

Parallel Corpora for Low-Resourced Middle Eastern Languages

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Background: Languages in the Middle East

- Remarkable linguistic diversity in the Middle East
- 400+ million people speaking lots of "languages"
- Only a handful of those languages are officially recognized



Background: Under-represented Middle Eastern Languages

- 60 varieties in the region identified as **endangered** by UNESCO [Moseley, 2010]
- Many face existential threats
 - Systematic assimilation campaigns
 - Limited educational resources
 - Younger generations have **fewer opportunities** to develop literacy in heritage languages
 - Lack of standardization

"An endangered language will progress if its speakers can make use of electronic technology."

- [Crystal, 2002] (Language Death)

Background: Limited Technological Progress for Middle Eastern NLP

Little or no progress in language and speech technologies

- Limited keyboard implementations and technological support
- Limited community support for resource development
- Lack of corpora, including parallel ones
- Only a few languages have Wikipedia portals
- 8,000–50,000 articles where available

	Resources								Tools			
	Grammar	Corpus	UniMorph	an An	WordNet	NLLB	Wiktionary	Wikipedia	LID	MT	Spell checke	ASR
Arabic	✓	√	\checkmark	√	√	\checkmark	√	√	✓	√	√	\checkmark
Hebrew	✓	\checkmark										
Turkish	✓	\checkmark	✓	\checkmark	\checkmark	\checkmark						
Persian	✓	✓	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	✓	✓	\checkmark	\checkmark
Northern Kurdish	1	✓	\checkmark	✓	X	✓	\checkmark	\checkmark	✓	✓	✓	\checkmark
Central Kurdish	✓	✓	\checkmark	X	✓	✓	✓	\checkmark	✓	✓	✓	\checkmark
Southern Kurdish	✓	\checkmark	\checkmark	X	X	X	X	Х	✓	X	X	X
Mazandarani	X	X	X	X	X	X	X	\checkmark	✓	Х	X	X
Gilaki	√	√	X	X	X	X	X	✓	√	X	X	X
Talysh	1	✓	Х	X	X	X	X	✓	Х	X	X	X
Zazaki	1	✓	Х	Х	Х	Х	√	√	√	Х	X	X
Hawrami	1	1	X	X	X	X	X	Х	1	X	X	X
Laki	1	X	X	X	X	X	X	X	X	X	X	X
Luri Bakhtiari	✓	X	X	X	X	X	X	X	X	X	X	X

PARME: Community-Driven Approach

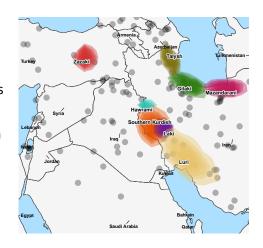
Objective: parallel corpora for under-represented Middle Eastern languages (PARME)

⇒ Community-driven participatory research [Nekoto et al., 2020]



PARME: Community Building

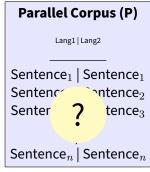
- Intensive campaign (Aug-Oct 2024)
- Twitter/X & LinkedIn for public announcements
- Direct outreach to academics and native speakers
- Cold messaging to published authors & experts
- Mixed Reception: enthusiasm but also skepticism
 - ⇒ 45 volunteers for eight severely under-resourced languages
 - ⇒ Each language is spoken by 0.3 to 5 million speakers



PARME: Data Collection

What sentences should the volunteers translate into their languages?

- Sample randomly? → limited resources and time
- Instead, select sentences strategically [Ambati et al., 2011]
 - Select sentences from a parallel corpus in Lang1-Lang2
 - Lang1: Language familiar to translator (Farsi, Turkish, Arabic or Kurdish)
 - Lang2: High-resource language (English)
 - Maximize lexical diversity
 - Enhance semantic richness



PARME: Strategic Data Selection

- 1. Start with Bilingual Parallel Corpus P
 - 2. Randomly select $N \operatorname{sentences} \Rightarrow C$
- 3. For each candidate: calculate diversity score
 - 4. Select $t \circ p n$ scoring sentence pairs
 - 5. Add to corpus ${\it C}$ Repeat until corpus size

N, n = 3000 / corpus size: 15,000 \Rightarrow a trilingual corpus (Lang1, Lang2, X)

Diversity Score Method:

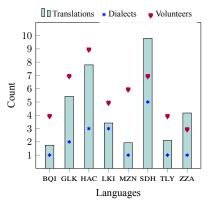
- **D** = Edit distance on Lang1 sentences
- **S** = Semantic similarity on Lang2 sentences

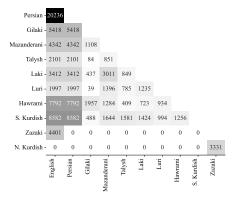
$$\mathsf{score}_i = \frac{\overline{D}_i}{\overline{\mathcal{S}}_i}$$

 $D\uparrow$, $S\downarrow \Rightarrow$ score \uparrow Rewards lexically diverse sentences with different semantic content

PARME: Manual Translation

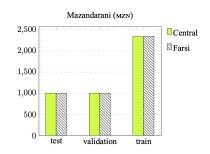
- Manually translate on spreadsheets
- Follow translation guidelines
 - Consistent orthography (per translator)
 - Translate into the standard variety, otherwise into your dialect
- At least two translators checked translations and assessed quality
- ullet \Rightarrow **36,384 translation pairs** in eight languages, 18 varieties and seven orthographies

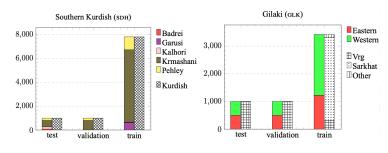




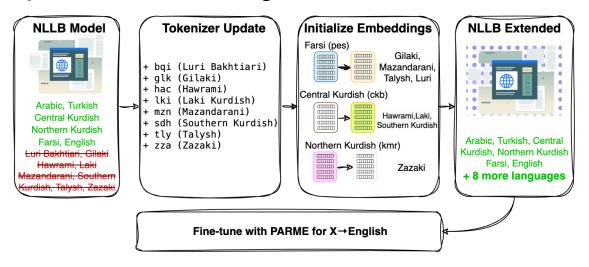
PARME: Evaluation Set

- Create representative test sets for multi-script multi-dialectal translations by
 - 1. Avoiding data contamination
 - 2. Maintaining orthographic consistency
 - 3. Representing all dialects
 - 4. Prioritizing cross-linguality
- ⇒ Approximately 1,000 test instances per language





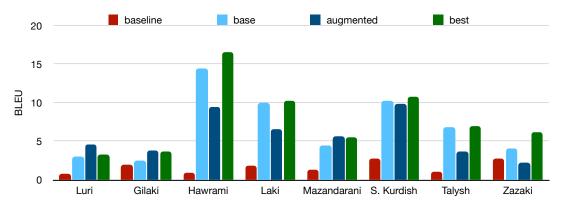
Experiments: NLLB Fine-tuning



^{*} No Language Left Behind covering 200 languages [Team et al., 2024]

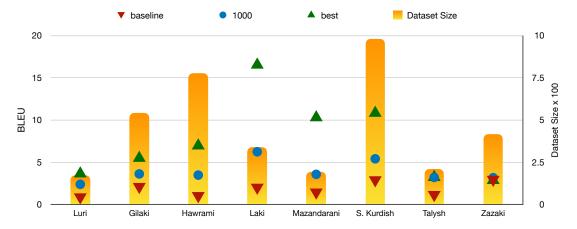
Experiments: Results

- Baseline (zero-shot) performance is poor
- Fine-tuning on PARME X-EN → substantial improvements
- Augmented fine-tuning on X-EN & X-(Lang1 \rightarrow EN) \rightarrow mixed performance
- Augmentation hurts languages initialized from CKB while beneficial for PES & KMR
- Best performance → fine-tuning on PARME X-EN for longer



Experiments: Data Size Impact

- Are performance improvements consistently proportional to data quantity?
- $\bullet \ \, \Rightarrow$ Data quality and hyperparameters matter as much as quantity



Conclusion

- First parallel corpora for 8 severely under-resourced ME languages (18-23M speakers)
- Community-driven success: Native speakers volunteering in participatory research
- Substantial MT improvements but much room for improvement
- Reproducible framework
- Opening new research avenues in NLP for ME languages
- Language preservation: Digital resources can enhance language prestige and vitality

"Technology as a pathway to linguistic diversity preservation and digital inclusion"

Thank You!

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Contact: sina.ahmadi@uzh.ch

Resources: https://github.com/DOLMA-NLP/PARME

Questions?



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