained on human judgments, scores relatively higher than others. Considering the model-free metrics, CHRF1 does well on semantics but poorly on fluency and grammar compared to BLEU. Not surprisingly, both MICROF₁ and MACROF₁, which rely solely on unigrams, are poor indicators of fluency and grammar compared to BLEU, however MACROF1 is clearly a better indicator of semantics than BLEU. The discrepancy between MICROF₁ and MACROF₁ regarding their agreement with fluency, grammar, and semantics is expected: micro-averaging pays more attention to function words (as they are frequent types) that contribute to fluency and grammar whereas macroaveraging pays relatively more attention to the content words that contribute to semantic adequacy.

The take away from this analysis is as follows: $MACROF_1$ is a strong indicator of semantic adequacy, however, it is a poor indicator of fluency. We recommend using either $MACROF_1$ or $CHRF_1$ when semantic adequacy and not fluency is a desired goal.

3.2 Machine Translation: WMT Metrics

In this section, we verify how well the metrics agree with human judgments using Workshop on Machine Translation (WMT) metrics task datasets for 2017–2019 (Bojar et al., 2017; Ma et al., 2018,

2019).⁷ We first compute scores from each MT metric, and then calculate the correlation τ with human judgments.

As there are many language pairs and translation directions in each year, we report only the mean and median of τ , and number of wins per metric for each year in Table 3. We have excluded BLEURT from comparison in this section since the BLEURT models are fine-tuned on the same datasets on which we are evaluating the other methods.⁸ CHRF₁ has the strongest mean and median agreement with human judgments across the years. In 2018 and 2019, both MACROF₁ and MICROF₁ mean and median agreements outperform BLEU whereas in 2017 BLEU was better than MACROF₁ and MICROF₁.

As seen in Section 3.1, MACROF₁ weighs towards semantics whereas MICROF₁ and BLEU weigh towards fluency and grammar. This indicates that recent MT systems are mostly fluent, and adequacy is the key discriminating factor amongst them. BLEU served well in the early era of statistical MT when fluency was a harder objective. Recent advancements in neural MT models such as Transformers (Vaswani et al., 2017) produce fluent outputs, and have brought us to an era where semantic adequacy is the focus.

3.3 Cross-Lingual Information Retrieval

In this section, we determine correlation between MT metrics and downstream cross-lingual information retrieval (CLIR) tasks. CLIR is a kind of information retrieval (IR) task in which documents in one language are retrieved given queries in another (Grefenstette, 2012). A practical solution to CLIR is to translate source documents into the query language using an MT model, then use a monolingual IR system to match queries with translated documents. Correlation between MT and IR metrics is accomplished in the following steps:

1. Build a set of MT models and measure their performance using MT metrics.

2. Using each MT model in the set, translate all source documents to the target language, build an IR model, and measure IR performance on translated documents.

3. For each MT metric, find the correlation between the set of MT scores and their corresponding set of IR scores. The MT metric that has a

⁶Fluency and grammar, which are elicited with nearly identical directions (Gardent et al., 2017), are identically correlated.

⁷http://www.statmt.org/wmt19/metrics-task.html

⁸https://github.com/google-research/bleurt

Year	Pairs		*BLEU	BLEU	$MACROF_1$	$MICROF_1$	CHRF ₁
		Mean	.751	.771	.821	.818	.841
2019	18	Median	.782	.752	.844	.844	.875
		Wins	3	3	6	3	5
		Mean	.858	.857	.875	.873	.902
2018	14	Median	.868	.868	.901	.879	.919
		Wins	1	2	3	2	6
		Mean	.752	.713	.714	.742	.804
2017	13	Median	.758	.733	.735	.728	.791
		Wins	5	4	2	2	6

Table 3: WMT 2017–19 Metrics task: Mean and median Kendall's τ between MT metrics and human judgments. Correlations that are not significant at $\alpha = 0.05$ are excluded from the calculation of mean, and median, and wins. See Appendix Tables 9, 10, and 11 for full details. *BLEU is pre-computed scores available in the metrics packages. In 2018 and 2019, both MACROF₁ and MICROF₁ outperform BLEU, MACROF₁ outperforms MICROF₁. CHRF₁ has strongest mean and median agreements across the years. Judging based on the number of wins, MACROF₁ has steady progress over the years, and outperforms others in 2019.

stronger correlation with the IR metric(s) is more useful than the ones with weaker correlations.

4. Repeat the above steps on many languages to verify the generalizability of findings.

An essential resource of this analysis is a dataset with human annotations for computing MT and IR performances. We conduct experiments on two datasets: firstly, on data from the 2020 workshop on *Cross-Language Search and Summarization of Text and Speech* (CLSSTS) (Zavorin et al., 2020), and secondly, on data originally from Europarl, prepared by Lignos et al. (2019) (Europarl).

3.3.1 CLSSTS Datasets

CLSSTS datasets contain queries in English (EN), and documents in many source languages along with their human translations, as well as querydocument relevance judgments. We use three source languages: Lithuanian (LT), Pashto (PS), and Bulgarian (BG). The performance of this CLIR task is evaluated using two IR measures: Actual Query Weighted Value (AQWV) and Mean Average Precision (MAP). AQWV⁹ is derived from Actual Term Weighted Value (ATWV) metric (Wegmann et al., 2013).

We use a single CLIR system (Boschee et al., 2019) with the same IR settings for all MT models in the set, and measure Kendall's τ between MT and IR measures. The results, in Table 4, show that MACROF₁ is the strongest indicator of CLIR downstream task performance in five out of six settings. AQWV and MAP have a similar trend in agreement to the MT metrics. CHRF₁ and BLEURT, which are strong contenders when generated text is directly evaluated by humans, do not indicate

CLIR task performance as well as $MACROF_1$, as CLIR tasks require faithful meaning equivalence across the language boundary, and human translators can mistake fluent output for proper translations (Callison-Burch et al., 2007).

3.3.2 Europarl Datasets

We perform a similar analysis to Section 3.3.1 but on another cross-lingual task set up by Lignos et al. (2019) for Czech \rightarrow English (CS-EN) and German \rightarrow English (DE-EN), using publicly available data from the Europarl v7 corpus (Koehn, 2005). This task differs from the CLSSTS task (Section 3.3.1) in several ways. Firstly, MT metrics are computed on test sets from the news domain, whereas IR metrics are from the Europarl domain. The domains are thus intentionally mismatched between MT and IR tests. Secondly, since there are no queries specifically created for the Europarl domain, GOV2 TREC topics 701-850 are used as domain-relevant English queries. And lastly, since there are no query-document relevance human judgments for the chosen query and document sets, the documents retrieved by BM25 (Jones et al., 2000) on the English set for each query are treated as relevant documents for computing the performance of the CS-EN and DE-EN CLIR setup. As a result, IR metrics that rely on boolean query-document relevance judgments as ground truth are less informative, and we use Rank-Based Overlap (RBO; p = 0.98) (Webber et al., 2010) as our IR metric.

We perform our analysis on the same experiments as Lignos et al. (2019).¹⁰ NMT models for CS-EN and DE-EN translation are trained using a convolutional NMT architecture (Gehring

⁹https://www.nist.gov/system/files/documents-/2017/10/26/aqwv_derivation.pdf

¹⁰https://github.com/ConstantineLignos/ mt-clir-emnlp-2019

	Domain	IR Score	BLEU	$MACROF_1$	$MICROF_1$	$CHRF_1$	BLEURTmean	BLEURTmedian
	I.,	AQWV	.429	×.363	.508	×.385	.451	.420
LT-EN	In	MAP	.495	.429	.575	.451	.473	.486
LI-EN	In Ent	AQWV	×.345	.527	.491	.491	.491	.477
	In+Ext	MAP	×.273	×.455	×.418	×.418	×.418	×.404
-	In	AQWV	.559	.653	.574	.581	.584	.581
PS-EN	111	MAP	.493	.632	.487	.494	.558	.554
I S-LIN	In+Ext	AQWV	.589	.682	.593	.583	.581	.571
	III+EXI	MAP	.519	.637	.523	.482	.536	.526
	In	AQWV	×.455	.550	.527	×.382	×.418	.418
BG-EN	111	MAP	.491	.661	.564	.491	.527	.527
DO-EN	Interve	AQWV	×.257	.500	×.330	×.404	×.367	×.367
	In+ext	MAP	×.183	×.426	×.257	×.330	×.294	×.294

Table 4: CLSSTS CLIR task: Kendall's τ between IR and MT metrics under study. The rows with Domain=In are where MT and IR scores are computed on the same set of documents, whereas Domain=In+Ext are where IR scores are computed on a larger set of documents that is a superset of segments on which MT scores are computed. **Bold** values are the best correlations achieved in a row-wise setting; values with \times are *not* significant at $\alpha = 0.05$.

	BLEU	MACROF ₁	MICROF ₁	$CHRF_1$	$\overline{\mathrm{BT}}$	ΒŤ
CS-EN	.850	.867	.850	.850	.900	.867
DE-EN	.900	.900	.900	.912	.917	.900

Table 5: Europarl CLIR task: Kendall's τ between MT metrics and RBO. BT and BT are short for BLEURTmean and BLEURTmedian. All correlations are significant at $\alpha = 0.05$.

et al., 2017) implemented in the FAIRSeq (Ott et al., 2019) toolkit. For each of CS-EN and DE-EN, a total of 16 NMT models that are based on different quantities of training data and BPE hyperparameter values are used. The results in Table 5 show that BLEURT has the highest correlation in both cases. Apart from the trained BLEURTmedian metric, MACROF₁ scores higher than the others on CS-EN, and is competitive on CS-EN. MACROF₁ is not the metric with highest IR task correlation in this setting, unlike in Section 3.3.1, however it is competitive with BLEU and CHRF₁, and thus a safe choice as a downstream task performance indicator.

4 Spotting Qualitative Differences Between Supervised and Unsupervised NMT with MACROF₁

Unsupervised neural machine translation (UNMT) systems trained on massive monolingual data without parallel corpora have made significant progress recently (Artetxe et al., 2018; Lample et al., 2018a,b; Conneau and Lample, 2019; Song et al., 2019; Liu et al., 2020). In some cases, UNMT yields a BLEU score that is comparable with strong¹¹ supervised neural machine transla-

tion (SNMT) systems. In this section we leverage $MACROF_1$ to investigate differences in the translations from UNMT and SNMT systems that have similar BLEU.

We compare UNMT and SNMT for English \leftrightarrow German (EN-DE, DE-EN), English \leftrightarrow French (EN-FR, FR-EN), and English \leftrightarrow Romanian (EN-RO, RO-EN). All our UNMT models are based on XLM (Conneau and Lample, 2019), pretrained by Yang (2020). We choose SNMT models with similar BLEU on common test sets by either selecting from systems submitted to previous WMT News Translation shared tasks (Bojar et al., 2014, 2016) or by building such systems.¹² Specific SNMT models chosen are in the Appendix (Table 12).

Table 6 shows performance for these three language pairs using a variety of metrics. Despite comparable scores in BLEU and only minor differences in MICROF₁ and CHRF₁, SNMT models have consistently higher MACROF₁ and BLEURT than the UNMT models for all six translation directions.

In the following section, we use a pairwise maximum difference discriminator approach to compare corpus-level metrics BLEU and MACROF₁ on a segment level. Qualitatively, we take a closer look at the behavior of the two metrics when comparing a translation with altered meaning to a translation with differing word choices using the metric.

¹²We were unable to find EN-DE and DE-EN systems with comparable BLEU in WMT submissions so we built standard Transformer-base (Vaswani et al., 2017) models for these using appropriate quantity of training data to reach the desired BLEU performance. We report EN-RO results with diacritic removed to match the output of UNMT.

¹¹though not, generally, the strongest

	BLEU	MACROF ₁	MICROF ₁	CHRF ₁		BLEURTmedian
	SN UN Δ	SN UN Δ	SN UN Δ	SN UN Δ	SN UN Δ	SN UN Δ
DE-EN	32.7 33.9 -1.2	38.5 33.6 4.9	58.7 57.9 0.8	59.9 58.0 1.9	.211026 .24	.285 .067 .22
EN-DE	24.0 24.0 0.0	24.0 23.5 0.5	47.7 48.1 -0.4	53.3 52.0 1.3	134204 .07	112197 .09
FR-EN	31.1 31.2 -0.1	41.6 33.6 8.0	60.5 58.3 2.2	59.1 57.3 1.8	.182 .066 .17	.243 .154 .09
EN-FR	25.6 27.1 -1.5	31.9 27.3 4.6	53.0 52.3 0.7	56.0 57.7 -1.7	.104 .042 .06	.096 .063 .03
RO-EN	30.8 29.6 1.2	40.3 33.0 7.3	59.8 56.5 3.3	58.0 54.7 3.3	.004058 .06	.045004 .04
EN-RO	31.2 31.0 0.2	34.6 31.0 3.6	55.4 53.4 2.0	59.3 56.7 2.6	.030046 .08	.027038 .07

Table 6: For each language direction, UNMT (UN) models have similar BLEU to SNMT (SN) models, and $CHRF_1$ and $MICROF_1$ have small differences. However, $MACROF_1$ scores differ significantly, consistently in favor of SNMT. Both corpus-level interpretations of BLEURT support the trend reflected by $MACROF_1$, but the value differences are difficult to interpret.

$\delta_{ m MACROF_1}$	Fav	Analysis	$\delta_{ m Bleu}$	Fav	Analysis
.071	S	S: synonym; U: untranslation, noun	.048	S	S: word order; U: word order, untranslation,
					ending
.064	S	S: synonym; U: untranslation	.046	S	S: spelling variation; U: synonym, word order, punctuation
055	U	U: no issues; S: untranslation	.044	S	S: extra determiner; U: paraphrase, synonym, <i>number</i> , <i>untranslation</i>
.052	S	S: synonym; U: untranslation, noun	.042	S	S: synonym; U: synonym, punctuation, extra adverb
045	U	U: no issues; S: untranslation	039	U	U: no issues; S: noun, verb
.044	S	S: synonym, word order; U: <i>subject</i> , <i>trunca-</i> <i>tion</i> , word order	037	U	U: no issues; S: punctuation
.044	S	S: synonym, tense; U: untranslation	034	U	U: no issues; S: symbol
.043	S	S: inflection, word order; U: number	032	U	U: no issues; S: adjective, noun
041	U	U: adjective, verb; S: omitted verb, untransla- tion	032	U	U: <i>untranslation</i> ; S: tense, <i>word order, mean-</i> <i>ing</i> , active/passive voice
.041	S	S: time, word order; U: time, nouns	031	U	U: <i>untranslation</i> ; S: word order, synonym, <i>ex-tra_conj</i>

Table 7: Analysis of the ten DE-EN test set segments with the most favoritism in SNMT (S) or UNMT (U), according to $MACROF_1$ (left) and BLEU (right). Fav is the favored system by metrics. The complete text of the sentences is in the Appendix, Tables 15 and 16.

4.1 Pairwise Maximum Difference Discriminator

We consider cases where a metric has a strong opinion of one translation system over another, and analyze whether the opinion is well justified. In order to obtain this analysis, we employ a pairwise segment-level discriminator from within a corpuslevel metric, which we call *favoritism*.

We extend the definition of *T* from Section 2 to $T = \{x, h_S, h_U, y\}$ where each of h_S and h_U is a separate system's hypothesis set for x.¹³ Let *M* be a corpus-level measure such that $M(h, y) \in \mathbb{R}$ and a higher value implies better translation quality. $M(h^{(-i)}, y^{(-i)})$ is the corpus-level score obtained by excluding $h^{(i)}$ and $y^{(i)}$ from *h* and *y*, respectively. We define the *benefit* of segment *i*, $\delta_M(i;h)$:

$$\delta_M(i;h) = M(h,y) - M(h^{(-i)}, y^{(-i)})$$

If $\delta_M(i;h) > 0$, then *i* is beneficial to *h* with respect to *M*, as the inclusion of $h^{(i)}$ increases the corpus-

level score. We define the *favoritism* of *M* toward *i* as $\delta_M(i;h_S,h_U)$:

$$\delta_M(i;h_S,h_U) = \delta_M(i;h_S) - \delta_M(i;h_U) \qquad (1)$$

If $\delta_M(i;h_S,h_U) > 0$ then *M* favors the translation of $x^{(i)}$ by system *S* over that in system *U*.

Table 7 reflects the results of a manual examination of the ten sentences in the DE-EN test set with greatest magnitude favoritism; complete results are in the Appendix, Tables 15 and 16. Meaningaltering changes such as *'untranslation'*, (wrong) *'time'*, and (wrong) *'translation'* are marked in *italics*, while changes that do not fundamentally alter the meaning, such as 'synonym,' (different) 'inflection,' and (different) 'word order' are marked in plain text.¹⁴

The results indicate that $MACROF_1$ generally favors SNMT, and with good reasons, as the favored translation does not generally alter sentence meaning, while the disfavored translation does. On

¹³The subscripts represent SNMT and UNMT in this case, though the definition is general.

¹⁴Some changes, such as 'word order' may change meaning; these are italicized or not on a case-by-case basis.

6^{th}	$\delta_{MACROF1}(i,h_S,h_U)$: .044, $\delta_{BLEU}(i,h_S,h_U)$:00087, $\delta_{BLEURT}(i,h_S,h_U)$: .97					
Ref	Ever since I joined Labour 32 years ago as a school pupil, provoked by the Thatcher government's neglect that had left my					
	comprehensive school classroom literally falling down, I've sought to champion better public services for those who need them most					
	- whether as a local councillor or government minister.					
SNMT	32 years ago, I joined Labour as a student because of the neglect of the Thatcher government, which had led to my classroom literally					
	collapsed, and as a result I tried to promote better public services for those who need it most, whether as a local council or ministers.					
UNMT	Last 32 years ago, as a student, because of the disdain for the Thatcher-era government, Labour joined Labour.					
Problems	SNMT: synonym, word_order UNMT: subject, truncation, word_order					

Table 8: An example of favoritism that illustrates the differences between $MACROF_1$ and BLEU. Translations of the DE-EN test sentence with sixth largest magnitude favoritism according to $MACROF_1$, along with the favoritism according to BLEU (not in the top ten). UNMT's translation does not include the second half of the sentence. $MACROF_1$ favors SNMT, but BLEU favors UNMT.

the other hand, for the ten most favored sentences according to BLEU, four do not contain meaningaltering divergences in the disfavored translation. Importantly, none of the sentences with greatest favoritism according to MACROF₁, all of which having meaning altering changes in the disfavored alternatives, appears in the list for BLEU. This indicates relatively bad judgment on the part of BLEU. One case of good judgment from MACROF₁ and bad judgment from BLEU regarding truncation is shown in Table 8.

From our qualitative examinations, MACROF₁ is better than BLEU at discriminating against untranslations and trucations in UNMT. The case is similar for FR-EN and RO-EN, except that RO-EN has more untranslations for both SNMT and UNMT, possibly due to the smaller training data. Complete tables and annotated sentences are in the Appendix, in Section C.

5 Related Work

5.1 MT Metrics

Many metrics have been proposed for MT evaluation, which we broadly categorize into modelfree or model-based. Model-free metrics compute scores based on translations but have no significant parameters or hyperparameters that must be tuned a priori; these include BLEU (Papineni et al., 2002), NIST (Doddington, 2002), TER (Snover et al., 2006), and CHRF₁ (Popović, 2015). Modelbased metrics have a significant number of parameters and, sometimes, external resources that must be set prior to use. These include METEOR (Banerjee and Lavie, 2005), BLEURT (Sellam et al., 2020), YiSi (Lo, 2019), ESIM (Mathur et al., 2019), and BEER (Stanojević and Sima'an, 2014). Model-based metrics require significant effort and resources when adapting to a new language or domain, while model-free metrics require only a test

set with references.

Mathur et al. (2020) have recently evaluated the utility of popular metrics and recommend the use of either CHRF₁ or a model-based metric instead of BLEU. We compare our MACROF₁ and MICROF₁ metrics with BLEU, CHRF₁, and BLEURT (Sellam et al., 2020). While Mathur et al. (2020) use Pearson's correlation coefficient (r) to quantify the correlation between automatic evaluation metrics and human judgements, we instead use Kendall's rank coefficient (τ), since τ is more robust to outliers than r (Croux and Dehon, 2010).

5.2 Rare Words are Important

That natural language word types roughly follow a Zipfian distribution is a well known phenomenon (Zipf, 1949; Powers, 1998). The frequent types are mainly so-called "stop words," function words, and other low-information types, while most content words are infrequent types. To counter this natural frequency-based imbalance, statistics such as inverted document frequency (IDF) are commonly used to weigh the *input* words in applications such as information retrieval (Jones, 1972). In NLG tasks such as MT, where words are the output of a classifier, there has been scant effort to address the imbalance. Doddington (2002) is the only work we know of in which the 'information' of an ngram is used as its weight, such that rare n-grams attain relatively more importance than in BLEU. We abandon this direction for two reasons: Firstly, as noted in that work, large amounts of data are required to estimate n-gram statistics. Secondly, unequal weighing is a bias that is best suited to datasets where the weights are derived from, and such biases often do not generalize to other datasets. Therefore, unlike Doddington (2002), we assign equal weights to all n-gram classes, and in this work we limit our scope to unigrams only.

While BLEU is a precision-oriented measure,

METEOR (Banerjee and Lavie, 2005) and CHRF (Popović, 2015) include both precision and recall, similar to our methods. However, neither of these measures try to address the natural imbalance of class distribution. BEER (Stanojević and Sima'an, 2014) and METEOR (Denkowski and Lavie, 2011) make an explicit distinction between function and content words; such a distinction inherently captures frequency differences since function words are often frequent and content words are often infrequent types. However, doing so requires the construction of potentially expensive linguistic resources. This work does not make any explicit distinction and uses naturally occurring type counts to effect a similar result.

5.3 F-measure as an Evaluation Metric

F-measure (Rijsbergen, 1979; Chinchor, 1992) is extensively used as an evaluation metric in classification tasks such as part-of-speech tagging, named entity recognition, and sentiment analysis (Derczynski, 2016). Viewing MT as a multi-class classifier is a relatively new paradigm (Gowda and May, 2020), and evaluating MT solely as a multiclass classifier as proposed in this work is not an established practice. However, we find that the F_1 measure is sometimes used for various analyses when BLEU and others are inadequate: The compare-mt tool (Neubig et al., 2019) supports comparison of MT models based on F_1 measure of individual types. Gowda and May (2020) use F_1 of individual types to uncover frequency-based bias in MT models. Sennrich et al. (2016) use corpuslevel unigram F_1 in addition to BLEU and CHRF, however, corpus-level F_1 is computed as MICROF₁. To the best of our knowledge, there is no previous work that clearly formulates the differences between micro- and macro- averages, and justifies the use of MACROF₁ for MT evaluation.

6 Discussion and Conclusion

We have evaluated NLG in general and MT specifically as a multi-class classifier, and illustrated the differences between micro- and macro- averages using MICROF₁ and MACROF₁ as examples (Section 2). MACROF₁ captures semantic adequacy better than MICROF₁ (Section 3.1). BLEU, being a micro-averaged measure, served well in an era when generating fluent text was at least as difficult as generating adequate text. Since we are now in an era in which fluency is taken for granted and semantic adequacy is a key discriminating factor, macroaveraged measures such as MACROF1 are better at judging the generation quality of MT models (Section 3.2). We have found that another popular metric, CHRF1, also performs well on direct assessment, however, being an implicitly micro-averaged measure, it does not perform as well as MACROF₁ on downstream CLIR tasks (Section 3.3.1). Unlike BLEURT, which is also adequacy-oriented, MACROF₁ is directly interpretable, does not require retuning on expensive human evaluations when changing language or domain, and does not appear to have uncontrollable biases resulting from data effects. It is both easy to understand and to calculate, and is inspectable, enabling fine-grained analysis at the level of individual word types. These attributes make it a useful metric for understanding and addressing the flaws of current models. For instance, we have used MACROF₁ to compare supervised and unsupervised NMT models at the same operating point measured in BLEU, and determined that supervised models have better adequacy than the current unsupervised models (Section 4).

Macro-average is a useful technique for addressing the importance of the long tail of language, and MACROF₁ is our first step in that direction; we anticipate the development of more advanced macroaveraged metrics that take advantage of higherorder and character n-grams in the future.

7 Ethical Consideration

Since many ML models including NMT are themselves opaque and known to possess data-induced biases (Prates et al., 2019), using opaque and biased evaluation metrics in concurrence makes it even harder to discover and address the flaws in modeling. Hence, we have raised concerns about the opaque nature of the current model-based evaluation metrics, and demonstrated examples displaying unwelcome biases in evaluation. We advocate the use of the MACROF₁ metric, as it is easily interpretable and offers the explanation of score as a composition of individual type performances. In addition, MACROF₁ treats all types equally, and has no parameters that are directly or indirectly estimated from data sets. Unlike MACROF₁, MICROF₁ and other implicitly or explicitly micro-averaged metrics assign lower importance to rare concepts and their associated rare types. The use of microaveraged metrics in real world evaluation could lead to marginalization of rare types.

Failure Modes: The proposed MACROF₁ metric is not the best measure of fluency of text. Hence we suggest caution while using MACROF₁ to draw fluency related decisions. MACROF₁ is inherently concerned with *words*, and assumes the output language is easily segmentable into word tokens. Using MACROF₁ to evaluate translation into alphabetical languages such as Thai, Lao, and Khmer, that do not use white space to segment words, requires an effective tokenizer. Absent this the method may be ineffective; we have not tested it on languages beyond those listed in Section B.

Reproducibility: Our implementation of $MACROF_1$ and $MICROF_1$ has the same user experience as BLEU as implemented in SACRE-BLEU; signatures are provided in Section A. In addition, our implementation is computationally efficient, and has the same (minimal) software and hardware requirements as BLEU. All data for MT and NLG human correlation studies is publicly available and documented. Data for reproducing the IR experiments in Section 3.3.2 is also publicly available and documented. The data for reproducing the IR experiments in Section 3.3.1 is only available to participants in the CLSSTS shared task.

Climate Impact: Our proposed metrics are on par with BLEU and such model-free methods, which consume significantly less energy than most model-based evaluation metrics.

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A Metrics Reproducibility

BLEU scores reported in this work are computed with the SACREBLEU library and have signature BLEU+case.mixed+lang.<xx>-<yy>+numrefs.1 +smooth.exp+tok.<TOK>+version.1.4.13, where <TOK> is zh for Chinese, and 13a for all other languages. MACROF₁ and MICROF₁ use the same tokenizer as BLEU. CHRF₁ is also obtained using SACREBLEU and has signature chrF1+lang.<xx>-<yy>+numchars.6 +space.false +version.1.4.13. BLUERT scores are from the *base* model of Sellam et al. (2020), which is fine-tuned on WMT Metrics ratings data from 2015-2018. The BLEURT model is retrieved from https://storage.googleapis.com/ bleurt-oss/bleurt-base-128.zip.

 $MACROF_1$ and $MICROF_1$ are computed using our fork of SACREBLEU as:

sacrebleu \$REF -m macrof microf < \$HYP.</pre>

B Agreement with WMT Human Judgments

Tables 9, 10, and 11 provide τ between MT metrics and human judgments on WMT Metrics task 2017– 2019. *BLEU is based on pre-computed scores in WMT metrics package, whereas BLEU is based on our recalculation using SACREBLEU. Values marked with [×]are not significant at $\alpha = 0.05$, and hence corresponding rows are excluded from the calculation of mean, median, and standard deviation.

Since MACROF₁ is the only metric that does not achieve statistical significance in the WMT 2019 EN-ZH setting, we carefully inspected it. Human scores for this setting are obtained without looking at the references by bilingual speakers (Ma et al., 2019), but the ZH references are found to have a large number of bracketed EN phrases, especially proper nouns that are rare types. When the text inside these brackets is not generated by an MT system, MACROF₁ naturally penalizes heavily due to the poor recall. Since other metrics assign lower importance to poor recall of such rare types, they achieve relatively better correlation to human scores than MACROF₁. However, since the τ values for EN-ZH are relatively lower than the other language pairs, we conclude that poor correlation of MACROF1 in EN-ZH is due to poor quality references. Some settings did not achieve statistical significance due to a smaller sample set as there were fewer MT systems submitted, e.g. 2017 CS-EN.

	*BLEU	BLEU	MACROF ₁	MICROF ₁	ChrF ₁
DE-CS	.855	.745	.964	.917	.982
DE-EN	.571	.655	.723	.695	.742
DE-FR	.782	.881	.927	.844	.915
EN-CS	.709	.954	.927	.927	.908
EN-DE	.540	.752	.741	.773	.824
EN-FI	.879	.818	.879	.848	.923
EN-GU	.709	.709	.600	.734	.709
EN-KK	.491	.527	.685	.636	.661
EN-LT	.879	.848	.970	.939	.881
EN-RU	.870	.848	.939	.879	.930
FI-EN	.788	.809	.909	.901	.875
FR-DE	.822	.733	.733	.764	.815
GU-EN	.782	.709	.855	.891	.945
KK-EN	.891	.844	.796	.844	.881
LT-EN	.818	.855	.844	.855	.833
RU-EN	.692	.729	.714	.780	.757
ZH-EN	.695	.695	.752	.676	.715
Median	.782	.752	.844	.844	.875
Mean	.751	.771	.821	.818	.841
SD	.124	.101	.112	.093	.095
EN-ZH	.606	.606	×.424	.595	.594
Wins	3	3	6	3	5

Table 9: WMT19 Metrics task: Kendall's τ between metrics and human judgments.

C UNMT and SNMT Models

The UNMT models follow XLM's standard architecture and are trained with 5 million monolingual sentences for each language using a vocabulary size of 60,000. We train SNMT models for $EN \leftrightarrow DE$ and select models with the most similar (or a slightly lower) BLEU as their UNMT counterparts on newstest2019. The DE-EN model selected is trained with 1 million sentences of parallel data and a vocabulary size of 64,000, and the EN-DE model selected is trained with 250,000 sentences of parallel data and a vocabulary size of 48,000. For $EN \leftrightarrow FR$ and $EN \leftrightarrow RO$, we select SNMT models from submitted systems to WMT shared tasks that have similar or slightly lower BLEU scores to corresponding UNMT models, based on NewsTest2014 for EN \leftrightarrow FR and *NewsTest2016* for EN \leftrightarrow RO.

Figure 1, which is a visualization of $MACROF_1$ for SNMT and UNMT models, shows that UNMT is generally better than SNMT on frequent types, however, SNMT outperforms UNMT on the rest leading to a crossover point in $MACROF_1$ curves. Since $MACROF_1$ assigns relatively higher weights to infrequent types than in BLEU, SNMT gains higher $MACROF_1$ than UNMT while both have approximately the same BLEU, as reported in Table 6.

A complete comparison of UNMT vs SNMT in different languages is in Table 12. A manual analysis of the ten sentences with the largest magnitude favoritism according to MACROF₁ and BLEU in

		$\star BLEU$	Bleu	$MACROF_1 \\$	$MICROF_1 \\$	CHRF ₁
-	DE-EN	.828	.845	.917	.883	.919
	EN-DE	.778	.750	.850	.783	.848
	EN-ET	.868	.868	.934	.906	.949
	EN-FI	.901	.848	.901	.879	.945
	EN-RU	.889	.889	.944	.889	.930
	EN-ZH	.736	.729	.685	.833	.827
	ET-EN	.884	.900	.884	.878	.904
	FI-EN	.944	.944	.889	.915	.957
	RU-EN	.786	.786	.929	.857	.869
	ZH-EN	.824	.872	.738	.780	.820
	EN-CS	1.000	1.000	.949	1.000	.949
	Median	.868	.868	.901	.879	.919
	Mean	.858	.857	.875	.873	.902
	SD	.077	.080	.087	.062	.052
	TR-EN	×.200	×.738	×.400	×.316	×.632
	EN-TR	×.571	×.400	.837	×.571	.849
	CS-EN	×.800	×.800	×.600	×.800	×.738
	Wins	1	2	3	2	6

Table 10: WMT18 Metrics task: Kendall's τ between metrics and human judgments.

	*Bleu	BLEU	$MACROF_1$	$MICROF_1$	CHRF ₁
DE-EN	.564	.564	.734	.661	.744
EN-CS	.758	.751	.767	.758	.878
EN-DE	.714	.767	.562	.593	.720
EN-FI	.667	.697	.769	.718	.782
EN-RU	.556	.556	.778	.648	.669
EN-ZH	.911	.911	.600	.854	.899
LV-EN	.905	.714	.905	.905	.905
RU-EN	.778	.611	.611	.722	.800
TR-EN	.911	.778	.674	.733	.907
ZH-EN	.758	.780	.736	.824	.732
Median	.758	.733	.735	.728	.791
Mean	.752	.713	.714	.742	.804
SD	.132	.110	.103	.097	.088
FI-EN	.867	.867	×.733	.867	.867
EN-TR	.857	.714	×.571	.643	.849
CS-EN	×1.000	×1.000	×.667	×.667	×.913
Wins	5	4	2	2	6

Table 11: WMT17 Metrics task: Kendall's τ between metrics and human judgments.

SNMT	UNMT	SNMT Name
32.7	33.9	Our Transformer
24.0	24.0	Our Transformer
31.1	31.2	OnlineA.0
25.6	27.1	PROMT-Rule-based.3083
30.8	29.6	Online-A.0
31.2	31.0	uedin-pbmt.4362
	32.7 24.0 31.1 25.6 30.8	24.024.031.131.225.627.130.829.6

Table 12: SNMT systems are selected such that their BLEU scores are approximately the same as the available pretrained UNMT models.

the FR-EN and RO-EN test sets is in Table 13 and Table 14. The complete texts of these sentences, their reference translations, and the system translations (including DE-EN mentioned in Sec 4), are shown in Tables 15, 16, 17, 18, 19, and 20.



Figure 1: SNMT vs UNMT MACROF₁ on the most frequent 500 types. UNMT outperforms SNMT on frequent types that are weighed heavily by BLEU however, SNMT is generally better than UNMT on rare types; hence, SNMT has a higher MACROF₁.

_	$\delta_{\mathrm{MACROF}_1}$	Fav	Analysis	$\delta_{ m Bleu}$	Fav	Analysis
	.044	S	S: synonym; U: <i>untranslation</i> , synonym	026	U	U: synonym; S: omitted adv, word order
	038	U	U: no issues; S: synonym	.025	S	S: no issues; U: determiner, word order
	.035	S	S: synonym; U: <i>untranslation</i> , synonym	.024	S	S: no issues; U: repetition, form
	034	U	U: no issues; S: synonym; word_order	.021	S	S: verb, synonym; U: untranslation, noun,
						<i>time</i> , synonym
	034	U	U: synonym; S: word order, verb_ref	.021	S	S: synonym; U: synonym
	033	U	U: no issues; S: synonym	021	U	U: omitted NER; S: synonym, word order
	.033	S	S: word order; U: <i>untranslation</i> , NER, word	021	U	U: untranslation; S: verb, word order
	order					
	.032	S	S: synonym; U: number, omitted noun, un-	021	U	U: synonym; S: extra preposition, synonym,
			translation, verb			word order
	.030	S	S: adj; U: untranslation	.021	S	S: no issues; U: NER
	.030	S	S: noun, synonym; U: noun, synonym	020	U	U: synonym; S: synonym, word order

Table 13: Analysis of the ten FR-EN test set segments with the most favoritism in SNMT (S) or UNMT (U), according to $MACROF_1$ (left) and BLEU (right). Fav is the favored system by metrics. Actual examples are shown in Appendix Tables 17 and 18.

δ_{MACROF_1}	Fav	Analysis	$\delta_{\rm BLEU}$	Fav	Analysis
1		7			
.131	S	S: word order; U: repetition, word order	.114	S	S: word order; U: repetition, word order
.063	S	S: noun, word order; U: repetition, untrans-	.089	S	S: no issues; U: omitted noun, omitted time,
		lation, noun			NER
.062	S	S: extra, untranslation; U: untranslation,	072	U	U: country, untranslation; S: noun, word
		сору			order
052	U	U: <i>untranslation x 3</i> , synonym; S: <i>untrans</i> -	045	U	U: synonym; S: synonym, word order
		<i>lation</i> , synonym			
052	U	U: untranslation, NER, synonym; S: NER,	041	U	U: untranslation; S: word order, subject
		synonym			,,
052	U	U: extra; S: untranslation	040	U	U: no issues; S: number, omitted preposition
050	Ū	U: adv; S: incoherent, adv	.039	S	S: extra, untranslation; U: untranslation,
.050	U	o. auv, o. meonereni, auv	.057	5	copy
050	U	U: active/passive voice, name; S: name	.036	S	S: no issues; U: <i>extra verb</i>
049	U	U: untranslation; S: untranslation, word or-	035	U	U: repetition, untranslation, S: verb, syn-
1015	-	der		-	onym, word order
.048	S	S: no issues; U: NER	.034	S	S: synonym; U: untranslation

Table 14: Analysis of the ten RO-EN test set segments with the most favoritism in SNMT (S) or UNMT (U), according to $MACROF_1$ (left) and BLEU (right). Fav is the favored system by metrics. Actual examples are shown in Appendix Tables 19 and 20.

$\delta_{ m MACROF_1}$	Source	Reference	SNMT	UNMT
	Es wird davon ausgegan- gen, dass sie über eine leistungsstarke Kanone, eine Reihe von Flugabwehr- und Schiffsabwehrraketen sowie einige Stealth-Technologien verfügen, wie z. B. re- duzierte Radar-, Infrarot- und akustische Signaturen.	It is understood they will feature a powerful cannon, an array of anti-aircraft and anti-ship missiles as well as some stealth technologies, such as reduced radar, in- frared and acoustic signa- tures.	It is assumed that they have a powerful cannon, a series of anti-aircraft and anti-ship missiles, as well as some steam technologies, such as reduced radar, infrared and acoustic signatures.	It is understood they have a powerful canon, a number of fluke and ship fire systems and some stealth-controlled technologies, such as re- duced radar, infrarot and akustic signposts.
.064	Eine Gruppe maskierter Pro- Separatisten, die von der Bere- itschaftspolizei zurückgehal- ten wurden, bewarfen sie mit Eiern und schleuderte Pulver- farbe und erzeugte in den Straßen, die normalerweise von Touristen überfüllt waren, dunkle Staubwolken.	A group of masked pro- separatists held back by riot police pelted them with eggs and hurled powder paint, creating dark clouds of dust in streets that would usually be thronged with tourists.	A group of masked pro- separatists held hostage by the riot police brought them to eggs and ignited pow- der paint and produced dark clouds in the streets that were usually crowded by tourists.	A group of masked pro- independence separforces, who were kept away by the Bereitschaftpolice, be- headed them with evocative paint and poured pulver- force paint and created dark Staubations in the streets normally clogged by tourists.
055	Il faut bien le faire.	Il faut bien le faire.	La gentillesse du personnel et la disponibil.	Il faut bien faire.
.052	In einem Abschnitt gibt es ein Bild eines Schlafsaals, in dem die Studenten auf Kaf- feetassen, Vorhänge, Trainer und Bücher klicken, um über die Auswirkungen von Kof- fein und Licht informiert zu werden und darüber, wie sich die sportliche Leistung durch Schlafmangel und die Bedeu- tung einer Schlafenszeitrou- tine beeinflusst.	In one section there is an image of a dorm room, where students click on cof- fee cups, curtains, trainers and books to be told about the effects of caffeine and light and how athletic perfor- mance is impacted by sleep deficiency, and the impor- tance of a bedtime routine.	In a section there is an im- age of a bedroom where stu- dents click on coffee cups, curtains, trainers, and books to be informed about the ef- fects of caffeine and light, and about how sporting per- formance is affected by lack of sleep and the importance of sleeping time routine.	In one section, there is a picture of a sleeping sauna where students click on coffee cups, forecourts, coaches and books to be educated about the impact of Koffein and light and about how athletic performance is influenced by sleep loss and the importance of a sleep day routine.
045		Nickel mining is also impor- tant in the province, but is mostly concentrated in Mo- rowali, on the opposite coast of Sulawesi.	Nickelergbau is also impor- tant in the province, but is mainly operated in More- wali, on the opposite coast of Sulawesi.	Nickel mining is also im- portant in the province, but is mostly operated in Mo- rowali, on the opposite coast of Sulawesi.
.044		Ever since I joined Labour 32 years ago as a school pupil, provoked by the Thatcher government's neglect that had left my comprehensive school class- room literally falling down, I've sought to champion better public services for those who need them most - whether as a local councillor or government minister.	32 years ago, I joined Labour as a student be- cause of the neglect of the Thatcher government, which had led to my classroom lit- erally collapsed, and as a re- sult I tried to promote bet- ter public services for those who need it most, whether as a local council or ministers.	Last 32 years ago, as a stu- dent, because of the disdain for the Thatcher-era govern- ment, Labour joined Labour.
.044	UN-Gesandter Staffan de Mistura hofft, bald die er- sten Treffen eines neuen Ausschusses aus Regierungs- und Oppositionsmitgliedern einzuberufen, um eine Nachkriegsverfassung für Syrien zu entwerfen und den Weg zu Wahlen zu ebnen.	UN envoy Staffan de Mis- tura is hoping to soon con- vene the first meetings of a new committee comprised of government and opposi- tion members to draft a post- war constitution for Syria and pave the way to elec- tions.	UN envoy Staffan de Mis- tura hopes to convene soon the first meetings of a new committee of government and opposition members to draw up a post-war constitu- tion for Syria and pave the way for elections.	U.N. Secretary General Staffan de Mistura hopes to soon join the first meet- ings of a new committee of government and oppo- sition leaders to design a Nachkriegsrewrite for Syria and clear the path to elections.
.043	CBS hatte 3,1 Millio- nen, NBC 2,94 Millionen, MSNBC 2,89 Millionen und CNN 2,52 Millionen, so Nielsen.	CBS had 3.1 million, NBC had 2.94 million, MSNBC had 2.89 million and CNN had 2.52 million, Nielsen said.	CBS had 3.1 million, NBC 2.94 million, MSNBC 2.89 million and CNN 2.52 mil- lion, says Nielsen.	CBS had 3.8 million, NBC 3.94 million, MSNBC 3.89 million and CNN 3.52 mil- lion, Nielsen said.
041	Den Rangers gelangen nur zwei Schüsse in der ersten Hälfte, aber der ehemalige Ibrox-Torhüter Liam Kelly war kaum von Lassana Coulibalys Kopfsprung und dem Treffer eines bisslosen Ovie Ejaria aus der Ruhe zu bringen.	Rangers managed just two first-half shots on target but former Ibrox goalkeeper Liam Kelly was barely trou- bled by Lassana Coulibaly's header and a tame Ovie Ejaria strike.	The Ranners only reach two shots in the first half, but the former Ibrox-Torkeeper Liam Kelly was hardly the head of Lassanna Coulibys and the hit of a bissloze Ovi Ejaria.	The Rangers managed only two shots in the first half but former Ibrox goalkeeper Liam Kelly was unlikely to be helped by Lassana Coulibaly's headfirst tackle and the goal of a bisected Ovie Ejaria.
.041	Liverpool tritt am MIttwoch um 15.00 Uhr im Stadio San Paolo in Neapel, Italien, gegen Napoli an.	Liverpool battles Napoli in the group stage of the Cham- pions League at 3 p.m. on Wednesday at Stadio San Paolo in Naples, Italy.	Liverpool will take place at 3 p.m. at the Stadio San Paolo in Naples, Italy, against Napoli.	Liverpool v Napoli at the MItch Stadium at 15.00 pm in Neapel, Italy, on MItch.

Table 15: Top 10 segments by $|\delta_{MACROF_1}(i, h_S, h_U)|$ on DE-EN.

δ_{BLEU} Source	Reference	SNMT	UNMT
.048 In der letzten Woche wurden mittlere Konzentrationen in Küstennähe und auf offener See in Pinellas County gemeldet, geringe bis hohe Konzen- trationen auf offener See in Hillsbor- ough County, Hintergrund- bis hohe Konzentrationen in Manatee County, Hintergrund- bis hohe Konzentratio- nen in Küstennähe und auf offener See in Sarasota County, Hintergrund- bis mittlere Konzentrationen in Char- lotte County, Hintergrund- bis hohe Konzentrationen in Küstennähe und auf hoher See in Lee County sowie geringe Konzentrationen in Collier County.	Medium concentrations in or offshore of Pinellas County have been reported in the past week, low to high concentrations offshore of Hillsborough County, back- ground to high concentra- tions in Manatee County, background to high concen- trations in or offshore of Sarasota County, background to medium concentrations in Charlotte County, back- ground to high concentra- tions in or offshore of Lee County, and low concentra- tions in Collier County.	Last week, average concen- trations were reported on the coast and open seas in Pinel- las County, low to high lev- els at open sea in Hillsbor- ough County, background to high concentrations in Man- tee County, high concentra- tions in coastal and open seas in Sarsota County, back- ground to medium concen- trations in Charlotte County, background to high shore and high sea levels in Lee County, and low concentrations in Collier County.	In the last week, moder- ate to high Konzentrof lead in Küstas County were re- ported in Pinellas County, low to high Konzentrof lead levels on open water in Hillsborough County, Hinter- grundto high levels in Man- atee County, Hintergrundto high to high Konzentrin Küs- tas and on open water in Sara- sota County and low Konzen- trationen in Charlotte County, Hintergrundto high to high Konzentrin Küstennähe and on open water in Sarasota County.
.046 Moskau hat wiederholt betont, dass die 11-Milliarden-Dollar-Pipeline Nord Stream 2, die die bestehende Pipeline-Kapazität auf 110 Milliar- den Kubikmeter verdoppeln soll, ein rein wirtschaftliches Projekt ist.	Moscow has repeatedly stressed that the \$11 billion Nord Stream 2 pipeline, which is set to double the existing pipeline capacity to 110 billion cubic meters, is a purely economic project.	Moscow has repeatedly stressed that the \$11 billion Nord Stream 2 pipeline, which is supposed to double the existing pipeline capacity to 110 billion cubic metres, is a purely economic project.	Moscow has repeatedly insisted that the 11-billion pipeline, Nord Stream 2, which will double the exist- ing Pipeline-capacity to 110 billion cubic feet, is a purely commercial project.
.044 Der NTS, der für die Betreuung von mehr als 270 historischen Gebäu- den, 38 wichtigen Gärten und 76.000 Hektar Land rund um das Land ve- rantwortlich ist, nimmt die Fleder- mäuse sehr ernst.	The NTS, which is respon- sible for the care of more than 270 historical build- ings, 38 important gardens and 76,000 hectares of land around the country, takes bats very seriously.	The NTS, which is responsi- ble for the care of more than 270 historic buildings, 38 im- portant gardens and 76,000 hectares of land around the country, takes the bats very seriously.	The NTS, responsible for managing more than 270 his- toric buildings, 38 key gar- dens and 74,000 acres of land around the country, said the Fledermäuse are very impor- tant.
.042 George W. Bush telefonierte mit Sen- atoren, um diese zu überreden, Herrn Kavanaugh zu unterstützen, der im Weißen Haus für Herrn Bush gear- beitet hatte und durch ihn seine Frau Ashley traf, die die persönliche Sekretärin von Herrn Bush war.	George W. Bush has been picking up the phone to call Senators, lobbying them to support Mr Kavanaugh, who worked in the White House for Mr Bush and through him met his wife Ashley, who was Mr Bush's personal sec- retary.	George W. Bush contacted senators to persuade them to support Mr Kavanaugh, who worked in the White House for Mr Bush and met his wife Ashley, who was Mr Bush's personal secretary.	George W. Bush spoke to sen- ators to help him overture to support Mr. Kavanaugh, who had worked in the White House for Mr. Bush and met through him his wife, Ashley, who was the personal secre- tary to Mr. Bush.
039 Eine Woche nachdem eine of- fizielle chinesische Zeitung eine vierseitige Anzeige in einer US- amerikanischen Tageszeitung auf den gegenseitigen Nutzen des US- China-Handels gestellt hatte, warf der US-amerikanische Botschafter in China Peking vor, die amerikanis- che Presse zur Verbreitung von Propaganda zu verwenden.	A week after an official Chi- nese newspaper ran a four- page ad in a U.S. daily tout- ing the mutual benefits of U.SChina trade, the U.S. ambassador to China accused Beijing of using the Amer- ican press to spread propa- ganda.	A week after an official Chi- nese newspaper published a four-page display in a US daily on the mutual benefit of US China trade, the US am- bassador to China published in Beijing to use the Ameri- can press for propaganda.	A week after an official Chi- nese newspaper published a four-page ad on the mutual benefit of the US-China trade, the U.S. ambassador to China accused Beijing of using the American press to spread pro- paganda.
037 Sie kümmern sich nicht darum, wen sie verletzen, wen sie überfahren müssen, um Macht und Kontrolle zu bekommen, das ist, was sie wollen, Macht und Kontrolle, wir werden es ihnen nicht geben.	They don't care who they hurt, who they have to run over in order to get power and control, that's what they want is power and control, we're not going to give it to them."	They do not care about who they hurt whom they must pass over to gain power and control, that is what they want, power and control, we will not give them.	They don't care who they hurt, who they have to pass to get power and control, that's what they want, power and control, we won't give it to them.
034 Mayorga behauptet, Ronaldo sei nach dem angeblichen Vorfall auf die Knie gefallen und habe ihr gesagt, er sei "zu 99 Prozent" ein "guter Kerl", der von den "ein Prozent" im Stich gelassen wurde.	Mayorga claims Ronaldo fell to his knees after the alleged incident and told her he was "99 percent" a "good guy" let down by the "one percent."	Mayorga claims that Ronaldo fell to the knees after the al- leged incident, saying that he was "99%" a" good guy" left in the lurch by the "one percent".	Mayorga claims Ronaldo fell on his knee after the alleged incident and told her he was "to 99 percent" a "good guy" who was left in the dark by the "one percent."
032 Palin, 29, aus Wasilla, Alaska, wurde wegen des Verdachts auf häus- liche Gewalt verhaftet. Gegen ihn liegt bereits ein Bericht über häus- liche Gewalt und Widerstand bei der Festnahme vor, so eine Meldung, die am Samstag von den Alaska State Troopers veröffentlicht wurde.	Palin, 29, of Wasilla, Alaska, was arrested on suspicion of domestic violence, interfer- ing with a report of domes- tic violence and resisting ar- rest, according to a report released Saturday by Alaska State Troopers.	Palin, 29, from Wasilla, Alaska, was arrested for alleged domestic violence, and a report on domestic violence and opposition to arrest has already been published on Saturday by the Alaska State Trooperator.	Palin, 29, of Wasilla, Alaska, was arrested on charges of do- mestic violence. – Against him, a report of domestic vi- olence and resistance in ar- rest was already released Sat- urday, according to a report released Saturday by Alaska State Troopers.
 032 "Ich habe [] nicht versteckt Fords Behauptungen, ich habe ihre Geschichte nicht geleakt", erzählte Feinstein dem Komitee, berichtete The Hill. 031 Briefings werden immer noch stat- 	"I did not hide Dr. Ford's allegations, I did not leak her story," Feinstein told the committee, The Hill re- ported. Briefings will still happen,	"Thave [] not hidden Ford's claims that I have not lived their history," told Finestein the committee, reported The Hill. briefing is still going to take	"I did not hide [Forman's claims, I didn't geleast her story," Feinstein told the committee, The Hill reported. Briefings will still take place,
tfinden, sagte Sanders, aber sollte die Presse die Chance haben, dem Präsidenten der Vereinigten Staaten die Fragen direkt zu stellen, so sei das unendlich besser, als mit ihr zu sprechen.	Sanders said, but "if the press has the chance to ask the president of the United States questions directly, that's in- finitely better than talking to me.	bleen	Sanders said, but if the press has the chance to ask the pres- ident of the United States di- rectly, so that is unendlich better than talking to her.

Table 16: Top 10 segments by $|\delta_{BLEU}(i, h_S, h_U)|$ on DE-EN.

$\delta_{ ext{Macrof}_1}$	Source	Reference	SNMT	UNMT
.044	Il ne fallait qu'en déployer les ac- cidents, et l'affaire, jacobinisme oblige, était confiée aux préfets et aux sous-préfets, interprètes au- torisés.	All it took was to highlight its mistakes and, in keeping with Jacobinism, the issue would be entrusted to prefects and sub-prefects - the authorised interpreters.	It should deploy the accidents, and the case, Jacobinism obliges, was entrusted to the prefects and the sub-prefects, authorized interpreters.	It only took to deploy the ac- cidents, and the matter, jacobi- nite oblige, was handed to the préfets and the sous-préfets, authorized interprètes.
038	Les spécialistes disent que les personnes sont systématique- ment contraintes à faire leurs aveux, malgré un changement dans la loi qui a été voté plus tôt dans l'année interdisant aux autorités de forcer quiconque à s'incriminer lui-même.	Experts say confessions are still routinely coerced, despite a change in the law earlier this year banning the authori- ties from forcing anyone to in- criminate themselves.	The experts say that people are systematically forced to make their confessions, de- spite a change in the law which was passed earlier this year prohibiting the authori- ties to force anyone to incrim- inating himself.	Experts say people are rou- tinely forced to make their confessions, despite a change in the law that was passed ear- lier in the year banning offi- cials from forcing anyone to incriminate themselves.
.035	Ils sont intersexués, l'intersexualité faisant partie du groupe de la soixantaine de maladies diagnostiquées comme désordres du développement sexuel, un terme générique désignant les personnes possé- dant des chromosomes ou des gonades (ovaires ou testicules) atypiques ou des organes sexuels anormalement développés.	They are intersex, part of a group of about 60 conditions that fall under the diagno- sis of disorders of sexual de- velopment, an umbrella term for those with atypical chro- mosomes, gonads (ovaries or testes), or unusually deve- loped genitalia.	They are intersex, intersex forming part of the Group of 60 diseases diagnosed as dis- orders of sexual development, a generic term for people with chromosomes or atypical go- nads (ovaries or testes) or ab- normally developed sexual or- gans.	They are intersexuzed, with intersexuality making up the group of the soixantaine of diseases diagnosed as disor- dered sexual development, a generic term dissignant peo- ple possessing chromosomes or gonades (ovaries or testic- ules) atypiques or anormally developed sexual organs.
034		The violence is becoming more and more deadly in spite of reinforced security mea- sures and large-scale military operations undertaken in re- cent months by Nouri Al Ma- liki's government, which is dominated by Shiites.	Such violence are more lethal despite measures enhanced se- curity and large-scale military operations launched by the Government of Nouri Al Ma- liki, the Shia-dominated for months.	Those violence is increasingly deadly in the face of increased security measures and major military operations launched for months by Nouri Al Ma- liki's government, dominated by Shiites.
034	Du côté du gouvernement, on estime que 29 954 membres des forces armées du prési- dent Bachar el-Assad ont trouvé la mort, dont 18 678 étaient des combattants des forces pro- gouvernementales et 187 des mil- itants du Hezbollah libanais.	On the government side, it said 29,954 are members of President Bashar Assad's armed forces, 18,678 are pro-government fighters and 187 are Lebanese Hezbollah militants.	On the side of the Govern- ment, it is estimated that 29 954 members of the armed forces of president Bachar Al- Assad died, whose 18 678 were 187 Lebanese Hezbollah militants and fighters of the pro-Government forces.	On the government side, one estimate says 29,954 mem- bers of President Bachar al- Assad's armed forces have found their way, including 18,678 were fighters from pro- government forces and 187 from Lebanese Hezbollah mil- itants.
033	Mercredi, le Centre américain de contrôle et de prévention des mal- adies a publié une série de di- rectives indiquant comment gérer les allergies alimentaires des en- fants à l'école.	On Wednesday, the Centers for Disease Control and Pre- vention released a set of guide- lines to manage children's food allergies at school.	Wednesday, the US Centre of disease prevention and control issued a set of guidelines in- dicating how to manage food allergies of children at the school.	Wednesday, the U.S. Centers for Disease Control and Pre- vention issued a series of di- rectives indicating how to han- dle children's food allergies at school.
.033	N'est-il pas surprenant de lire dans les colonnes du Monde à quelques semaines d'intervalle d'une part la reproduction de la correspondance diplomatique américaine et d'autre part une condamnation des écoutes du Quai d'Orsay par la NSA ?	And is it not surprising to read in the pages of Le Monde, on the one hand, a reproduc- tion of diplomatic correspon- dence with the US and, on the other, condemnation of the NSA's spying on the Ministry of Foreign Affairs on the Quai d'Orsay, within a matter of weeks?	Is it not surprising to read in the columns of the world a few weeks apart on the one hand the reproduction of American diplomatic cor- respondence and on the other hand a condemnation of the Quai d'Orsay by the NSA lis- tens?	Isn't it surprising to read in the Times' pages just weeks apart of one side's reproduction of the American diplomatic cor- respondance and of another a condamnation of the Quai d' Orsay's écoutes by the NSA?
.032	Les ministères appellent à présent les personnes qui au- raient été mordues, griffées, égratignées, ou léchées sur une muqueuse ou sur une peau lésée par ce chaton ou dont l'animal aurait été en contact avec ce chaton entre le 8 et le 28 octobre à contacter le 08.11.00.06.95 entre 10 heures et 18 heures à partir du ler novembre.	The ministries are currently asking anyone who might have been bitten, clawed, scratched or licked on a mu- cous membrane or on dam- aged skin by the kitten, or who own an animal that may have been in contact with the kit- ten between 08 to 28 October, to contact them on 08 11 00 06 95 between 10am and 6pm from 01 November.	Departments now call peo- ple who have been bitten, scratched, scratched or licked on mucous membranes or skin injured by this kitten or where the animal would have been in contact with this kitten be- tween 8 and 28 October to contact the 08.11.00.06.95 be- tween 10 a.m. and 6 p.m. from November 1.	The at present call on people who may have been morbid, griffon, egregious or layed on a mug or on a skin léché by this chateau or whose an- imal may have been in con- tact with that chateau between 8 and 28 October to contact 08.11.00.095 between 10 and 18 November.
.030	A cette IIIe République, moment central et créateur, Pierre Nora a montré beaucoup d'intérêt et même de tendresse: saluant ceux qui se sont alors employés à ré- parer la fracture révolutionnaire, en enseignant aux écoliers tout ce qui dans l'ancienne France préparait obscurément la France moderne et en leur proposant une version unifiée de leur histoire.	Pierre Nora has shown has shown great interest and even tenderness for this Third Re- public: he salutes those who tried at the time to repair the divide created by the Rev- olution by teaching students about everything in the for- mer France that obscurely paved the way for the modern France, and by offering them a unified version of their his- tory.	This third Republic, while central and creator, Pierre Nora has shown great interest and even tenderness: saluting those who then worked to re- pair the revolutionary divide, by teaching students what in the former France preparing darkly modern France and of- fering them a version unified in their history.	At this IIIe République, central and creator moment, Pierre Nora showed much interest and even tendresse: praising those who then helped to repair the revolu- tionary fracture, teaching schoolchildren everything in the former French Republic that obscurantly prepared modern France and offering them a unifying version of their history.
.030	La théorie dominante sur la façon de traiter les enfants pourvus d'organes sexuels ambigus a été lancée par le Dr John Money, de l'université Johns-Hopkins, qui considérait que le genre est mal- léable.	The prevailing theory on how to treat children with ambigu- ous genitalia was put forward by Dr. John Money at Johns Hopkins University, who held that gender was malleable.	The prevailing theory about how to treat children with ambiguous sexual organs was launched by Dr. John Money of the Johns Hopkins Univer- sity, who considered that the genre is malleable. ROF ₁ (i, h_S, h_U) on FR-EN.	The dominant theory about how to treat children armed with ambigüous sex organs was launched by Dr. John Money, of Johns-Hopkins University, who considered the genre maudlin.

Table 17: Top 10 segments by $|\delta_{MACROF_1}(i, h_S, h_U)|$ on FR-EN. 1154

$\delta_{ m BLEU}$	Source	Reference	SNMT	UNMT
026	Mais le représentant Bill Shuster (R-Pa.), président du Comité des transports de la Chambre des représentants, a déclaré qu'il le considérait aussi comme l'alternative la plus viable à long terme.	But Rep. Bill Shuster (R-Pa.), chairman of the House Trans- portation Committee, has said he, too, sees it as the most vi- able long-term alternative.	But Congressman Bill Shus- ter (R - PA.), Chairman of the House of representa- tives Transportation Commit- tee, said that he considered as the most viable alternative in the long term.	But Rep. Bill Shuster (R-Pa.), chairman of the House Trans- portation Committee, said he also considered it the most vi- able long-term alternative.
.025	Les neuf premiers épisodes de Sheriff Callie's Wild West seront disponibles à partir du 24 novem- bre sur le site watchdisneyju- nior.com ou via son application pour téléphones et tablettes.	The first nine episodes of Sher- iff Callie's Wild West will be available from November 24 on the site watchdisneyju- nior.com or via its application for mobile phones and tablets.	The first nine episodes of Sher- iff Callie's Wild West will be available from November 24 on the site watchdisneyju- nior.com or via its application for phones and tablets.	Sheriff first nine episodes of Sheriff Callie's Wild West will be available as of Novem- ber 24 on the watchdisneyju- nior.com website or via its ap- plication for phones and com- puter
.024	Le président Xi Jinping, qui a pris ses fonctions en mars dernier, a fait de la lutte contre la cor- ruption une priorité nationale, es- timant que le phénomène consti- tuait une menace à l'existence- même du Parti communiste.	President Xi Jinping, who took office last March, has made the fight against corrup- tion a national priority, believ- ing that the phenomenon is a threat to the very existence of the Communist Party.	President Xi Jinping, who took office in March, has made the fight against corrup- tion a national priority, be- lieving that the phenomenon posed a threat to the very exis- tence of the Communist Party.	puters. President Xi Jinping, who took office in March, has made fighting corruption a na- tional priority, saying the phe- nomenon posed a threat to the Communist Party's existence- free existence.
.021	Un peu plus tôt, sur la route menant à Bunagana, poste- frontière avec l'Ouganda, des militaires aidés de civils chargeaient un lance-roquettes multiple monté sur un camion flambant neuf des FARDC, de- vant assurer la relève d'un autre engin pilonnant les positions du M23 sur les collines.	A little earlier, on the road to Bunagana, the frontier post with Uganda, soldiers assisted by civilians loaded up a mul- tiple rocket launcher mounted on a brand new truck belong- ing to the FARDC, intended to take over from another device pounding the positions of the M23 in the hills.	Earlier, on the road leading to Bunagana border post with Uganda, soldiers helped civil- ians loaded a multiple rocket launcher mounted on a truck brand new FARDC, to ensure succession of another engine pounding the positions of the M23 in the hills.	A day earlier, on the road leading to Bunagana, a post- code with Uganda, military personnel aided by civilians were loading a multiple rocket lance-roquettes fire on a flam- bant neuf FARDC truck, ex- pected to provide the lead for another device pilfering M23 positions on the hills.
.021	Il y a, avec la crémation, "une vi- olence faite au corps aimé", qui va être "réduit à un tas de cen- dres" en très peu de temps, et non après un processus de décom- position, qui "accompagnerait les phases du deuil".	With cremation, there is a sense of "violence committed against the body of a loved one", which will be "reduced to a pile of ashes" in a very short time instead of after a process of decomposition that "would accompany the stages of grief".	There, with the cremation, "a violence made to the beloved body", which will be "reduced to a pile of ashes" in a very short time, and not after a process of decomposition, which "would accompany the phases of mourning".	There is, with cremation, "vio- lence done to the loved one," which is going to be "reduced to a tas of ashes" in very lit- tle time, and not after a pro- cess of disablement, which would "accompany the phases of grief."
.021	Scott Brown, le capitaine du Celtic Glasgow, a vu son appel rejeté et sera bien suspendu pour les deux prochains matches de Ligue des champions de son club, contre l'Ajax et l'AC Milan.	Scott Brown, Glasgow Celtic captain, has had his appeal re- jected and will miss his club's next two Champion's League matches, against Ajax and AC Milan.	Scott Brown, the captain of the Glasgow Celtic, saw his appeal dismissed and will be well suspended for the next two matches of the champions League for his club against Ajax and AC Milan.	Scott Brown, the Celtic cap- tain, has had his appeal re- jected and will be well sus- pended for his club's next two Champions League matches, against Ajax and AC Milan.
021	Les irréductibles du M23, soit quelques centaines de combat- tants, étaient retranchés à près de 2000 mètres d'altitude sur les collines agricoles de Chanzu, Runyonyi et Mbuzi, proches de Bunagana et Jomba, deux local- ités situées à environ 80 km au nord de Goma, la capitale de la province du Nord-Kivu.	The diehards of the M23, who are several hundreds in number, had entrenched them- selves at an altitude of almost 2,000 metres in the farmland hills of Chanzu, Runyonyi and Mbuzi, close to Buna- gana and Jomba, two towns located around 80km north of Goma, the capital of North Kivu province.	The irreducible m23, or a few hundred fighters, were cut off to nearly 2000 metres above sea level on the hills agricul- tural Chanzu, Runyonyi and Mbuzi, near Bunagana and Jomba, located about 80 km north of Goma, the capital of the province of North Kivu.	The irréductibles M23, or some hundred fighters, were retranchés at nearly 2000 feet of altitude on the agricultural hills of Chanzu, Runyonyi and Mbuzi, close to Buna- gana and Jomba, two towns located about 80 miles north of Goma, the capital of North Kivu province.
021	Il a indiqué que le nouveau tri- bunal des médias « sera toujours partial car il s'agit d'un prolonge- ment du gouvernement » et que les restrictions relatives au con- tenu et à la publicité nuiraient à la place du Kenya dans l'économie mondiale.	He said the new media tri- bunal "will always be bi- ased because it's an exten- sion of the government," and that restrictions on content and advertising would damage Kenya's place in the global economy.	He said as the new media tri- bunal ' will be always par- tial because it is an exten- sion of the Government "and content and advertising restric- tions hurt instead of the Kenya into the world economy.	He said the new media tri- bunal "will always be partial because it is a extension of the government" and that re- strictions relating to content and advertising would hurt Kenya's place in the global economy.
.021	Dans "Les Fous de Benghazi", il avait été le premier à révéler l'existence d'un centre de com- mandement secret de la CIA dans cette ville, berceau de la révolte libyenne.	In "Les Fous de Benghazi", he was the first to reveal the ex- istence of a secret CIA com- mand centre in the city, the cradle of the Libyan revolt.	In "Les Fous de Benghazi", he was the first to reveal the ex- istence of a secret CIA com- mand center in this town, cra- dle of the Libyan revolt.	In "The Facts of Libya," he had been the first to reveal the existence of a secret CIA com- mand center in that city, the birthplace of the Libyan upris- ing.
020	un projet pilote de 90 M\$ Î année dernière qui aurait porté sur envi- ron 10 000 voitures.	The U.S. Senate approved a \$90-million pilot project last year that would have involved about 10,000 cars.	The US Senate has approved a pilot project of 90 MS last year which would have cov- ered about 10,000 cars.	The U.S. Senate approved a \$90 million pilot project last year that would have focused on about 10,000 cars.

Table 18: Top 10 segments by $|\delta_{BLEU}(i,h_S,h_U)|$ on FR-EN.

croF ₁ Source	Reference	SNMT	UNMT
.131 David Grimal are o cariera interna- tionala de violonist solo, de-a lun-	David Grimal has an international career as a solo violinist. Dur-	David Grimal has an international career as a solo violinist, along	David Grimal has an internation solo violinist, Peter Csaba, Heinric
gul careia a sustinut regulat con-	ing the last 20 years he held reg-	which claimed concerts regularly	Schiff, Lawrence Foster, Emmanu
certe în ultimii 20 de ani pe princi-	ular concerts on the main clas-	in the past 20 years in key scenes	Krivine, Mikhail Pletnev, Mikha
palele scene de muzica clasica ale	sical music stages of the world	of classical music of the world and	Fruyhbeck Orchestra of Burgos ar
lumii și cu orchestre prestigioase	with prestigious orchestras such as	with prestigious orchestras such	Peter Eotvos, Andris Philharmoniqu
cum ar fi Orchestre de Paris,	the Orchestre de Paris, Orchestre	as the Orchestre de Paris, Or-	de Radio France, Russian Nation
Orchestre Philharmonique de Ra-	Philharmonique de Radio France,	chestre Philharmonique de Radio	Orchestra, the French National C
dio France, Russian National Or-	Russian National Orchestra, Or-	France, Russian National Orches-	chestra of Lyon, the Berliner Syr
chestra, Orchestre National de Lyon, Chamber Orchestra of Eu-	chestre National de Lyon, Cham- ber Orchestra of Europe, Berliner	tra, Orchestre National de Lyon, the Chamber Orchestra of Eu-	phoniker, the New Japan Philharmon the Orchestre de l'Opera of Lyon, th
rope, Berliner Symphoniker, New	Symphoniker, New Japan Phil-	rope, the Berliner Symphoniker,	Mozarteum Orchestra of Lyon, th
Japan Philharmonic, Orchestre de	harmonic, Orchestre de l'Opera	the New Japan Philharmonic, Or-	Jerusalem Symphony Orchestra ar
l'Òpera de Lyon, Mozarteum Or-	de Lyon, Salzburg Mozarteum	chestre de l'Opera de Lyon The	Sinfonia of Paris, under the direction
chestra Salzburg, Jerusalem Sym-	Orchestra, Jerusalem Symphony	Salzburg Mozarteum Orchestra,	of conductor like Orchestra: the N
phony Orchestra și Sinfonia Varso-	Orchestra and Sinfonia Varsovia	the Jerusalem Symphony Orches-	tional Orchestra, the National Orche
via, sub conducerea unor diri-	under the direction of conduc- tors such as Christoph Eschen-	tra, and Sinfonia Varsovia, un-	tra, the French Chamber Orchestra, the
jori precum Christoph Eschen- bach, Michel Plasson, Michael	bach, Michel Plasson, Michael	der the direction of conductors such as Christoph Eschenbach,	Berliner Symphoniology Orchestra an Sinfonia Sinfonia and Sinfonia Si
Schnwandt, Peter Csaba, Hein-	Schnwandt, Peter Csaba, Hein-	Michel Plasson, Michael Schn-	fonia Sinfonia, the Orchestra of th
rich Schiff, Lawrence Foster, Em-	rich Schiff, Lawrence Foster, Em-	wandt, Peter Csaba, Heinrich	Paris Symphony Orchestra, the Russi
manuel Krivine, Mikhail Pletnev,	manuel Krivine, Mikhail Pletnev,	Schiff, Lawrence Foster, Em-	National Orchestra, the National O
Rafael Fruhbeck de Burgos și Pe-	Rafael Fruhbeck de Burgos and	manuel Krivine, Mikhail Pletnev,	chestra, the National Opera of Europ
ter Eotvos, Andris Nelsons, Chris-	Peter Eotvos, Andris Nelsons,	Rafael Fruhbeck de Burgos, and	the Berliner Symphoniker, the Berlin
tian Arming.	Christian Arming.	Peter Eotvos, Andris Nelsons,	Symphoniker Orchestra, the Orchest
		Christian Arming.	of Lyon, the Mozarteum and Sinfon
063 O parte importantă a magazinului	A good portion of the start	An important part of the store	Sinfonia Sinfonia.
.063 O parte importantă a magazinului este dedicată raionului uriaș	A good portion of the store is given over to the massive	An important part of the store is devoted to giant rayon of	A significant part of the store is devote to the huge frozen fried chicken count
de congelatoare, unde veti găsi	freezer section, where you'll find	freezers, where you'll find curry	to the huge frozen fried chicken count where you will find curry leaves, peac
frunze de curry, pepene și ghimbir	frozen curry leaves, bitter melon	leaves, melon and Ginger frozen	and frozen ghrelish, whole fish, fish
congelate, rațe întregi, pește,	and galangal, whole ducks, fish,	whole fish, ducks, beef bile and	blood and bone stock, pork carcass, t
sânge și bilă de vită, carcase de	beef blood and bile, pork cas-	blood, carcasses of pork, fish balls,	ditional fish carnouts, semi-cooked c
porc, chiftele de pește, cârnați	ings, fish balls, regional sausages,	sausages, traditional dishes and	nacies and many others.
tradiționali, semi-preparate și	commercially-prepared foods and	more.	
multe altele.	more.		
.062 Persoanele interesate vor putea să	Those interested will find out how	Persons interested will be able to learn how to achieve harmonious	People interested will be able to lea
afle cum să realizeze creații sculp- turale florale armonioase (ike-	to make harmonious sculptural flo- ral creations (ikebana), how to	sculptural floral design (ikebana),	how to complete armonious creati contrasting (grafic, drawing, paintin
bana), cum să surprindă sufle-	capture the soul of the surround-	how to capture the soul of the sur-	Creations, how to capture the hearts
tul elementelor înconjurătoare în	ing elements in plastic compo-	rounding elements in plastic com-	surrounding elements in artistic co
compoziții plastice folosind arta	sitions using the Japanese ink	positions using the Japanese art	positions using the Japanese art
japoneză a pictării în tuș (Sumie)	painting art (Sumie) or how to	of painting in India ink (Sumie)	painting in tus (Sumie) or how to e
sau cum să își exprime propria	express their own individuality	or how to express their own in-	press their own individualistic ide
individualitate și creativitate prin	and creativity through the mul-	dividuality and creativity through	tity through the plurivalent universe
universul plurivalent al artelor fru-	tifaceted universe of Fine Arts	the universe plurivalent of fine	beautiful arts (grafic, drawing, paintir
moase (grafic, desen, pictură), a declarat Sorin Mazilu, profesorul	(graphic, drawing, painting) said Sorin Mazilu, the teacher of these	arts (painting, drawing, graphics), said Sorin Mazilu, teacher of such	painting), said Mr. Mazilu, the professor of these courses.
acestor cursuri.	courses.	courses.	sor of these courses.
.052 Umorul îmbrățișărilor frecvente	The humour of Joey and Chan-	His humor frequently asked	Ummaline frequent imbratisations
ale lui Joey și Chandler, mo-	dler's frequent hugs, moments	îmbrățișărilor of Joey and Chan-	Joey and Chandler, the moments wh
mentele în care se uită la fotbal	watching football on the comfy	dler, the times when looking at	he looks at football in his comfy
în fotoliile comode și pasiunea lui	chairs, and Ross's pining for	football in comfortable armchairs	toland Ross's passion for Rachel can
Ross pentru Rachel au venit din	Rachel, came from the knowledge	and Ross's passion for Rachel	from knowing that men can assimila
faptul că se știe că bărbații se pot	that yes, men can all relate to this,	came from the fact that it is known	with the situation, even though off
asocia cu această situație, chiar dacă adesea se feresc să îsi ex-	even if they often hold back from fully exploring their feelings.	that men may be associated with this case, even if often keep out to	they are shy about fully exploring the feelings.
ploreze complet sentimentele.	fully exploring their reenings.	explore fully the sentiments.	reenings.
.052 În plus, zilele următoare vom face	In addition, in the coming days	In addition, the coming days we	In addition, next week we will ma
recepția lucrărilor de la pasarela	we will accept the works from	will make the reception of the	the receptia work on the pasarela lin
care leagă UPU de clinicile de	the bridge linking the ER to the	works from the footbridge link-	ing the UPU to the cardiology, hepat
cardiologie, hepatologie și gas-	cardiology, hepatology and gas-	ing the UPU of Cardiology clin-	ogy and gastroenterology clinicians
troenterologie pentru a asigura	troenterology clinics to ensure the	ics, Hepatology and gastroenterol-	ensure the transport in decent mann
transportul în condiții decente a	proper transfer of patients from	ogy to ensure decent conditions of	of the patient from the UPU to neig
bolnavului din UPU în clinicile	the ER to the neighbouring clinics	transportation of the patient from	bouring clinicians, the hospital's ch
vecine, a mai explicat managerul	explained the manager of the med-	the UPU in neighboring clinics,	executive explained.
unității medicale.	ical unit.	explained Manager medical unit.	While the Arab madia are 6
.052 Desi presa araba anunta ca despar- tirea lui Sanmartean de Ittihad e	Although Arab media announced that the separation between San-	Although the Arabic press an-	While the Arab media say Sa martean's departure from Ittihad is i
iminenta, impresarul antreorului	martean and Ittihad is imminent,	nounces that the breakup of his impending Sanmartean Ittihad e,	minent, impressions of coach Ladisl
Ladislau Boloni, Arcadie Zaporo-	coach Ladislau Boloni's agent, Ar-	impresario, Arcadie antreorului	Boloni's agent Arcadie Zaporojanu s
janu, spune ca acest lucru nu se va	cadia Zaporojanu, says this will no	Baghery Isaac says that this will	this will never happen.
mai intampla.	longer happen.	no longer happen.	
05 Cei mai putini sunt cei care au fi-	The least numerous are those who	Most are few people who have	Most are those who completed their
nalizat doctoratul - 0,67 - și tinerii	completed the doctorate courses	completed their Ph.d0.67-and	gree - 0.67 - and young people who f
care au terminat liceul ori scoala	- 0.67 - and young people who	young people who have completed	ished college or high school - 6.20 p
generala - 6,20%.	have finished high school or mid- dle school - 6.20%.	grade school-high school times 6,20%.	cent.
05 Politica clubului a facut din Vi-	The club's policy made Viitorul	The Club's policy has made the	The club's policy made the Viitorul
itorul o echipa de urmarit pentru	a team to be watched by foreign	team watched for foreign teams,	team to watch for overseas teams,
echipele din strainatate, astfel ca	teams, and so Hagi has sold this	such as Hagi has sold in this sum-	Hagi sold him for £1.5 million - Iar
Hagi a vandut în aceasta vara de	summer in the amount of 1.5 mil-	mer of 1.5 million euros-Ibarra	Hagi (Fiorentina - £1 million) and Jo
1,5 milioane de euro - Ianis Hagi	lion Euros - Ianis Hagi (Fiorentina	H (Fiorentina-EUR 1 million) and	Mitrita (Bari - £500 million).
(Fiorentina - 1 milion de euro) și	- 1 million euro) and Alexandru	Alexander Duane (Bari-500 thou-	
Alexandru Mitrita (Bari - 500 mii	Mitrita (Bari - 500 thousand euro).	sand euros).	
euro).	These instants of the literation	The sector should be a first first first	These included the second
.049 Au inclus amestecul de pungași	They included the medley of tax-	They included a mixture of tax rev-	They included the mixture of unpa
neplătitori de taxe, străini care	shy rascals, phone-hacking for-	enue, pungași foreign fraudau tele-	tax pungas, foreigners who telepho
fraudau telefonic și pseudo non-domi care dețin majoritatea	eigners and pseudo non-doms who own most of our great newspapers.	phone and non-pseudo domi who hold most major newspapers.	and pseudo non-doms who own mo of the key newspapers.
ziarelor importante.	e most of our great newspapers.	note most major newspapers.	er die key newspapers.
.048 Constantin Brâncuși se naște la 19	Constantin Brancusi was born on	Constantin Brâncuși February 19,	Mr. Brancusi was born Feb. 19, 187
	19 February 1876 in Hobita, a	was born in 1876, a small Certi-	in Hobson, a small village in the tov
februarie 1876, în Hobița, un mic	17 1 condary 1070 in 1100na, a		
	small village from Peştişani, Gorj	fied village in Peștișani commune,	of Peston, Pa., at the foot of the mou
februarie 1876, în Hobița, un mic		fied village in Peştişani commune, Gorj County, at the foot of the	of Peston, Pa., at the foot of the mot tains.

Table 19: Top 10 segments by $|\delta_{MACROF_1}(i, h_S, h_U)|$ on RO-EN.

$\delta_{ m BLEU}$ Source	Reference	SNMT	UNMT
.114 David Grimal are o cariera internationala de violonist solo, de-a lungul carcia a sustinut regulat concerte în ultimii 20 de ani pe principalele scene de muzica clasica ale lumii și cu orchestre prestigioase cum ar fi Orchestre de Radio France, Russian National Orchestra, Orchestre Philharmonique de Radio France, Russian National Orchestra, Orchestre National de Lyon, Chamber Orchestra of Europe, Berliner Symphoniker, New Japan Philharmonic, Orchestre de l'Opera de Lyon, Mozarteum Orchestra Salzburg, Jerusalem Symphony Orchestra și Sinfonia Varsovia, sub conducerea unor dirijori precum Christoph Eschenbach, Michel Plasson, Michael Schwandt, Peter Csaba, Heinrich Schiff, Lawrence Foster, Emmanuel Krivine, Mikhail Pletnev, Rafael Fruhbeck de Burgos și Peter Eotvos, Andris Nelsons, Christian Arming.	David Grimal has an international ca- reer as a solo violinist. During the last 20 years he held regular con- certs on the main classical music stages of the world with prestigious orchestras such as the Orchestre de Paris, Orchestre Philharmonique de Radio France, Russian National Orchestra, Orchestre National de Lyon, Chamber Orchestra of Europe, Berliner Symphoniker, New Japan Philharmonic, Orchestre de l'Opera de Lyon, Salzburg Mozarteum Or- chestra, Jerusalem Symphony Or- chestra and Sinfonia Varsovia un- der the direction of conductors such as Christoph Eschenbach, Michel Plasson, Michael Schnwandt, Peter Csaba, Heinrich Schiff, Lawrence Foster, Emmanuel Krivine, Mikhail Pletnev, Rafael Fruhbeck de Burgos and Peter Eotvos, Andris Nelsons, Christian Arming.	David Grimal has an international ca- reer as a solo violinist, along which claimed concerts regularly in the past 20 years in key scenes of classical music of the world and with pres- tigious orchestras such as the Or- chestre de Paris, Orchestre Philhar- monique de Radio France, Russian National Orchestra, Orchestre Na- tional de Lyon, the Chamber Or- chestra of Europe, the Berliner Sym- phoniker, the New Japan Philhar- monic, Orchestre de l'Opera de Lyon The Salzburg Mozarteum Orches- tra, the Jerusalem Symphony Or- chestra, and Sinfonia Varsovia, un- der the direction of conductors such as Christoph Eschenbach, Michel Plasson, Michael Schnwandt, Peter Csaba, Heinrich Schiff, Lawrence Foster, Emmanuel Krivine, Mikhail Pletnev, Rafael Fruhbeck de Burgos, and Peter Eotvos, Andris Nelsons, Christian Arming.	David Grimal has an international solo violinist, Peter Csaba, Heinrich Schiff, Lawrence Foster, Emmanuel Krivine, Mikhail Pletnev, Mikhail Fruyhbeck Or- chestra of Burgos and Peter Eotvos, An- dris Philharmonique de Radio France, Russian National Orchestra, the French National Orchestra of Lyon, the Berliner Symphoniker, the New Japan Philhar- monic, the Orchestre de l'Opera of Lyon, the Mozarteum Orchestra of Lyon, the Jerusalem Symphony Orchestra and Sinfo- nia of Paris, under the direction of conduc- tor like Orchestra: the National Orchestra, the National Orchestra, the French Cham- ber Orchestra; the Berliner Symphoniol- ogy Orchestra and Sinfonia Sinfonia and Sinfonia Sinfonia Sinfonia is the Orchestra of the Paris Symphony Orchestra, the Rus- sian National Orchestra, the National Or- chestra, the National Opera of Europe, the Berliner Symphoniker, the Berliner Sym- phoniker Orchestra, the Orchestra of Lyon, the Mozarteum and Sinfonia Sinfonia Sin- fonia.
.089 Editia de toamna va avea loc la Chisinau (Casa Armatei, 2-3 oc- tombrie), Iasi (Palas, 6-7 octombrie), Brasov (Casa Armatei, 14-15 oc- tombrie), Cluj Global (Sala Poli- valenta, 20-21 octombrie), Cluj IT (Sala Polivalenta, 22-23 octombrie), Sibiu (Centrul de Afaceri, 28-29 oc- tombrie) și Targu-Mures (Teatrul Na- tional, 5 noiembrie).	The autumn edition will take place in Chisinau (Army House 2-3 Octo- ber), Iasi (Palas 6-7 October), Brasov (Army House 14-15 October), Cluj Global (Polyvalent Hall 20-21 Octo- ber) Cluj IT (Polyvalent Hall 22-23 October), Sibiu (Business Centre 28- 29 October) and Targu-Mures (Na- tional Theatre, November 5).	Autumn edition will take place in Chisinau (Army House, 2-3 Oc- tober), Iasi (Palas, 6-7 October), Brasov (Army House, 14-16 Octo- ber), Global (Sala Polivalenta, 20-21 October), Cluj (Sala Polivalenta, 22- 23 October), Sibiu (Business Cen- tre, 28-29 October) and Targu-Mures (National Theatre, November 5).	The fall season will be in Washington (House, 2-3), Washington (Palas, Oct. 6- 7), Brasov (House, Oct. 14-15), San Fran- cisco Global (East Coast, Oct. 20-21), San Francisco IT (East Coast, Oct. 22- 23), Sibiu (Center for Business, Oct. 28- 29) and Targu-Mures (the National Center, Nov. 5).
072 Realizatorii studiului mai transmit ca "romanii simt nevoie de ceva mai multa aventura în viata lor (24%), urmat de afectiune (21%), bani (21%), siguranta (20%), nou (19%), sex (19%), respect 18%, in- credere 17%, placere 17%, conectare 17%, cunoastere 16%, protectie 14%, importanta 14%, invatare 12%, liber- tate 11%, autocunoastere 10% și con- trol 7%".	The study's conductors transmit that "Romanians feel the need for a little more adventure in their lives (24%), followed by affection (21%), money (21%), safety (20%), new things (19%), sex (19%) respect 18%, con- fidence 17%, pleasure 17%, connec- tion 17%, knowledge 16%, protec- tion 14%, importance 14%, learning 12%, freedom 11%, self-awareness 10% and control 7% ".	The filmmakers may study trans- mitted as "Romanians feel in need of something more adventure in their lives (24%), followed by affec- tion (21%), money (21%), security (20%), new (19%), sex (19%), re- spect, trust and 18% 17% 17% plea- sure, 17%, 16%, knowledge protec- tion 14%, 14%, 12%, liberty learning 11%, self-awareness and control 10% 7%".	Reporters also deliver that "Americans feel need for some more adventure in their lives (24%), followed by affection (21%), money (21%), safety (20%), new (19%), sex (19%), respect 18%, confidence 17%, pleasure 17%, connection 17%, knowl- edge 14%, protection 14%, importance 14%, learning 12%, freedom 11%, auto- cunoastere 10% and control 7%."
045 Fed ar trebui să pună problema stabilității financiare pe primul loc doar în cazul unei crize majore, cum a fost cutremurul de pe piață din 2008, a declarat Adam S. Posen, fost membru al comisiei de stabilire a ratei dobânzii din cadrul Bank of England.	The Fed should put financial stabil- ity concerns first only during a major crisis, such as the 2008 market melt- down, said Adam S. Posen, a former member of the Bank of England's rate-setting committee.	The Fed should put the issue of finan- cial stability in the first place only in the event of major crises such as the earthquake in the market since 2008, "said Adam s. Posen, a former mem- ber of the Commission for determin- ing the interest rate within the Bank of England.	The Fed should put financial stability first only in a major crisis, such as the 2008 credit crunch, said Adam S. Posen, a for- mer member of the Bank of England's sta- bilisation committee.
041 Atunci când susținătorilor lui Clinton li se adresează o întrebare deschisă referitor la motivul pentru care ar dori ca ea să câştige cursa, răspunsul majoritar este că deține experiența adecvată (16%), urmat de faptul că a venit momentul ca o femeie să fie președinte (13%) și că este cel mai bun candidat pentru această funcție (10%).	When Clinton's supporters are asked in an open-ended question why they want her to be the nominee, the top answer is that she has the right expe- rience (16 percent), followed by it's time for a woman president (13 per- cent), and that she is the best candi- date for the job (10 percent).	When Clinton supporters are asked to answer an open question regarding why you want her to win the race, the answer is that majority holds appro- priate experience (16%), followed by the fact that the time has come for a woman to be President (13%), and that is the best candidate for this po- sition (10%).	When Clinton supporters are asking an open question as to why she would like to win the race, the majoritar answer is that she holds the right experience (16 percent), followed by the fact that it has come time for a woman to be president (13 percent) and that she is the best candidate for the job (10 percent).
 040 Clinton este acum susținută de 47% din alegătorii democrați (în scădere de la 58%), în timp ce Sanders este pe locul doi, cu 27% (în urcare față de 17%). 	Clinton now has the backing of 47 percent of Democratic primary vot- ers (down from 58 percent), while Sanders comes in second, with 27 percent (up from 17 percent).	Clinton is now supported by 47% of the voters Democrats (down 62%), while Sanders is in second place with 27% (in relation to climb 17%).	Clinton is now backed by 47 percent of Democratic voters (down from 58 per- cent), while Sanders is second with 27 per- cent (up from 17 percent).
.039 Persoanele interesate vor putea să afle cum să realizeze creații sculpturale florale armonioase (ikebana), cum să surprindă sufle- tul elementelor înconjurătoare în compoziții plastice folosind arta japoneză a pictării în tuş (Sumie) sau cum să îşi exprime propria individualitate şi creativitate prin universul plurivalent al artelor fru- moase (grafic, desen, pictură), a declarat Sorin Mazilu, profesorul acestor cursuri.	Those interested will find out how to make harmonious sculptural flo- ral creations (ikebana), how to cap- ture the soul of the surrounding ele- ments in plastic compositions using the Japanese ink painting art (Sumie) or how to express their own indi- viduality and creativity through the multifaceted universe of Fine Arts (graphic, drawing, painting) said Sorin Mazilu, the teacher of these courses.	Persons interested will be able to learn how to achieve harmonious sculptural floral design (ikebana), how to capture the soul of the sur- rounding elements in plastic com- positions using the Japanese art of painting in India ink (Sumie) or how to express their own individu- ality and creativity through the uni- verse plurivalent of fine arts (paint- ing, drawing, graphics), said Sorin Mazilu, teacher of such courses.	People interested will be able to learn how to complete armonious creative contrast- ing (grafic, drawing, painting) Creations, how to capture the hearts of surrounding elements in artistic compositions using the Japanese art of painting in tus (Sumie) or how to express their own individualistic identity through the plurivalent universe of beautiful arts (grafic, drawing, painting, painting), said Mr. Mazilu, the professor of these courses.
.036 Companiile au vacante cateva mii de locuri de munca, oportunitati de internship și stagii de prac- tica, iar o parte dintre ele sunt deja anuntate pe site-ul oficial www.targuldecariere.ro.	Companies have thousands of vacant jobs, internship opportunities and in- ternships, and some of them are al- ready posted on the official website www.targuldecariere.ro.	Companies have thousands of vacant jobs, internship opportunities, and in- ternships, and some of them are al- ready announced on the official web- site www.targuldecariere.ro.	Companies have filled several thou- sand jobs, internships and training trips, and a number of them are al- ready announced on the official website www.targuldecarier.com.
035 Antrenorul celor de la Brisbane Broncos, Wayne Bennett, a făcut o uşoară referire la Storm după victo- ria echipei sale în meciul de califi- care cu North Queensland Cowboys, jucat sâmbătă seară, când a numit meciul respectiv o "demonstrație" a ligii de rugby și a declarat că cele două echipe din Queensland nu se prea pricep la wrestling.	Brisbane Broncos coach Wayne Ben- nett made a thinly veiled reference to the Storm after his side's qualify- ing final win over North Queensland Cowboys on Saturday night when he called that game a "showcase" of the rugby league and said the two Queensland weren't "too big" into wrestling.	The coach of the Brisbane Broncos, Wayne Bennett, made a slight refer- ence to the Storm after his team's victory in the qualifying match with North Queensland Cowboys, played Saturday night, when he called the match a "demonstration" of rugby league and said that the two teams from Queensland is not too good at wrestling.	Brisbane Broncos coach Wayne Bennett made a slight reference to the Storm af- ter their team's victory in the qualifying match against the North Queensland Cow- boys on Saturday night when he called the game a "demonstration" of the rugby league league league and said the two Queensland sides were not really pricep to wrestling.
.034 David Grimal recunoaste ca Les Dis- sonaces reprezinta "un lux", pre- cizand ca niciunul dintre muzicienii care fac parte din orchestra nu de-	David Grimal admits that Les Dis- sonaces is "a luxury", adding that none of the musicians who are part of the orchestra does not depend on	David Grimal admits that Les Dis- sonaces represents "a luxury", stat- ing that none of the musicians who are part of the orchestra does not de-	David Grimal admits Les Dissonaces rep- resent "a luxury," saying none of the musi- cians who make up the orchestra depinde of the success of this project to live.