# Can informal genres be better translated by tuning on automatic semantic metrics?

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### **Abstract**

Even though the informal language of spoken text and web forum genres presents great difficulties for automatic semantic role labeling, we show that surprisingly, tuning statistical machine translation against the SRL-based objective function, MEANT, nevertheless leads more robustly to adequate translations of these informal genres than tuning against BLEU or TER. The accuracy of automatic semantic parsing has been shown to degrade significantly on informal genres such as speech or tweets, compared to formal genres like newswire. In spite of this, human evaluators preferred translations from MEANTtuned systems over the BLEU- or TERtuned ones by a significant margin. Error analysis indicates that one of the major sources of errors in automatic shallow semantic parsing of informal genres is failure to identify the semantic frame for copula or existential senses of "be". We show that MEANT's correlation with human adequacy judgment on informal text is improved by reconstructing the missing semantic frames for "be". Our tuning approach is independent of the translation model architecture, so any SMT model can potentially benefit from the semantic knowledge incorporated through our approach.

## 1 Introduction

We present the first known results of robustly improving the adequacy of statistical machine translation (SMT) for *informal* genres by tuning against the semantic frame based objective function MEANT. Using both spoken language and

web forum datasets, we show that tuning against MEANT outperforms two SMT baselines that follow the common practice of tuning against the ngram based BLEU metric or the edit distance based TER metric. We investigate results across a battery of automatic evaluation metrics, as well as subjective human evaluation of translation adequacy. Since automatic semantic parsing has been shown to fare worse on informal genres, where the robustness of the POS tagging and syntactic parsing that it depends on suffers, it is surprising that tuning against a semantic frame based objective function nonetheless performs more robustly than tuning against non-semantic metrics. Our results encouragingly suggest that further improvement of semantic frame based objective functions for training SMT will be a fruitful direction for raising the utility of machine translation on informal language, and not only formal text genres.

Previous work on improving machine translation of informal text has mostly focused on using domain adaptation techniques instead of incorporating semantics, because the accuracy of automatic shallow semantic parsers has been reported to drop by around 10% on speech data (Favre et al., 2010) and by more than 30% on web data like tweets (Liu et al., 2010). Yet something more must be done; common errors in machine translation of formal text caused by semantic role confusions of the kind that plague state-of-the-art MT systems are even more glaring for informal texts. Semantic role confusion errors in SMT are mainly the consequence of using fast and cheap lexical n-gram based objective functions such as BLEU to drive development. Despite enforcing fluency, it has been established that these metrics do not enforce translation utility adequately and often fail to preserve meaning closely (Callison-Burch et al., 2006; Koehn and Monz, 2006).

We recently showed that the translation adequacy for formal news is improved by replacing the surface oriented metrics like BLEU or TER with a semantic objective function, when tuning the parameters of MT systems (Lo et al., 2013). We now ask whether the same approach of tuning MT systems against a semantic objective function might improve translation adequacy even for informal texts, despite the fact that automatic semantic parsing is known to be of lower accuracy. If the objective function for tuning SMT sufficiently reflects preservation of meaning, this should ultimately drive continuing progress toward higher utility. Our fully-automatic, semantic MT evaluation metric, MEANT (Lo et al., 2012), measures similarity between MT output and reference translations as judged via semantic frames. For formal text genres, we have shown that MEANT correlates more highly with human adequacy judgment than all other automatic MT evaluation metrics. Although MEANT has not been shown to work well for informal language genres where it is likely to suffer from degraded semantic parsing accuracy, since a high MEANT score is also contingent on correct lexical choices as well as syntactic and semantic structures, we hypothesize that perhaps tuning against MEANT would nonetheless improve both translation adequacy and fluency even on speech and web forum data.

The proposed approach of incorporating semantic structures into SMT by tuning the model against a semantic frame based evaluation metric is independent of assumptions about the underlying translation model architecture. Therefore, MT systems from different SMT approaches (such as hierarchical, phrase based, or synchronous/transduction grammar based) or those applying other techniques for informal data (such as domain adaptation from formal to informal text, or integration of linguistic features) could also benefit from the semantic information incorporated through our approach.

## 2 Related work

Relatively little work has been done toward addressing the problem of biasing the translation decisions of an SMT system to produce adequate translations for informal text that correctly preserve who did what to whom, when, where and why (Pradhan et al., 2004). There has been a recent surge of work aimed at incorporating seman-

tics into the SMT pipeline; however, none attempts to improve translation quality on *informal* text due to the difficulty of semantic parsing. Below, we describe some of the attempts to (a) improve informal text translation quality using domain adaptation techniques and (b) incorporate semantic role labeling information into the SMT pipeline and present a brief survey of evaluation metrics that focus on rewarding semantically valid translations.

## 2.1 Semantics in SMT

There is a rising trend of work aimed at incorporating semantics into various stages of the SMT pipeline, for example, preprocessing the input (Komachi *et al.*, 2006; Wu *et al.*, 2011), training treeto-string MT models (Liu and Gildea, 2010; Aziz *et al.*, 2011), training reordering model (Xiong *et al.*, 2012) and reordering the output in the postprocessing stage (Wu and Fung, 2009). All these approaches are orthogonal to the present question of whether to train toward a semantic objective function. In fact, any of the above models could potentially benefit from the proposed approach.

## 2.2 Adapting SMT for formal genres to informal genres

The major challenges for machine translation in the informal genres are (1) the data demonstrates a large variety of grammar issues, such as disfluencies, incomplete sentences and misspellings; and (2) only small volumes of high-quality parallel training data are available (Mei and Kirchhoff, 2010). Rao et al. (2007) and Wang et al. (2010) proposed to remove disfluency in preprocessing stage; Bertoldi et al. (2010) introduced a model to recover the misspelled words before translation; Banerjee et al. (2011) addressed the data sparsity problem by mixing data from comparable domain into the training of both the translation model and the language model whereas He and Deng (2011) proposed to classify the training data into in-domain or out-of-domain for training two independent translation model and then combine the two models using a system combination approach. Mei and Kirchhoff (2010) incorporated document-level semantics, such as topic of the discourse, through contextual modelling. Again, all these approaches are orthogonal to our approach of incorporating semantics into SMT by tuning against a semantic objective function and any of the above models could potentially benefit from tuning with semantic metrics.

### 2.3 MT evaluation metrics

Lo et al. (2013) showed that tuning against BLEU (Papineni et al., 2002) or TER (Snover et al., 2006) does not sufficiently drive SMT into making decisions to produce adequate translations. Other similar n-gram based or edit distance based metrics, such as NIST (Doddington, 2002), METEOR (Banerjee and Lavie, 2005), CDER (Leusch et al., 2006) and WER (Nießen et al., 2000) also suffer from the same problem of failing to adequately reflect translation utility and correctly bias translation model for producing adequate translation.

On the other hand, no work has been done towards tuning SMT systems against more linguistically motivated MT evaluation metrics because of expensive run time costs. ULC (Giménez and Màrquez, 2007, 2008) is an aggregated metric that incorporates a large set of linguistic features, including several semantic features and shows improved correlation with human judgement of translation quality (Callison-Burch et al., 2007; Giménez and Màrquez, 2007; Callison-Burch et al., 2008; Giménez and Màrquez, 2008). Lambert et al. (2006) tuned on QUEEN, a simplified version of ULC, that discards the semantic features of ULC and bases on pure lexical similarity. Although tuning on QUEEN produced slightly more preferable translations than solely tuning on BLEU, the metric does not make use of any semantic features and thus fails to exploit any potential gains from tuning to a semantic objective function. TINE (Rios et al., 2011) is a recall-oriented metric which aims to preserve the basic event structure but only performs comparably to BLEU and worse than METEOR on correlation with human adequacy judgment.

In contrast, Lo et al. (2013) show that a MT system tuned against MEANT (Lo et al., 2012) produces more adequate translations in formal news genre as evaluated both quantitatively and qualitatively. Precisely, MEANT is computed as follows:

- 1. Apply an automatic shallow semantic parser on both the references and MT output.
- Apply maximum weighted bipartite matching algorithm to align the semantic frames between the references and MT output by the lexical similarity of the predicates.
- 3. For each pair of aligned semantic frames,

- (a) Compute the lexical/phrasal similarity scores to determine the similarity of the semantic role fillers
- (b) Apply maximum weighted bipartite matching algorithm to align the semantic role fillers between the reference and MT output according to their lexical/phrasal similarity.
- 4. Compute the weighted f-score over the matching role labels of these aligned predicates and role fillers.

$$\begin{array}{rcl} M_{i,j} & \equiv & \operatorname{total} \# \operatorname{ARG} \operatorname{j} \operatorname{of aligned frame} \operatorname{i} \operatorname{in} \operatorname{MT} \\ R_{i,j} & \equiv & \operatorname{total} \# \operatorname{ARG} \operatorname{j} \operatorname{of aligned frame} \operatorname{i} \operatorname{in} \operatorname{REF} \\ S_{i,\operatorname{pred}} & \equiv & \operatorname{similarity} \operatorname{of predicate} \operatorname{in} \operatorname{aligned} \operatorname{frame} \operatorname{i} \\ S_{i,j} & \equiv & \operatorname{similarity} \operatorname{of} \operatorname{ARG} \operatorname{j} \operatorname{in} \operatorname{aligned} \operatorname{frame} \operatorname{i} \\ w_{\operatorname{pred}} & \equiv & \operatorname{weight} \operatorname{of} \operatorname{similarity} \operatorname{of} \operatorname{ARG} \operatorname{j} \\ w_{\operatorname{pred}} & \equiv & \operatorname{weight} \operatorname{of} \operatorname{similarity} \operatorname{of} \operatorname{ARG} \operatorname{j} \\ m_i & \equiv & \operatorname{\#tokens} \operatorname{filled} \operatorname{in} \operatorname{aligned} \operatorname{frame} \operatorname{i} \operatorname{of} \operatorname{MT} \\ r_i & \equiv & \operatorname{\#tokens} \operatorname{filled} \operatorname{in} \operatorname{aligned} \operatorname{frame} \operatorname{i} \operatorname{of} \operatorname{REF} \\ \operatorname{total} \# \operatorname{tokens} \operatorname{in} \operatorname{MT} \\ & \operatorname{total} \# \operatorname{tokens} \operatorname{in} \operatorname{REF} \\ & \operatorname{total} \# \operatorname{tokens} \operatorname{in} \operatorname{REF} \\ & \operatorname{precision} & = & \frac{\sum_i m_i \frac{w_{\operatorname{pred}} S_{i,\operatorname{pred}} + \sum_j w_j S_{i,j}}{w_{\operatorname{pred}} + \sum_j w_j M_{i,j}}}{\sum_i m_i} \\ & \operatorname{precall} & = & \frac{\sum_i r_i \frac{w_{\operatorname{pred}} S_{i,\operatorname{pred}} + \sum_j w_j S_{i,j}}{w_{\operatorname{pred}} + \sum_j w_j S_{i,j}}}{\sum_i r_i} \\ & \\ & \operatorname{recall} & = & \frac{\sum_i r_i \frac{w_{\operatorname{pred}} S_{i,\operatorname{pred}} + \sum_j w_j S_{i,j}}{w_{\operatorname{pred}} + \sum_j w_j R_{i,j}}}}{\sum_i r_i} \\ & \\ & \end{array}$$

where  $m_i$  and  $r_i$  are the weights for frame i that estimate the degree of contribution of the frame to the overall meaning of the sentence in the MT/REF respectively.  $M_{i,j}$  and  $R_{i,j}$  are the total counts of argument of type j in frame i in the MT and REF respectively.  $S_{i,\text{pred}}$  and  $S_{i,j}$  are the lexical/phrasal similarities of the predicates and role fillers of the arguments of type j between the MT and REF.  $w_{\text{pred}}$  and  $w_j$  are the weights of the the predicates and role fillers of the arguments of type j in the MT and REF. There are 12 weights for the set of semantic role labels in MEANT as defined in Lo and Wu (2011b) and they are determined by optimizing the correlation with human adequacy judgments using grid search (Lo and Wu, 2011a).

## 3 Tuning SMT against MEANT

We now show that using MEANT as an objective function to drive minimum error rate training (MERT) of state-of-the-art MT systems improves MT utility in the informal genres.

Aiming at improving SMT adequacy of informal genres, we set up two experiments on public speech TED talk data and web forum data. The TED talk MT system is trained on the IWSLT2012 Chinese-English parallel TED talk training data consisting of over 130k sentences pairs. The development and

Table 1: Translation quality of MT system tuned against MEANT, BLEU and TER on TED talk data

| TED talk    | BLEU ↑ | NIST ↑ | METEOR no_syn ↑ | METEOR ↑ | WER ↓ | CDER ↓ | TER ↓ | MEANT↑ |
|-------------|--------|--------|-----------------|----------|-------|--------|-------|--------|
| BLEU-tuned  | 12.09  | 4.36   | 38.14           | 41.28    | 83.87 | 68.55  | 80.83 | 22.70  |
| TER-tuned   | 9.63   | 3.67   | 32.75           | 35.19    | 74.00 | 59.24  | 72.31 | 20.41  |
| MEANT-tuned | 11.24  | 4.22   | 38.57           | 41.96    | 80.97 | 66.21  | 78.10 | 22.74  |

Table 2: Translation quality of MT system tuned against MEANT, BLEU and TER on web forum data

| forum       | BLEU ↑ | NIST ↑ | METEOR no_syn ↑ | METEOR ↑ | WER↓  | CDER ↓ | TER ↓ | MEANT ↑ |
|-------------|--------|--------|-----------------|----------|-------|--------|-------|---------|
| BLEU-tuned  | 9.58   | 4.10   | 31.77           | 34.63    | 80.09 | 64.54  | 76.12 | 17.11   |
| TER-tuned   | 6.94   | 2.21   | 28.55           | 30.85    | 76.15 | 57.96  | 74.73 | 15.39   |
| MEANT-tuned | 7.92   | 3.11   | 30.40           | 33.08    | 77.32 | 61.01  | 74.64 | 17.27   |

test sets of the TED talk data consist of 934 and 1664 sentences respectively with one reference. The web forum MT system is trained on a large collection of Chinese-English LDC newswire data with a small portion of web forum data released under DARPA GALE and BOLT projects. The development and test sets were a held-out subset of the DARPA BOLT phase 1 web forum data which consist of 2,000 sentences and 1,697 sentences respectively with one reference. We experimented two different MT systems: a Moses phrase based MT system trained on the TED talk data and a Moses hierarchical MT system trained on the BOLT web forum data. The TED talk Moses phrase based system leverage a 6-gram language model trained on the English side of the parallel training data while the web forum Moses hierarchical system uses a 5gram language model trained on a large volume of mixed newswire and web forum data.

We use ZMERT (Zaidan, 2009) to tune the MT system because it is a highly competitive, robust, and reliable implementation of MERT that is also fully configurable and extensible for incorporating new evaluation metrics. In this experiment, we use a MEANT implementation along the lines described in Lo *et al.* (2012) and Tumuluru *et al.* (2012) but we incorporate a variant of the aggregation function proposed in Mihalcea *et al.* (2006) for phrasal similarity of role fillers because it normalizes the phrase length better than geometric mean.

## 4 Results

Of course, tuning against any metric would maximize the performance of the SMT system on that particular metric; it would be overfitting. In the following, we avoid comparing on metrics too similar to the one that the system was tuned on. This is because Cer *et al.* (2010) showed that tuning on

METEOR, TER and their variations would do well on metrics similar to what they were tuned on but perform particularly poorly on the other metrics. Therefore, it is less meaningful to evaluate a system on metrics similar to what they were tuned on.

A far more worthwhile goal would be to bias the SMT system to produce adequate translations while achieving the best scores across all the metrics. In addition, we believe a good translation is one from which the reader could successfully understand at least the meaning of the source sentence, instead of just being fluent in the target language. With these as our objectives, we present the results of comparing MEANT-tuned systems against the baseline as evaluated on commonly used automatic metrics and human judgement of adequacy.

## 4.1 Cross-evaluation using automatic metrics

Tables 1 and 2 show that MEANT-tuned systems achieve the best scores across all other metrics on both TED talk and web forum data, when avoiding the comparison on metrics too similar to the one that the system was tuned on (the cells shaded in grey in the table). METEOR is grouped with BLEU and NIST because they are all n-gram based MT evaluation metrics. Our results indicate that MEANT-tuned system maintains a balance between lexical choices and word order as it performs well on n-gram based metrics that reward correct lexical choices and edit distance metrics that penalize incorrect word order. This is not surprising as a high MEANT score relies on a high degree of semantic structure matching, which is contingent upon correct lexical choices as well as syntactic and semantic structures.

Table 3: No. of sentences ranked the most adequate by human evaluators for each system in the web forum experiment.

|                 | Eval 1 | Eval 2 |
|-----------------|--------|--------|
| BLEU-tuned (B)  | 47     | 42     |
| TER-tuned (T)   | 28     | 23     |
| MEANT-tuned (M) | 59     | 68     |
| B=T             | 0      | 0      |
| M=B             | 8      | 9      |
| M=T             | 4      | 4      |
| M=B=T           | 4      | 4      |

## 4.2 Human subjective evaluation

In line with our original objective of biasing SMT systems toward producing adequate translations, we conduct a manual evaluation to judge the translation utility of outputs produced by MEANT-, BLEU- and TER-tuned systems in the web forum experiment. Following the manual evaluation protocol of Lambert *et al.* (2006), we randomly sampled 150 sentences from the MT output of the three systems in web forum domain. The output of our system and the two baselines along with the input sentence and the reference translation was presented to human evaluators. Two evaluators were instructed to choose the most adequate translation from the three MT output. The inter-evaluator agreement was 70%.

Table 3 indicates that output of the MEANT-tuned system is ranked adequate more frequently compared to BLEU- and TER-tuned baselines. We performed the right-tailed two proportion significance test on human evaluation of the SMT system outputs for both the genres. The MEANT-tuned system generates more adequate translations on web forum data than the TER-tuned system at the 99% significance level. The MEANT-tuned system is ranked more adequate than the BLEU-tuned system at the 95% significance level. The high inter-evaluator agreement and the significance tests confirm that MEANT-tuned system is better at producing adequate translations compared to BLEU or TER-tuned systems.

Although one might expect an SRL dependent metric such as MEANT to perform poorly on the domain of informal text, it nonetheless significantly outperforms the baselines at the task of generating adequate output. This indicates that the design of the MEANT evaluation metric is robust enough to tune an SMT system toward adequate

output on informal text domains despite the short-comings of automatic shallow semantic parsing.

## 5 Error analysis

To better understand why the MEANT-tuned MT system still able to outperform the BLEU- and the TER-tuned system despite reports pointing out the deficiency of automatic semantic parser on informal text data, we conduct a thorough error analysis. We specifically look into cases when the automatic shallow semantic parser fails to construct a parse from the sentence. Besides analysing the development and the test set of the two previous mentioned experiments, we also look for the same phenomenon in the MetricsMaTr 2008 broadcast news data set to ensure a more reliable result from the error analysis.

Table 4 shows that over 14% of the sentences in the TED talk demonstrate no semantic parse and on average over 8% of the sentences in the web forum data set has no semantic parse. The failure of the automatic semantic parser to provide any parse for MEANT to score the sentence would result in a zero MEANT score on those sentences. We further investigate into the cases when the automatic semantic parser fails to identify any semantic frames in the sentences so as to understand how to incorporate semantic information in those cases.

## 5.1 Failure to label the "be" semantic frames

Surprisingly, Table 5 shows that the failure of automatic semantic parsing on identifying semantic frames did not result from the ungrammatical sentences in the informal data. Instead, the major source of errors in automatic shallow semantic parsing of informal genres is failure to identify the semantic frame for copula or existential senses of "be" in perfectly grammatical sentences. The following is the Propbank (Palmer *et al.*, 2005) framesets definition of the predicate "be":

Roleset be.01: copula

Roles: ARG1-topic, ARG2-comment

Roleset be.02: existential Roles: ARG1-thing that is

Roleset be.03: auxiliary

Roles: do not tag

The following are the examples from the TED talk and the web forum data showing the usage of the three senses of "be":

Table 4: Number of sentences with no automatic semantic parsing output in each data set

| dataset        | genre          | #sentences | #no semantic parse | %no semantic parse |
|----------------|----------------|------------|--------------------|--------------------|
| TED-dev        | public talk    | 934        | 138                | 14.78%             |
| TED-test       | public talk    | 1664       | 237                | 14.24%             |
| BOLT P1-dev    | forum          | 2000       | 229                | 11.45%             |
| BOLT P1-test   | forum          | 1697       | 100                | 5.89%              |
| MetricsMaTr 08 | broadcast news | 221        | 9                  | 4.07%              |

Table 5: Detailed breakdown of the sentences with no semantic frame identified by the automatic semantic parser. (#"be" is the number of sentences that has at least one grammatical and valid semantic frame of the copula or existential sense of "be"; #no verb ( $\leq 10$ ) and #no verb (> 10) are the number of sentences that has no verb in the sentence with the sentence length is "less than or equal to 10" or "greater than 10" respectively; #other is the number of sentences that do not fall into any of the previous categories.)

| dataset        | genre          | #no parse | #"be" | #no verb (≤10) | #no verb (> 10) | #others |
|----------------|----------------|-----------|-------|----------------|-----------------|---------|
| TED-dev        | public talk    | 138       | 110   | 20             | 3               | 5       |
| TED-test       | public talk    | 237       | 191   | 38             | 3               | 5       |
| BOLT P1-dev    | forum          | 229       | 166   | 56             | 6               | 1       |
| BOLT P1-test   | forum          | 100       | 81    | 4              | 5               | 10      |
| MetricsMaTr 08 | broadcast news | 9         | 9     | 0              | 0               | 0       |

## Example 5.1 (copula)

A language is a flash of the human spirit.

## Example 5.2 (existential)

There is no feed.

**Example 5.3 (auxiliary)** [ARG0 *The sun*] *is* [PRED *rising*].

In Example 5.1 and Example 5.2, the automatic semantic parser fails to identify the verb "be" as the predicate and thus fails to construct the corresponding semantic frame. In Example 5.3, the predicate "rising" is correctly identified and the auxiliary sense of "be" is not tagged following the Propbank frameset definition.

We believe that the failure is due to the differences in language usage between formal and informal genres. In formal text genres (i.e. which the automatic semantic parser is trained on), "be" is always used as auxiliary verb together with the present or past participle to realize different tense or voice in grammar. Since Propbank declares that the auxiliary sense of "be" should not be labeled, the automatic semantic parser is biased heavily not to label "be" as the predicate. However, as we can see in Table 5, up to 11% of the sentences in informal genres are having the copula or the existential sense of "be" as the predicate.

Apart from the problem of "be", another common situation for automatic parser fails to identify semantic frames in informal text data is that there is

no predicate verb at all in the sentence. We breakdown the statistics of sentences with no semantic frame according to the sentence length and find that most of these sentences are short phrases with less than 10 words. For these cases, from experiments on similar phenomena, we observed that using the phrasal similarity function for the role fillers in MEANT was still better than BLEU or other ngram based metrics for evaluating translation adequacy at sentence-level, because only very rarely do these sentences have sufficient n-gram counts for the metrics to accurately differentiate translation quality. The following are the examples of the no-verb sentences from the TED talk and the web forum data:

### Example 5.4 (no verb, sentence length<10)

2011 Summer Davos Forum

## Example 5.5 (no verb, sentence length>10)

A photo of Wang Lei , counselor for the Arts Department 2008 , Henan University

## 5.2 Reconstructing the frames of "be"

The results of the error analysis have lead to another interesting research question: is it possible to further improve the performance of the MEANT-tuned SMT system by reconstruct the missing semantic frames of "be" correctly? We manually reconstruct the missing semantic frames of "be" in the MetricsMaTr 2008 data set. We compute the MEANT scores of the three system outputs in

the MetricsMaTr data set and correlate the scores against the adequacy judgments as provided in the MetricsMaTr 2008 data set.

With the reconstructed semantic frames of "be", the correlation of MEANT with human adequacy judgment on MetricsMaTr 2008 broadcast news data set improves from 14.67 to 16.00. These results suggest that correctly reconstructing the missing semantic frames of "be" would possibly serve as a future basis for further improvements the translation quality for informal text.

### 6 Conclusion

We have presented the first ever results to demonstrate that even for informal language genres, tuning an SMT system against MEANT robustly produces more adequate translation than tuning against BLEU or TER, as measured across a battery of commonly used automated metrics, as well as via human subjective evaluation. Surprisingly, tuning against MEANT succeeds in producing adequate output significantly more frequently even though it depends on automatic semantic parses which are notoriously hard to find in these informal genres. We argue that by rewarding preservation of even a portion of the meaning of the translations as captured by semantic frames during the training process, the gains from constraining the SMT system to make more accurate lexical choices and reordering outweigh the losses from fragile automatic SRL on informal language.

We believe that tuning on MEANT will prove equally useful for SMT systems based on any architecture, particularly where the model does not otherwise incorporate semantic information to improve the adequacy of the translations produced. Our encouraging results even on informal language suggest that using MEANT-like objective functions to tune SMT would drive sustainable development of MT toward higher utility. Future work of this line of research includes using more stable and efficient optimization techniques, such as MIRA or PRO, to tune MT systems against MEANT; and tuning MT systems for other language pairs against MEANT.

## Acknowledgment

This material is based upon work supported in part by the Defense Advanced Research Projects Agency (DARPA) under BOLT contract no. HR0011-12-C-0016, and GALE contract

nos. HR0011-06-C-0022 and HR0011-06-C-0023; by the European Union under the FP7 grant agreement no. 287658; and by the Hong Kong Research Grants Council (RGC) research grants GRF620811, GRF621008, and GRF612806. Any opinions, findings and conclusions or recommendations expressed in this material are those of the authors and do not necessarily reflect the views of DARPA, the EU, or RGC.

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