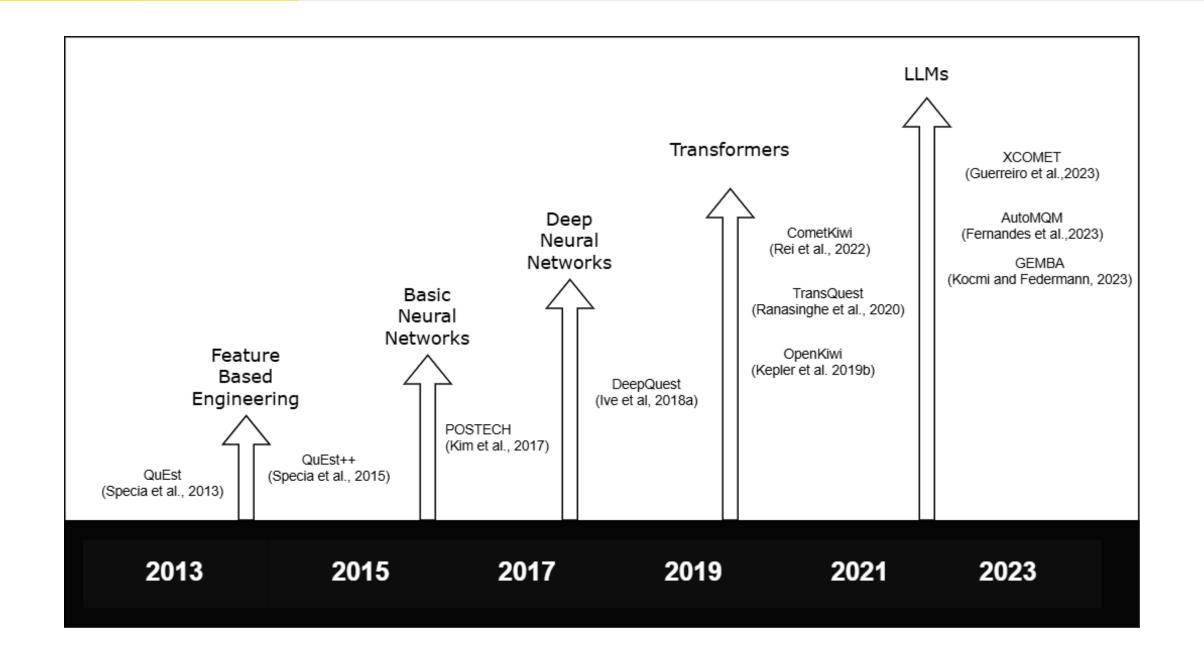
ALOPE: Adaptive Layer Optimization for Translation Quality Estimation using Large Language Models

Archchana Sindhujan, **Diptesh Kanojia**, Constantin Orasan, Shenbin Quan , Cha Chu Chin Matthew



Introduction

- **Evaluating the quality** of the translation is considered a challenging cross-lingual task to **identify the reliability** of the translation and towards improvement of the translation systems.
- Traditional Machine Translation evaluation methods are Reference-based which are resource-intensive.
- Quality estimation (QE) aims to estimate the quality of a translated text without the need for a reference translation - Reduces cost and effort.
 - At different granularity levels Segment-level, Word-level, Error Span Annotation, MQM



Quality Estimation @ Sentence/Segment-level

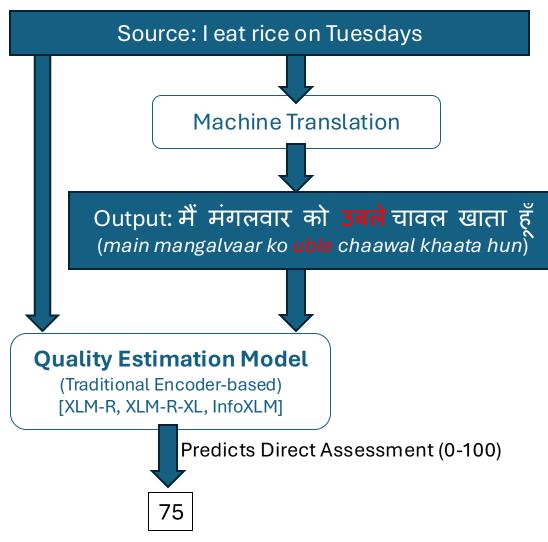
To provide a reliable, automatic measure of translation quality, crucial for system development and user-facing applications – without using a reference.

- Metrics like BLEU, chrF, MetricX need a reference.
- MT is subjective multiple references free order.

Quality Estimation (QE) is task of assessing the quality of machine-translated text in the <u>absence of</u> a human reference translation.

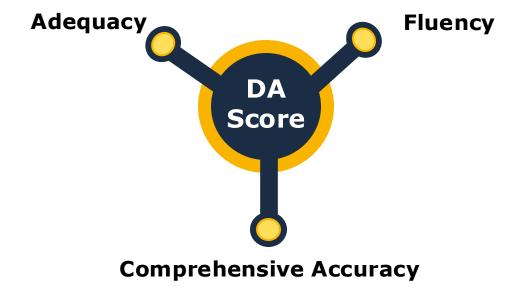
 Segment-Level QE focuses on assigning a quality score to a translated sentence, typically a Direct Assessment (DA) score from 0-100.

• Word-Level QE focuses on tagging each token in source and MT output with a OK/BAD tag, given the translation errors.



Direct Assessment (DA) score

Direct Assessment evaluates machine-translated content by having human assessors rate its quality on a scale of 0 to 100.

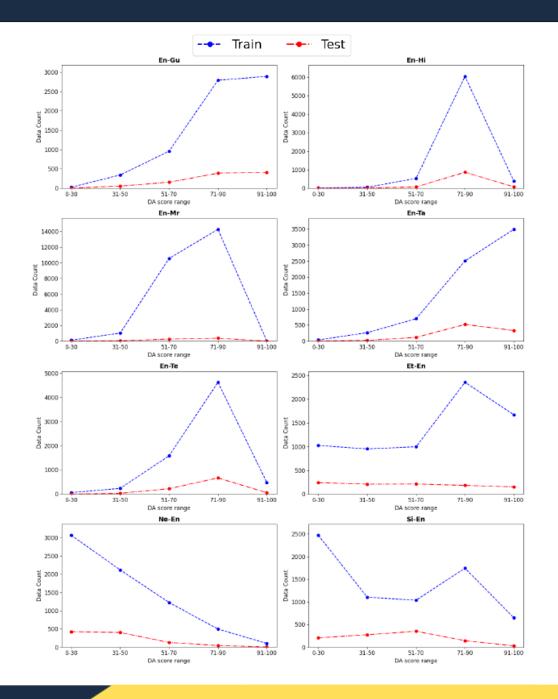


Overall Score	Translation conveys source meaning?	How much translation conveys to source?
1 - 10	Completely inaccurate.	The MT output is unintelligible. Studying the meaning of the sentence is hopeless; even allowing for context, one feels that guessing would be too unreliable. • The translation is incomprehensible, and machine translated output contains a mix of languages/dialects [which are not in the target language] (Adequacy) • None of the keywords are translated in the target language. (Adequacy) • There are major grammatical errors and typos (Fluency)
11 - 30	Inaccurate but contains some keywords.	 Incomprehensible translation (Fluency) Translation contains some keywords but not all (Adequacy) There are numerous grammatical errors and typos.
31 -50	Partially. Target reflects partially the source.	The general idea of the MT output is intelligible only after considerable study. • Translation is only partially understandable, and the overall meaning is not conveyed. (Adequacy and Fluency) • Translation contains some keywords (Adequacy) • There are many grammatical errors and typos (Fluency)
51 - 70	Yes. Target reflects the overall meaning.	The MT output is generally clear and intelligible. Despite some inaccuracies or infelicities of the sentence, one can understand (almost) immediately what it means. • Translation is understandable and reflects the source meaning. (Fluency) • Translation contains most keywords (Adequacy) • Only minor grammar errors (Fluency)
71 - 90	Yes. Target reflects the source meaning without errors.	The MT output is perfectly clear and intelligible. It is grammatical and reads like ordinary text. • Translation is very closed to the source meaning (Fluency) • Translation contains all keywords (Adequacy) • No errors but there are better word choices in the target language. (Adequacy)
91 - 100	Yes. Target reflects source meaning without errors.	 Perfect translation (Adequacy and Fluency) Accurately reflects the meaning of the source (Fluency) No errors

Dataset

- Focus on low-resource language pairs for QE.
- We utilize the DA score dataset from WMT QE shared task for our experiments.
 - En-Gu : English to ગુજરાતી (Gujarati)
 - En-Hi : English to हिन्दी (Hindi)
 - En-Mr : English to **मराठी** (Marathi)
 - En-Ta : English to தமிழ் (Tamil)
 - En-Te : English to తెలుగు (Telugu)

- Et-En : **eesti keel** (Estonian) to English
- Ne-En : नेपाली (Nepali) to English
- Si-En : සිංහල (Sinhala) to English



GEMBA(Kochmi et. al., 2023) Prompt

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Score the following translation from {Source Language} to {Target Language} on a continuous scale from 0 to 100, where score of zero means "no meaning preserved" and score of one hundred means "perfect meaning and grammar".

{Source Language} source: {Source Sentence}

{Target Language} translation: {Translated Sentence}

Score:
```

Experimental Settings

- **Zero-Shot**: Model generates outputs using pre-trained knowledge and the generalization ability.
- Standard Instruction-Tuning: Adapting a model using a dataset that includes explicit instructions for specific tasks.
- **ALOPE :** Enhances fine-tuning by leveraging regression headers across multiple Transformer layers.

Models

- All experiments conducted with open-source LLMs with 8B parameters or less:
 - ➤ Llama-2-7B
 - > Llama 3.1-8B
 - Llama 3.2-3B
 - > Aya-expanse-8B

Challenges for LLM-based QE

- LLMs are **optimized for next-token prediction**, excelling at generative tasks but struggling with regression-based goals like quality estimation which require numerical and cross-lingual reasoning.
 - This limits their ability to capture fine-grained relationships between language features and numerical scores.
- Although LLMs refine context through multiple Transformer layers, standard practice usually <u>predicts with the final layer for output</u>.

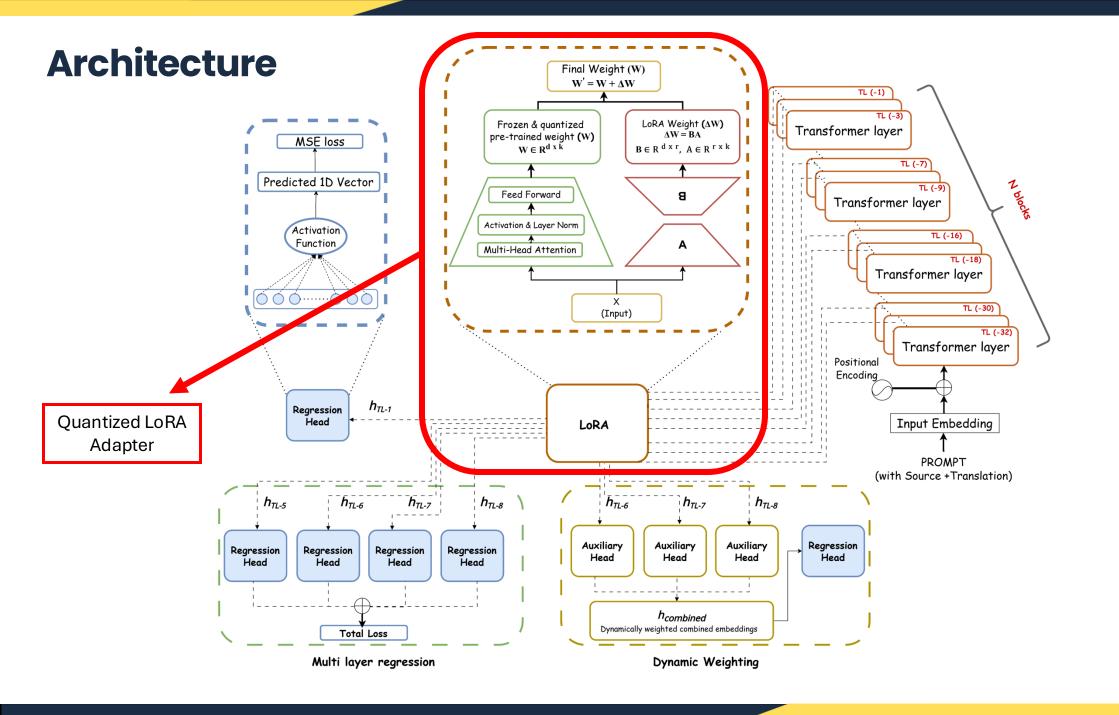
Why ALOPE (Adaptive Layer Optimization)?

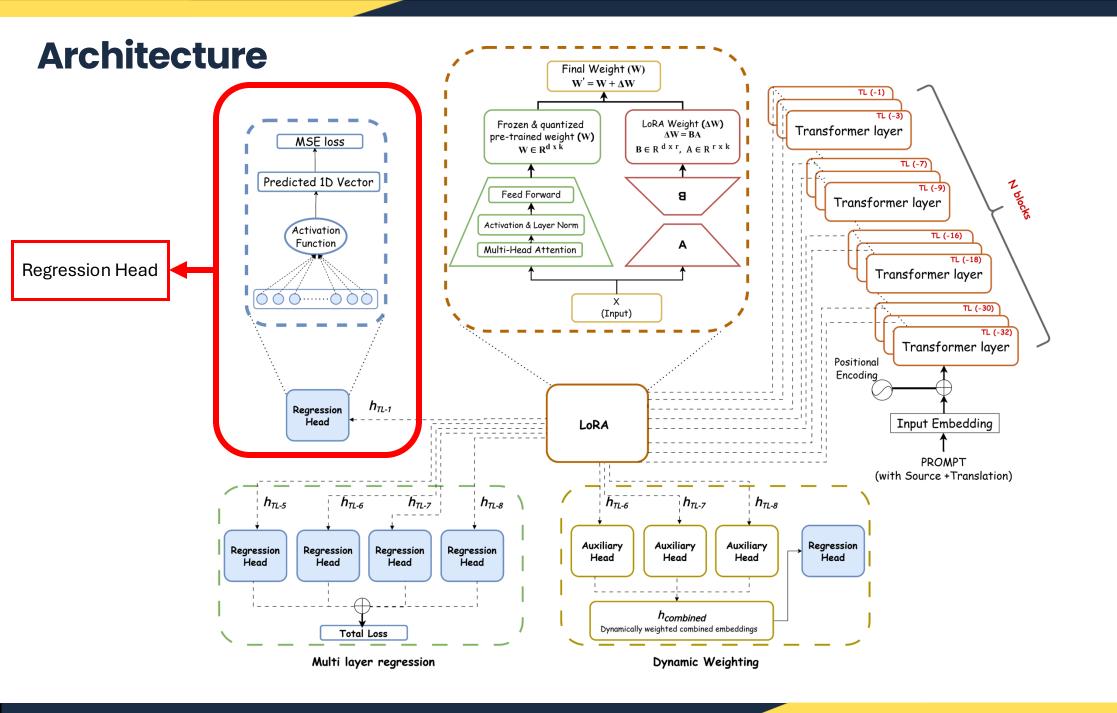
- ALOPE enables LLMs to perform regression by integrating a regression head within the low-rank adapted Transformer architecture.
- We <u>analyze the impact of adaptive regression heads</u> to determine which Transformer layers contribute most to <u>enhancing the performance of LLMs for QE</u> with low-resource language pairs.

Architecture Final Weight (W) $W' = W + \Delta W$ TL (-1) LoRA Weight (ΔW) $\Delta W = BA$ Frozen & quantized Transformer layer pre-trained weight (W) $W \in \mathbb{R}^{d \times k}$ MSE loss $B \in R^{d \times r}$, $A \in R^{r \times k}$ TL (-7) Predicted 1D Vector Feed Forward В Transformer layer Activation & Layer Norm Activation TL (-16) Function Multi-Head Attention Transformer layer TL (-30) (Input) TL (-32) Transformer layer Positional Encoding h_{TL-1} Regression Input Embedding Head LoRA PROMPT (with Source +Translation) h_{TL-7} h_{TL-8} h_{TL-8} h_{TL-6} h_{TL-7} h_{TL-5} h_{TL-6} **Auxiliary Auxiliary** Auxiliary Regression Regression Regression Regression Regression Head Head Head Head Head Head Head Head h_{combined} Dynamically weighted combined embeddings Total Loss Multi layer regression Dynamic Weighting

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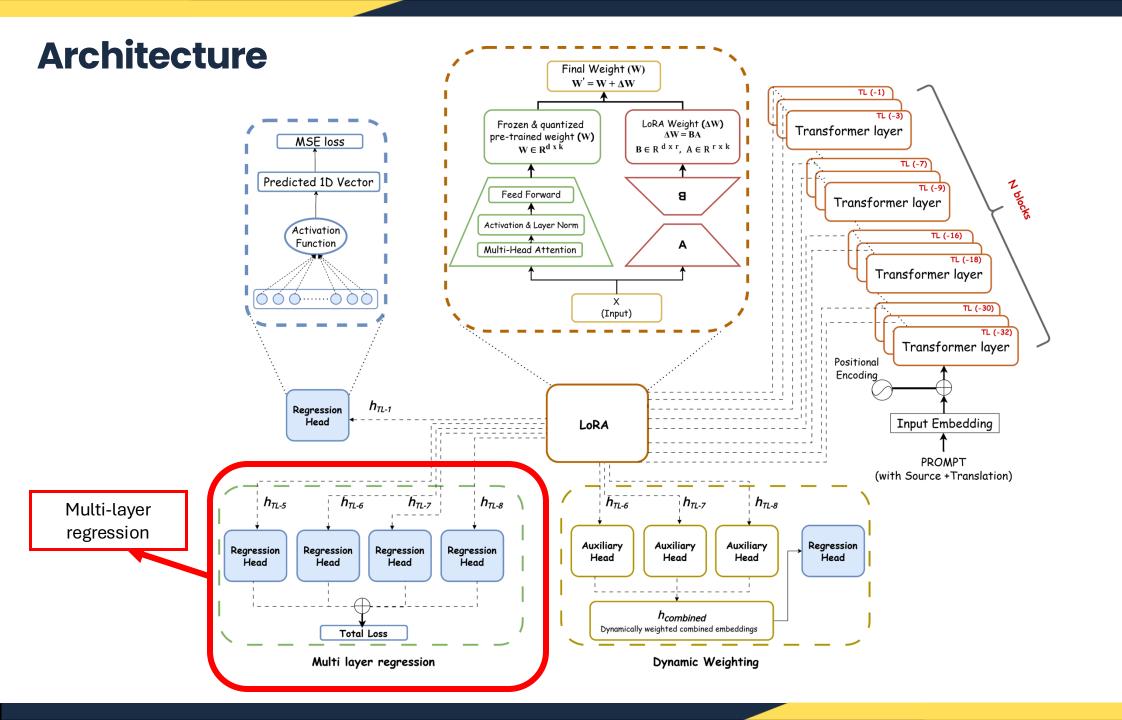




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Multi layer regression Dynamic Weighting

Architecture Final Weight (W) $W' = W + \Delta W$ TL (-1) LoRA Weight (ΔW) $\Delta W = BA$ Frozen & quantized Transformer layer pre-trained weight (W) $W \in R^{d \times k}$ MSE loss $B \in R^{d \times r}$, $A \in R^{r \times k}$ TL (-7) Predicted 1D Vector Feed Forward В Transformer layer Activation & Layer Norm Activation TL (-16) Function Multi-Head Attention Transformer layer ÓÓÓ......ÒÒÒ TL (-30) (Input) TL (-32) Transformer layer Positional Encoding h_{TL-1} Regression Input Embedding Head LoRA PROMPT (ith Source +Translation) h_{TL-7} h_{TL-8} \bar{h}_{TL-6} h_{TL-8} h_{TL-5} h_{TL-7} h_{TL-6} **Auxiliary Auxiliary** Auxiliary Regression Regression Regression Regression Regression Head Head Head Head Head Head Head Head Dynamic Weighting h_{combined} Dynamically weighted combined embeddings Total Loss Multi layer regression Dynamic Weighting



Evaluation Metrics for DA prediction in QE

- Spearman's correlation between predicted scores and human DA mean scores
- Applied Williams test to compare correlation significance between models and to identify whether top models outperform others statistically.

ALOPE		1	ı			1				
ALOIL	Δ	Model	En-Gu	En-Hi	En-Mr	En-Ta	En-Te	Et-En	Ne-En	Si-En
	/ \	llama-2-7b	0.563	0.414	0.609	0.525	0.356	0.742	0.596	0.565
Final Transformer	<u>+</u>	llama 3.1-8B	0.594	0.469	0.620	0.567	0.363	0.734	0.647	0.547
Layer	ř	llama 3.2-3B	0.604	0.477	0.636	0.580	0.348	0.735	0.674	0.543
		aya-expanse-8b	0.068	0.178	0.219	-0.006	0.275	0.115	0.012	0.077
		llama-2-7b	0.567	0.336	0.542	0.484	0.317	0.739	0.606	0.573
Intermediate	-7	llama 3.1-8B	0.590	0.477	0.625	0.528	0.388	0.744	0.638	0.544
Transformer Layer	TL (-7)	llama 3.2-3B	0.606	0.479	0.617	0.585	0.369	0.751	0.664	0.553
		aya-expanse-8b	0.538	0.447	0.597	0.528	0.347	0.741	0.646	0.544
		llama-2-7b	0.360	0.301	0.361	0.254	0.293	0.405	0.164	0.049
Intermediate	TL (-11)	llama 3.1-8B	0.514	0.412	0.609	0.438	0.304	0.148	0.554	0.493
Transformer Layer		llama 3.2-3B	0.594	0.476	0.605	0.610	0.373	0.748	0.678	0.560
	-	aya-expanse-8b	0.490	0.411	0.572	0.445	0.336	0.569	0.453	0.439
		llama-2-7b	0.540	0.381	0.585	0.482	0.308	0.751	0.580	0.569
Intermediate	.16	llama 3.1-8B	0.558	0.453	0.602	0.523	0.350	0.737	0.652	0.513
Transformer Layer	TL (-16)	llama 3.2-3B	0.557	0.459	0.597	0.547	0.338	0.745	0.682	0.567
	-	aya-expanse-8b	0.467	0.390	0.557	0.481	0.314	0.727	0.576	0.540
		llama-2-7b	0.470	0.405	0.544	0.460	0.338	0.684	0.508	0.534
Lower-level	TL (-20)	llama 3.1-8B	0.484	0.394	0.553	0.321	0.172	0.649	0.524	0.494
Transformer Layer		llama 3.2-3B	0.430	0.408	0.579	0.303	0.286	0.601	0.488	0.464
	-	aya-expanse-8b	0.437	0.300	0.488	0.263	0.287	0.483	0.438	0.395
		llama-2-7b	0.500	0.421	0.538	0.379	0.239	0.630	0.507	0.472
Lower-level	(-24)	llama 3.1-8B	0.421	0.378	0.552	0.330	0.290	0.515	0.530	0.464
Transformer Layer) 	llama 3.2-3B	0.443	0.376	0.507	0.367	0.299	0.559	0.528	0.487
	\ - /	aya-expanse-8b	0.375	0.319	0.440	0.337	0.220	0.393	0.407	0.345
		<u> </u>								

	Model	En-Gu	En-Hi	En-Mr	En-Ta	En-Te	Et-En	Ne-En	Si-En
	llama-2-7b	0.563	0.414	0.609	0.525	0.356	0.742	0.596	0.565
(-1)	llama 3.1-8B	0.594	0.469	0.620	0.567	0.363	0.734	0.647	0.547
 	llama 3.2-3B	0.604	0.477	0.636	0.580	0.348	0.735	0.674	0.543
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	aya-expanse-8b	0.437	0.300	0.488	0.263	0.287	0.483	0.438	0.395
	llama-2-7b	0.500	0.421	0.538	0.379	0.239	0.630	0.507	0.472
(-24)	llama 3.1-8B	0.421	0.378	0.552	0.330	0.290	0.515	0.530	0.464
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	aya-expanse-8b	0.375	0.319	0.440	0.337	0.220	0.393	0.407	0.345

	Model	En-Gu	En-Hi	En-Mr	En-Ta	En-Te	Et-En	Ne-En	Si-En	Avg.
	llama-2-7b	0.563	0.414	0.609	0.525	0.356	0.742	0.596	0.565	0.546
(1	llama 3.1-8B	0.594	0.469	0.620	0.567	0.363	0.734	0.647	0.547	0.567
TL (-1)	llama 3.2-3B	0.604	0.477	0.636	0.580	0.348	0.735	0.674	0.543	0.5 <i>7</i> 5
F	aya-expanse-8b	0.068	0.178	0.219	-0.006	0.275	0.115	0.012	0.077	0.117
	Avg	0.457	0.385	0.521	0.416	0.335	0.581	0.482	0.433	
	llama-2-7b	0.567	0.336	0.542	0.484	0.317	0.739	0.606	0.573	0.520
[C	llama 3.1-8B	0.590	0.477	0.625	0.528	0.388	0.744	0.638	0.544	0.567
TL (-7)	llama 3.2-3B	0.606	0.479	0.617	0.585	0.369	0.751	0.664	0.553	0.5 <i>7</i> 8
F	aya-expanse-8b	0.538	0.447	0.507	0.528	0.347	0.741	0.646	0.544	0.549
	Avg	0.575	0.435	0.595	0.531	0.355	0.744	0.639	0.554	
	llama-2-7b	0.360	0.301	0.361	0.254	0.293	0.405	0.164	0.049	0.273
1)	llama 3.1-8B	0.514	0.412	0.609	0.438	0.304	0.148	0.554	0.493	0.434
TL (-11)	llama 3.2-3B	0.594	0.476	0.605	0.610	0.373	0.748	0.678	0.560	0.581
	aya-expanse-8b	0.490	0.411	0.572	0.445	0.336	0.569	0.453	0.439	0.464
	Avg	0.489	0.400	0.537	0.437	0.327	0.467	0.462	0.385	
	llama-2-7b	0.540	0.381	0.585	0.482	0.308	0.751	0.580	0.569	0.524
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TL (-16)	llama 3.2-3B	0.557	0.459	0.597	0.547	0.338	0.745	0.682	0.567	0.561
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	Avg	0.530	0.421	0.585	0.508	0.327	0.740	0.622	0.547	
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	aya-expanse-8b	0.437	0.300	0.488	0.263	0.287	0.483	0.438	0.395	0.386
	Avg	0.455	0.377	0.541	0.337	0.271	0.604	0.490	0.472	
	llama-2-7b	0.500	0.421	0.538	0.379	0.239	0.630	0.507	0.472	0.461
4	llama 3.1-8B	0.421	0.378	0.552	0.330	0.290	0.515	0.530	0.464	0.435
TL (-24)	llama 3.2-3B	0.443	0.376	0.507	0.367	0.299	0.559	0.528	0.487	0.446
🗗	aya-expanse-8b	0.375	0.319	0.440	0.337	0.220	0.393	0.407	0.345	0.354
	Avg	0.435	0.373	0.509	0.353	0.262	0.524	0.493	0.442	

	Model	En-Gu	En-Hi	En-Mr	En-Ta	En-Te	Et-En	Ne-En	Si-En	Avg.
	llama-2-7b	0.563	0.414	0.609	0.525	0.356	0.742	0.596	0.565	0.546
	llama 3.1-8B	0.594	0.469	0.620	0.567	0.363	0.734	0.647	0.547	0.567
Т. (-1)	llama 3.2-3B	0.604	0.477	0.636	0.580	0.348	0.735	0.674	0.543	0.5 <i>7</i> 5
=	aya-expanse-8b	0.068	0.178	0.219	-0.006	0.275	0.115	0.012	0.077	0.117
	Avg	0.457	0.385	0.521	0.416	0.335	0.581	0.482	0.433	
	llama-2-7b	0.567	0.336	0.542	0.484	0.317	0.739	0.606	0.573	0.520
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TL (-7)	llama 3.2-3B	0.606	0.479	0.617	0.585	0.369	0.751	0.664	0.553	0.5 <i>7</i> 8
	aya-expanse-8b	0.538	0.447	0.597	0.528	0.347	0.741	0.646	0.544	0.549
	Avg	0.5 <i>7</i> 5	0.435	0.595	0.531	0.355	0.744	0.639	0.554	
	llama-2-7b	0.360	0.301	0.361	0.254	0.293	0.405	0.164	0.049	0.273
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	aya-expanse-8b	0.490	0.411	0.572	0.445	0.336	0.569	0.453	0.439	0.464
	Avg	0.489	0.400	0.537	0.437	0.327	0.467	0.462	0.385	
	llama-2-7b	0.540	0.381	0.585	0.482	0.308	0.751	0.580	0.569	0.524
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=	aya-expanse-8b	0.467	0.390	0.557	0.481	0.314	0.727	0.576	0.540	0.506
	Avg	0.530	0.421	0.585	0.508	0.327	0.740	0.622	0.547	
	llama-2-7b	0.470	0.405	0.544	0.460	0.338	0.684	0.508	0.534	0.493
(0)	llama 3.1-8B	0.484	0.394	0.553	0.321	0.172	0.649	0.524	0.494	0.449
TL (-20)	llama 3.2-3B	0.430	0.408	0.579	0.303	0.286	0.601	0.488	0.464	0.445
=	aya-expanse-8b	0.437	0.300	0.488	0.263	0.287	0.483	0.438	0.395	0.386
	Avg	0.455	0.377	0.541	0.337	0.271	0.604	0.490	0.472	
	llama-2-7b	0.500	0.421	0.538	0.379	0.239	0.630	0.507	0.472	0.461
4	llama 3.1-8B	0.421	0.378	0.552	0.330	0.290	0.515	0.530	0.464	0.435
TL (-24)	llama 3.2-3B	0.443	0.376	0.507	0.367	0.299	0.559	0.528	0.487	0.446
	aya-expanse-8b	0.375	0.319	0.440	0.337	0.220	0.393	0.407	0.345	0.354
	Avg	0.435	0.373	0.509	0.353	0.262	0.524	0.493	0.442	

	Model	En-Gu	En-Hi	En-Mr	En-Ta	En-Te	Et-En	Ne-En	Si-En	Avg.
	llama-2-7b	0.563	0.414	0.609	0.525	0.356	0.742	0.596	0.565	0.546
	llama 3.1-8B	0.594	0.469	0.620	0.567	0.363	0.734	0.647	0.547	0.567
TL (-1)	llama 3.2-3B	0.604	0.477	0.636	0.580	0.348	0.735	0.674	0.543	0.5 <i>7</i> 5
=	aya-expanse-8b	0.068	0.178	0.219	-0.006	0.275	0.115	0.012	0.077	0.117
	Avg	0.457	0.385	0.521	0.416	0.335	0.581	0.482	0.433	
	llama-2-7b	0.567	0.336	0.542	0.484	0.317	0.739	0.606	0.573	0.520
	llama 3.1-8B	0.590	0.477	0.625	0.528	0.388	0.744	0.638	0.544	0.567
TL (-7)	llama 3.2-3B	0.606	0.479	0.617	0.585	0.369	0.751	0.664	0.553	0.5 <i>7</i> 8
=	aya-expanse-8b	0.538	0.447	0.597	0.528	0.347	0.741	0.646	0.544	0.549
	Avg	0.5 <i>7</i> 5	0.435	0.595	0.531	0.355	0.744	0.639	0.554	
	llama-2-7b	0.360	0.301	0.361	0.254	0.293	0.405	0.164	0.049	0.273
1	llama 3.1-8B	0.514	0.412	0.609	0.438	0.304	0.148	0.554	0.493	0.434
Т. (-11)	llama 3.2-3B	0.594	0.476	0.605	0.610	0.373	0.748	0.678	0.560	0.581
	aya-expanse-8b	0.490	0.411	0.572	0.445	0.336	0.569	0.453	0.439	0.464
	Avg	0.489	0.400	0.537	0.437	0.327	0.467	0.462	0.385	
	llama-2-7b	0.540	0.381	0.585	0.482	0.308	0.751	0.580	0.569	0.524
9	llama 3.1-8B	0.558	0.453	0.602	0.523	0.350	0.737	0.652	0.513	0.548
TL (-16)	llama 3.2-3B	0.557	0.459	0.597	0.547	0.338	0.745	0.682	0.567	0.561
=	aya-expanse-8b	0.467	0.390	0.557	0.481	0.314	0.727	0.576	0.540	0.506
	Avg	0.530	0.421	0.585	0.508	0.327	0.740	0.622	0.547	
	llama-2-7b	0.470	0.405	0.544	0.460	0.338	0.684	0.508	0.534	0.493
6	llama 3.1-8B	0.484	0.394	0.553	0.321	0.172	0.649	0.524	0.494	0.449
ТL (-20)	llama 3.2-3B	0.430	0.408	0.579	0.303	0.286	0.601	0.488	0.464	0.445
	aya-expanse-8b	0.437	0.300	0.488	0.263	0.287	0.483	0.438	0.395	0.386
	Avg	0.455	0.377	0.541	0.337	0.271	0.604	0.490	0.472	
	llama-2-7b	0.500	0.421	0.538	0.379	0.239	0.630	0.507	0.472	0.461
4	llama 3.1-8B	0.421	0.378	0.552	0.330	0.290	0.515	0.530	0.464	0.435
ТL (-24)	llama 3.2-3B	0.443	0.376	0.507	0.367	0.299	0.559	0.528	0.487	0.446
	aya-expanse-8b	0.375	0.319	0.440	0.337	0.220	0.393	0.407	0.345	0.354
	Avg	0.435	0.373	0.509	0.353	0.262	0.524	0.493	0.442	

	Model	En-Gu	En-Hi	En-Mr	En-Ta	En-Te	Et-En	Ne-En	Si-En	Avg.
	llama-2-7b	0.563	0.414	0.609	0.525	0.356	0.742	0.596	0.565	0.546
	llama 3.1-8B	0.594	0.469	0.620	0.567	0.363	0.734	0.647	0.547	0.567
TL (-1)	llama 3.2-3B	0.604	0.477	0.636	0.580	0.348	0.735	0.674	0.543	0.5 <i>7</i> 5
=	aya-expanse-8b	0.068	0.178	0.219	-0.006	0.275	0.115	0.012	0.077	0.117
	Avg	0.457	0.385	0.521	0.416	0.335	0.581	0.482	0.433	
	llama-2-7b	0.567	0.336	0.542	0.484	0.317	0.739	0.606	0.573	0.520
	llama 3.1-8B	0.590	0.477	0.625	0.528	0.388	0.744	0.638	0.544	0.567
TL (-7)	llama 3.2-3B	0.606	0.479	0.617	0.585	0.369	0.751	0.664	0.553	0.5 <i>7</i> 8
=	aya-expanse-8b	0.538	0.447	0.597	0.528	0.347	0.741	0.646	0.544	0.549
	Avg	0.5 <i>7</i> 5	0.435	0.595	0.531	0.355	0.744	0.639	0.554	
	llama-2-7b	0.360	0.301	0.361	0.254	0.293	0.405	0.164	0.049	0.2 <i>7</i> 3
1	llama 3.1-8B	0.514	0.412	0.609	0.438	0.304	0.148	0.554	0.493	0.434
Т. (-11)	llama 3.2-3B	0.594	0.476	0.605	0.610	0.373	0.748	0.678	0.560	0.581
≓	aya-expanse-8b	0.490	0.411	0.572	0.445	0.336	0.569	0.453	0.439	0.464
	Avg	0.489	0.400	0.537	0.437	0.327	0.467	0.462	0.385	
	llama-2-7b	0.540	0.381	0.585	0.482	0.308	0.751	0.580	0.569	0.524
(9)	llama 3.1-8B	0.558	0.453	0.602	0.523	0.350	0.737	0.652	0.513	0.548
TL (-16)	llama 3.2-3B	0.557	0.459	0.597	0.547	0.338	0.745	0.682	0.567	0.561
=	aya-expanse-8b	0.467	0.390	0.557	0.481	0.314	0.727	0.576	0.540	0.506
	Avg	0.530	0.421	0.585	0.508	0.327	0.740	0.622	0.54/	
	llama-2-7b	0.470	0.405	0.544	0.460	0.338	0.684	0.508	0.534	0.493
(0;	llama 3.1-8B	0.484	0.394	0.553	0.321	0.172	0.649	0.524	0.494	0.449
TL (-20)	llama 3.2-3B	0.430	0.408	0.579	0.303	0.286	0.601	0.488	0.464	0.445
=	aya-expanse-8b	0.437	0.300	0.488	0.263	0.287	0.483	0.438	0.395	0.386
	Avg	0.455	0.377	0.541	0.337	0.271	0.604	0.490	0.472	
	llama-2-7b	0.500	0.421	0.538	0.379	0.239	0.630	0.507	0.472	0.461
(4)	llama 3.1-8B	0.421	0.378	0.552	0.330	0.290	0.515	0.530	0.464	0.435
TL (-24)	llama 3.2-3B	0.443	0.376	0.507	0.367	0.299	0.559	0.528	0.487	0.446
F	aya-expanse-8b	0.375	0.319	0.440	0.337	0.220	0.393	0.407	0.345	0.354
	Avg	0.435	0.373	0.509	0.353	0.262	0.524	0.493	0.442	

Approaches with ALOPE vs SFIT

En-Gu

En-Te

-8 to -11

-12 to -16

-1 to -7

-8 to -11

-12 to -16

-1 to -7

-8 to -11

-12 to -16

-1 to -7

-8 to -11

-12 to -16

0.7

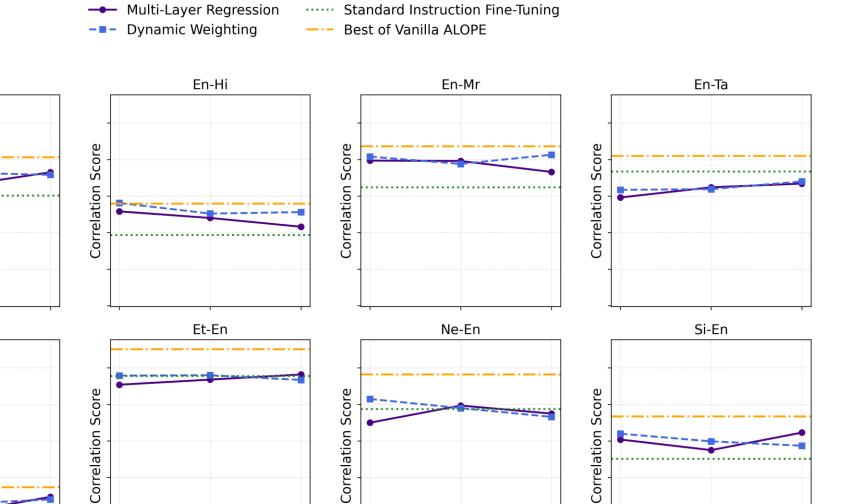
Correlation Score

0.3

0.7

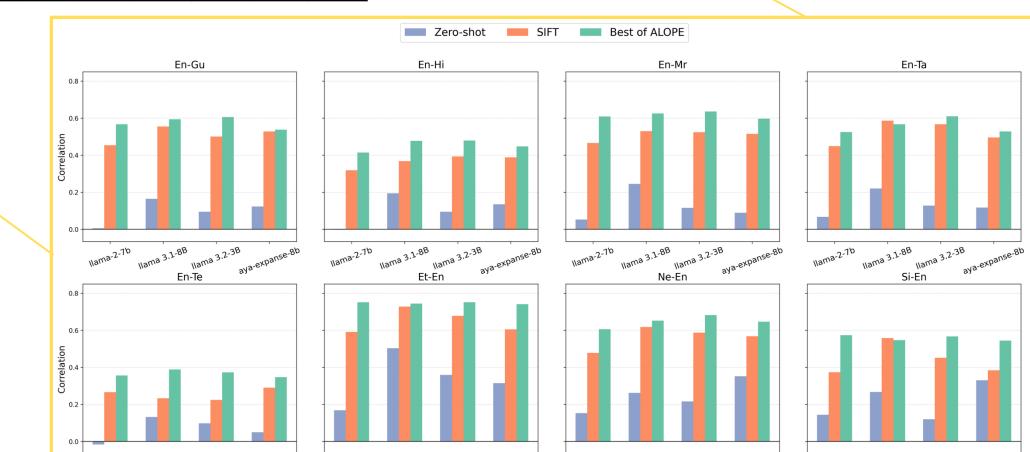
Correlation Score

-1 to -7



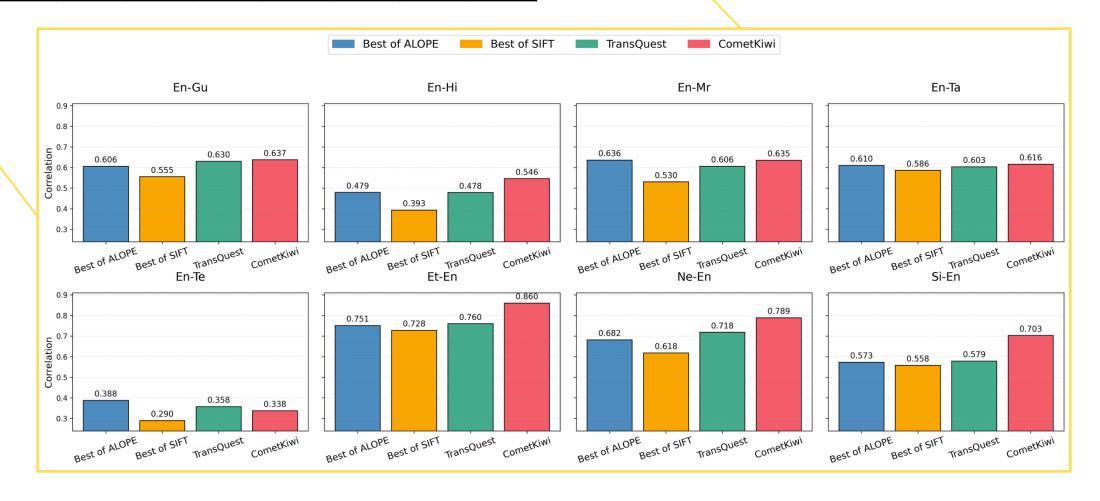
	Model	En-Gu	En-Hi	En-Mr	En-Ta	En-Te	Et-En	Ne-En	Si-En
Τ	llama-2-7b	0.006	-0.002	0.053	0.067	-0.016	0.168	0.153	0.144
-shot	llama 3.1-8B	0.164	0.194	0.245	0.220	0.132	0.503	0.262	0.267
Zero-	llama 3.2-3B	0.095	0.095	0.116	0.128	0.098	0.359	0.216	0.120
Ž	aya-expanse-8B	0.123	0.135	0.089	0.117	0.049	0.315	0.352	0.330
	llama-2-7b	0.454	0.319	0.466	0.449	0.266	0.591	0.478	0.374
SIFT	llama 3.1-8B	0.555	0.368	0.530	0.586	0.233	0.728	0.618	0.558
<u>S</u>	llama 3.2-3B	0.501	0.393	0.524	0.567	0.224	0.678	0.587	0.451
	aya-expanse-8B	0.528	0.388	0.515	0.496	0.290	0.605	0.568	0.384
l	llama-2-7b	0.567	0.414	0.609	0.525	0.356	0.751	0.606	0.573
t of DPE	llama 3.1-8B	0.594	0.477	0.625	0.567	0.388	0.744	0.652	0.547
Best ALOI	llama 3.2-3B	0.606	0.479	0.636	0.610	0.373	0.751	0.682	0.567
	aya-expanse-8B	0.538	0.447	0.597	0.528	0.347	0.741	0.646	0.544

ALOPE vs. Zero-shot vs. SIFT

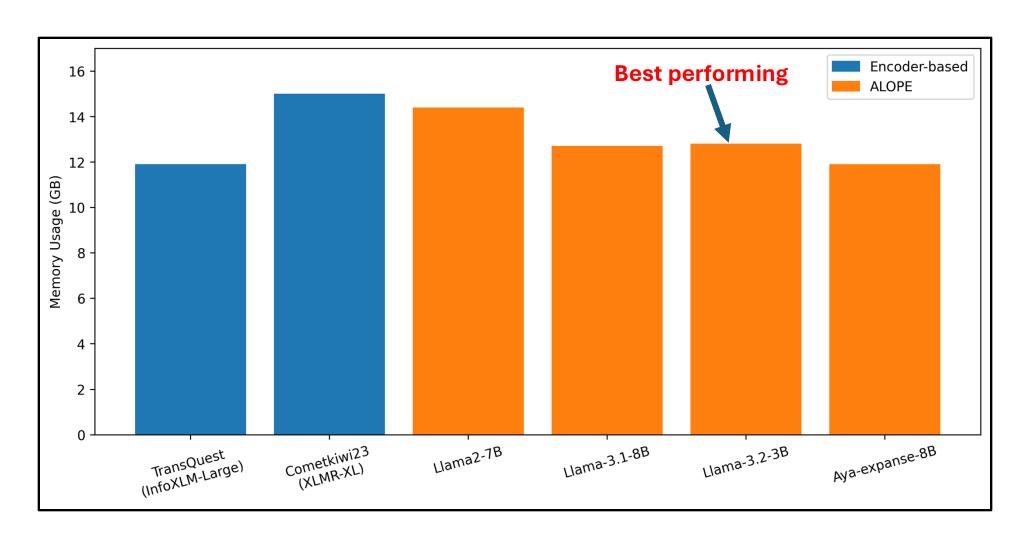


Approach	En-Gu	En-Hi	En-Mr	En-Ta	En-Te	Et-En	Ne-En	Si-En
Best ALOPE	0.606	0.479	0.636	0.610	0.388	0.751	0.682	0.573
Best from SIFT	0.555	0.393	0.530	0.586	0.290	0.728	0.618	0.558
TransQuest	0.630	0.478	0.606	0.603	0.358	0.760	0.718	0.579
CometKiwi	0.637	0.546	0.635	0.616	0.338	0.860	0.789	0.703

ALOPE vs SIFT vs SOTA Approaches



Computational Efficiency



Conclusion

Novel framework

Regression heads + LoRA on LLMs

Performance

Outperforms SIFT; matches / exceeds the performance of SOTA approaches QE

Findings

- Intermediate layers (TL-7, TL-11) => most effective for low-resource cross-lingual QE
- Mid=> layers stabilize earlier when English is target
- LLaMA-3.2 => best overall despite small size

Extension to the framework

Dynamic weighting & multi-head regression improve baseline SFIT

Efficiency

Competitive GPU memory, practical & scalable

Future vision

- Enhancing ALOPE for error reasoning and automatic post-editing
- Look out for ALOPE-RL, and ALOPE-APE branches of this work!

Thank You ©

Questions?