



Vuyko Mistral

Adapting LLMs for Low-Resource Dialectal Translation

Roman Kyslyi¹, Yuliia Maksymiuk², Ihor Pysmennyi¹

¹National Technical University of Ukraine "Igor Sikorsky Kyiv Polytechnic Institute"


²Ukrainian Catholic University





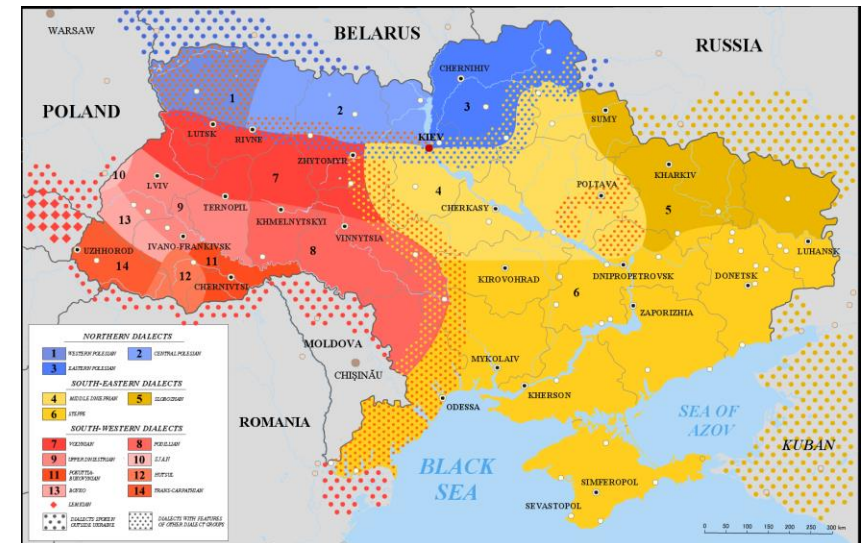
Mistral:

Ой, єй, виджу, шо си зібрало багато людів на конференцію! Та й файно!



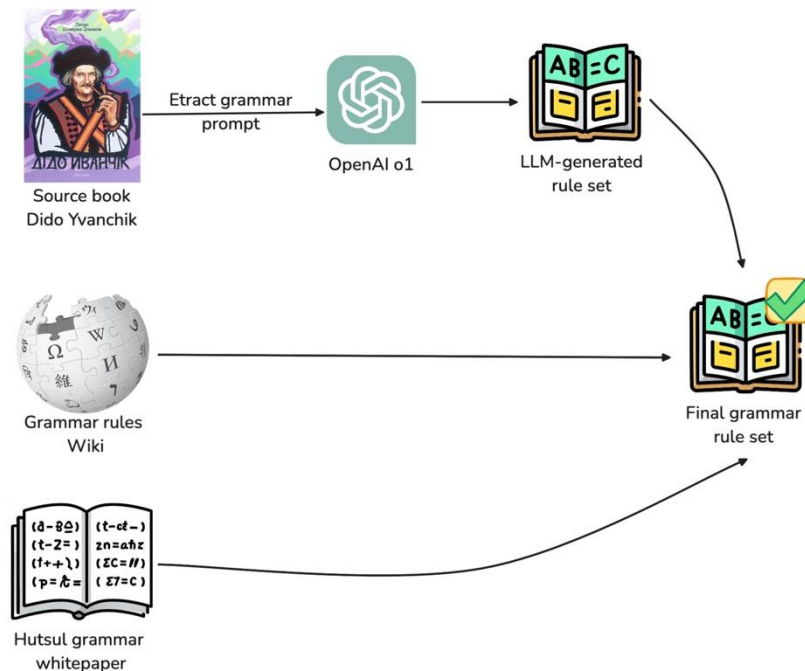
Goal of study

- Most research is centered on high-resource languages and their standard variants underrepresenting dialects that lack sufficient textual resources and standardized orthographies
- To be preserved they need to be visible to modern LLM toolkit and research, which is challenging due to lack of resources



Hutsul dialect

- We've chosen Hutsul dialect as it is still used in Carpathians (21000 as per 2001 census)
- It has several written sources which can be used for training
- We used hybrid approach to construct grammar rule set



Linguistic characteristics of dialect:

- *Phonetics*: vowel transformations, such as changing vowels "є" instead of "а", "я"(ya) (example: "як" → "єк", "ягода" → "єгода" ("yak" → "yek", "yahoda" → "yehoda")).
- *Morphology*: unique case endings (-єдь, -сі) ('-yed', '-si') and preserved dual forms дві яблуці ("two apples", with dual form "yablutsi" instead of plural "yabluka").
- *Lexicon*: Romanian, Polish and German borrowings such as "бринза" (cheese) and "шпацірувати" (go for a walk).



Hutsul dialect

Reference (Hutsul): "Прoшумавси у вечер, ек зробивси в діда в обох хатах гармідер."
(Eng: *He came to his senses in the evening, after a mess started in both of the grandfather's houses.*)

GPT-4o: "Отетавсі аж увечеру, ек зчинивсі в гїда в обидвох осєдочьках гармідер."
(Eng: *He snapped out of it only in the evening, after a chaos started in both of the old man's*

Mistral (manual + synthetic): "Отямивси аж увечері, коли зчинився в діда в обох хатах гармідер."
(Eng: *He came to his senses in the evening, when a chaos broke out in both grandfather's houses.*)

BLEU: 31.46, chrF++: 46.62, TER: 54.55, Fluency: 4, Adequacy: 5, Dialect: 3

Mistral (manual only): "Отємивси аж увечері, ек зчєнивси в діда в обох хатах гармідер."
(Eng: *He came to his senses only in the evening, after stirring up a mess in both of the grandfather's houses.*)

BLEU: 34.39, chrF++: 56.92, TER: 45.45, Fluency: 4, Adequacy: 5, Dialect: 4

LLAMA (manual + synthetic): "Отямивси аж у вечере, коли зчинивси в діда в обох хатях гармідер."
(Eng: *He came to his senses in the evening, when a mess started in both grandfather's houses.*)

BLEU: 20.45, chrF++: 46.42, TER: 63.64, Fluency: 4, Adequacy: 4, Dialect: 3

Fun Examples

Source (Ukr.):

Иванчік сказав: “Час і мені йти на толоку”

Model output:

Иванчік **вікажу**: “Чес і міні иди на толоку”

Real Hutsul:

Иванчік сказав: “Чес и міні йти на толоку”

Source (Ukr.):

Та й так над тим буком у руках «бога» й «чорта»

Model output:

Та й так над тим буком у руках «**біга**» й «**мольфа**»

Real Hutsul:

Тай так mocno над тим буком у руках «бога» й «чорта»

Source (Ukr.):

Аж кришив її: З золотого моря
Вилетіла золота золотариця

Model output:

Аж кришив її: «З золотого моря
Валентина золота золотариця

Real Hutsul:

Аж кришив її: «З золотого моря
Вилетіла золота золотариця





Our contribution

First digital resources for the Hutsul dialect

- a 9 852-pair Hutsul ↔ Standard-Ukrainian parallel corpus and
- a 7 320-entry dialect-to-standard dictionary, compiled from literature, folklore and blogs

Data-scarcity solution via advanced RAG augmentation – we designed a rule-guided RAG pipeline that injects phonological and morphological rules into GPT-4o prompts, creating **52 142 high-quality synthetic sentence pairs** further filtered with automatic alignment checks

Parameter-efficient adaptation of compact open LLMs – two open-source models (Mistral-7B-Instruct v0.3 and LLaMA-3.1 8B) are fine-tuned with LoRA/QLoRA on the combined manual + synthetic data, making dialect translation feasible on a single consumer GPU

Comprehensive, dialect-aware evaluation framework – performance is judged with BLEU, chrF++, TER and a GPT-4o “LLM-judge” scoring Fluency, Adequacy and Dialectal Quality, mitigating the blind spots of standard n-gram metrics for non-standard orthographies

.

Empirical finding: small fine-tuned models beat GPT-4o – the best 7 B model surpasses zero-shot GPT-4o across all automatic metrics and in LLM-based human-like scores, demonstrating the value of dialect-specific tuning even with modest model sizes

Open-source commitment – all code, data, prompts and trained LoRA weights are released on GitHub to encourage further work on Ukrainian dialects and other low-resource varieties

Data collection

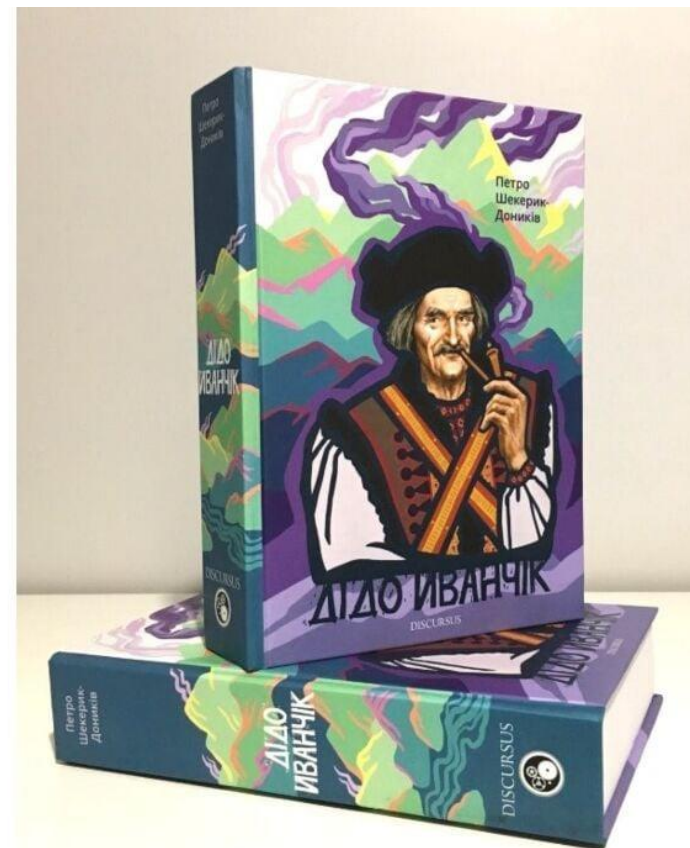
9 852 manually aligned sentence pairs

- Hutsul originals from Дідо Иванчік (foundational dialect novel), ethnographic transcripts, folklore websites and dialect blogs
- Standard-Ukrainian references taken from a modern bilingual edition or hand-translated

7 320-entry Hutsul ↔ Ukrainian dictionary

- Scraped Дідо Иванчік + five open-access lexicons (“Hutsul Hovir”, “Dictionary of Hutsul Words”, etc.)

(!) Lexicon is biased toward the vocabulary found in folk-lore, thus lacks diversity in news, science, or politics.



Synthetic data generation pipeline

Grammar Rule Extraction

- Prompt GPT-4o to distill phonological, morphological, and syntactic rules unique to the Hutsul dialect.
- Output: structured rule set → reusable prompt template.

RAG Index Creation

- Embed every sentence of “Дідо Иванчик” with text-embedding-3-large and store in a vector index to act as an authentic dialect reference base.

Candidate Retrieval

- Sample Standard-Ukrainian lines from the UberText corpus; for each, retrieve the top-3 semantically closest Hutsul sentences from the index.

Prompt Assembly

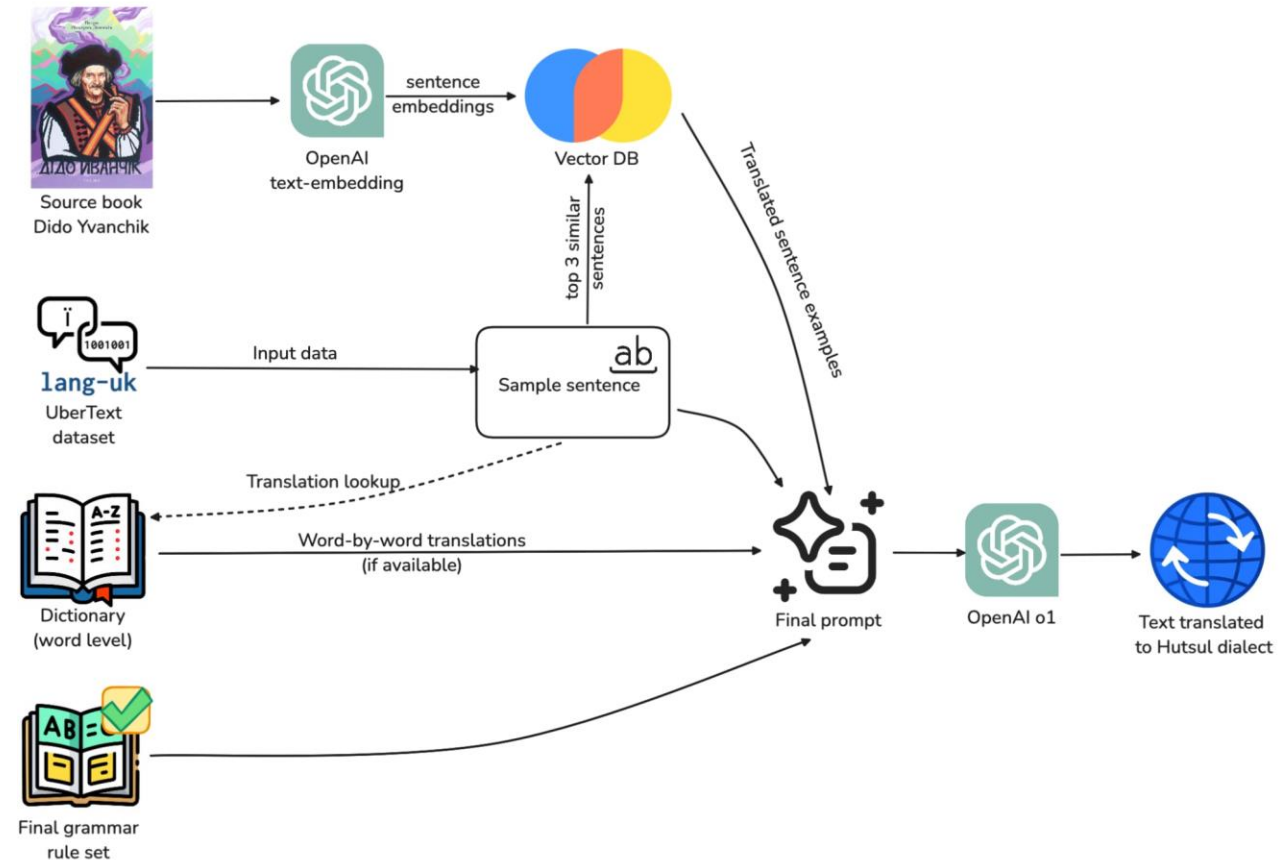
- extracted grammar rules
- three retrieved Hutsul examples
- source Standard-Ukrainian sentence

⇒ one rich prompt ready for generation.

Dialect Generation & Filtering

- Instruct GPT-4o to translate the source into Hutsul, adhering to the rules and style cues. Post-process with alignment/character-ratio filters to keep only high-quality synthetic pairs.

Result: 52 k clean Standard ↔ Hutsul sentence pairs that enlarge the training corpus times





Alignment Quality Metrics

Three metrics to check if sentence pairs align well:

1.U-src (Unaligned source)

- Proportion of source language characters that cannot be aligned to target
- **Lower values = better alignment quality**

2.U-tgt (Unaligned target)

- Proportion of target language characters that cannot be aligned to source
- **Lower values = better alignment quality**

3.X (Crossing alignments)

- Proportion of word alignment pairs that cross/swap positions
- **Shows structural differences** between source and target

Calculated using automatic word alignment tools

Metric	Original Dataset	Synthetic (Raw)	Synthetic (Filtered)
U-src	0.260	0.139	0.005
U-tgt	0.265	0.136	0.005
X	0.022	0.033	0.019

Table 1: Alignment quality metrics comparison between the original dataset, raw synthetic dataset, and synthetic dataset after alignment-based filtering.

Model fine-tuning and evaluation

- **Mistral-7B-Instruct v0.3** – Chosen for its performance-to-size ratio. It outperforms some larger models on many benchmarks, supports multilingual instructions, and includes explicit support for Ukrainian.
- **LLaMA-3.1 8B Instruct** – The instruction-tuned version of LLaMA 3.1 8B. This model has a strong multilingual support and improved instruction-following ability, making it a good candidate for low-resource translation.
- Each model was trained for 3 epochs using LoRa on two dataset variants:
 - a manually created Hutsul–Ukrainian parallel corpus
 - an extended version that included combined manual and filtered synthetic data.

Model	BLEU	chrF++	TER	Fluency	Adequacy	Dialect
GPT-4o	56.64	65.90	34.34	3.76	4.30	3.22
LLaMA (manual annotated + synthetic)	69.02	74.92	22.90	4.11	4.72	3.33
LLaMA (manual annotated only)	59.98	72.61	28.62	4.13	4.72	3.38
Mistral (manual annotated only)	62.36	75.65	28.62	4.14	4.74	3.35
Mistral (manual annotated + synthetic)	74.35	81.89	22.90	4.18	4.72	3.60

Table 2: Automatic and LLM-based evaluation results. BLEU, chrF++, and TER are computed with sacreBLEU. Fluency, adequacy, and dialect quality are rated by GPT-4o (1–5 scale).



Conclusions and future work

- We have created comprehensive and high-quality Hutsul ↔ Standard-Ukrainian parallel corpus as well as
- Novel method of generating synthetic dataset for low-resource dialects was developed
- Proof-of-concept: small open LLMs can outperform GPT-4o when dialect-tuned
- What's next? **Enrichment!**



Гєкүю за увагу! (Thanks!)

