

# Guiding Large Language Models to Post-edit Machine Translation with Error Annotations

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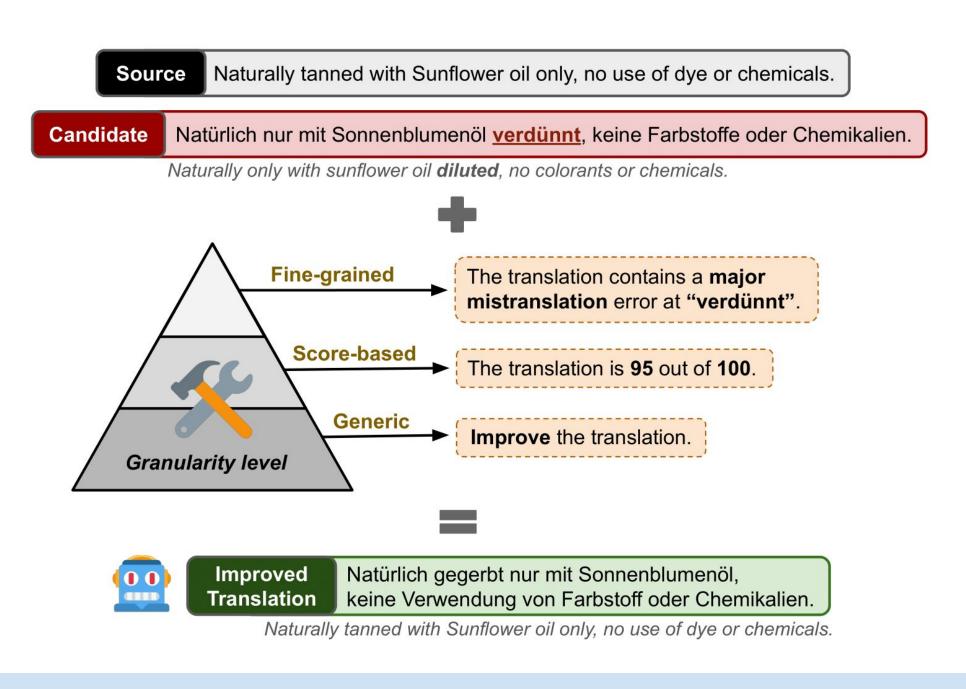
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#### **Motivation**

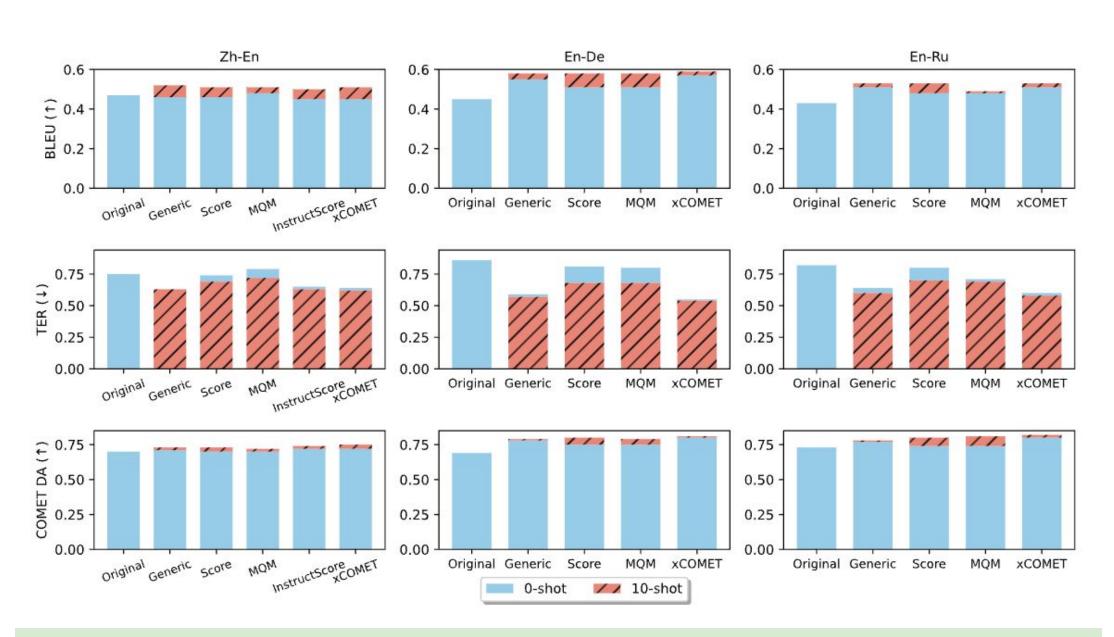
- Performance of LLMs remains uneven with variation across models, languages, and translation directions.
- Exploit the complementary strengths of LLMs and supervised Machine Translation system.

#### Can smaller LLMs use fine-grained feedback to refine their translations?



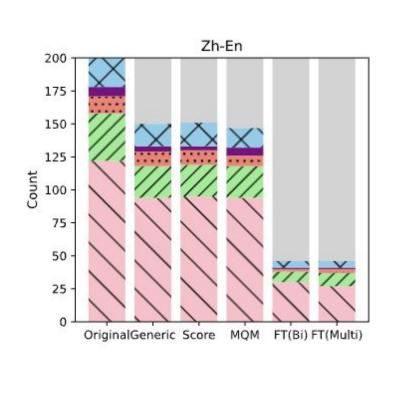
## **Prompting Results**

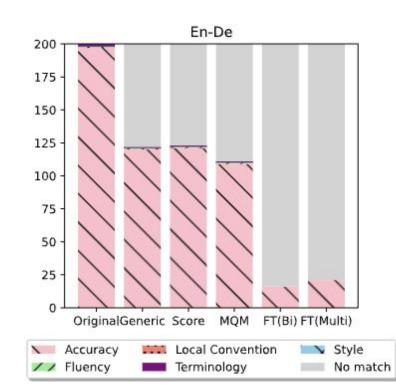
- Prompting LLaMA-2 to post-edit with any forms of feedback consistently improve MT performance.
- Most forms of our tested feedback, regardless of granularity, converge to similar performance.

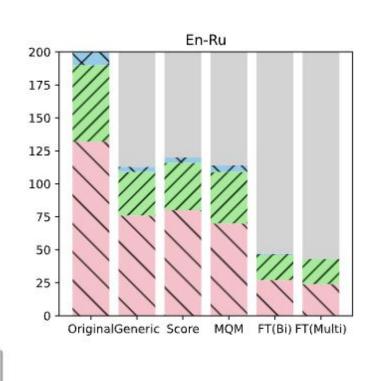


# **Error Analysis**

Fine-tuning with error annotations help align LLM behavior with the provided feedback.







# **Types of Feedback**

Category	Prompt				
Generic	Improve the translation from English to German without any explanation.  English: The newer items are bagged only.  German: Neue Gegenstände werden nur mit Gepäck versehen.  Improved German:				
Score	Improve the translation from English to German without any explanation. This translation is scored 85 out of 100. English: The newer items are bagged only.  German: Neue Gegenstände werden nur mit Gepäck versehen.  Improved German:				
Fine-grained	Improve the translation from English to German based on the identified errors without any explanation.  (1) There is a major mistranslation error at "mit Gepäck versehen".  English: The newer items are bagged only.  German: Neue Gegenstände werden nur mit Gepäck versehen.  Improved German:				

## Fine-tuning Results

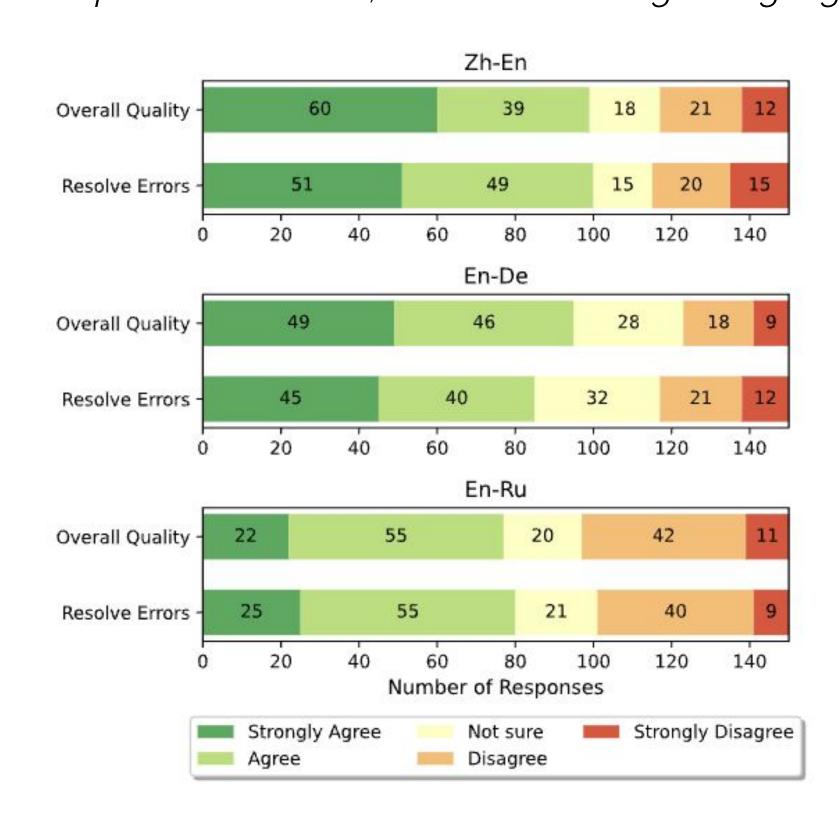
#### **Automatic Evaluation**

Fine-tuning LLaMA-2 with error annotated translations gives an extra boost on top of prompting results.

Language	Type	BLEU (†)	TER $(\downarrow)$	$\mathbf{COMET}_{\mathrm{DA}}$ (†)
	Original	0.47	0.75	0.70
	prompt (k=0)	$0.48^{\dagger}$	$0.72^{\dagger}$	$0.70^{\dagger}$
Zh-En	prompt ( $k=10$ )	$0.51^{\dagger}$	$0.65^{\dagger}$	$0.72^{\dagger}$
	FT (Bi)	$0.53^{\dagger}$	$0.63^{\dagger}$	$0.76^{\dagger}$
	FT (Multi)	$0.53^{\dagger}$	$0.61^{\dagger}$	$0.76^{\dagger}$
	Original	0.45	0.86	0.69
	prompt $(k=0)$	$0.51^{\dagger}$	$0.68^{\dagger}$	$0.75^{\dagger}$
En-De	prompt ( $k=10$ )	$0.58^{\dagger}$	$0.56^{\dagger}$	0.79 <sup>†</sup>
	FT (Bi)	$0.56^{\dagger}$	$0.58^{\dagger}$	0.79 <sup>†</sup>
	FT (Multi)	$0.59^{\dagger}$	$0.55^{\dagger}$	$0.79^{\dagger}$
	Original	0.43	0.82	0.73
	prompt $(k=0)$	$0.48^{\dagger}$	$0.69^{\dagger}$	$0.74^{\dagger}$
En-Ru	prompt ( $k=10$ )	$0.49^{\dagger}$	$0.67^{\dagger}$	$0.79^{\dagger}$
	FT (Bi)	$0.51^{\dagger}$	$0.65^{\dagger}$	$0.80^{\dagger}$
	FT (Multi)	0.52†	0.63	0.80 <sup>†</sup>

#### **Human Evaluation**

Fine-tuning rewrites for naturalness in the target language. (better explains the context, flows better in target language)



### **Discussions**

- Post-editing MT does not require the largest proprietary LLMs and can be done with smaller open-source models.
- Building on LLMs fine-tuned for many translation related tasks is a promising direction for encouraging transfer learning from limited amounts of annotation.