### TRANSLATION

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### TEAM MEMBERS

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### PROJECT OVERVIEW

This project implements a bidirectional Arabic-English neural machine translation system using transformer-based models from the MarianMT framework. The system provides high-quality translations in both directions while being computationally efficient.

# TECHNICAL ARCHITECTURE

### MODEL SPECIFICATIONS

• BASE MODEL: HELSINKI-NLP MARIANMT (TRANSFORMER ARCHITECTURE)

• ARABIC-TO-ENGLISH MODEL:

LAYERS: 6 ENCODER, 6 DECODER

HIDDEN SIZE: 512

**ATTENTION HEADS: 8** 

PARAMETERS: ~77 MILLION

FINETUNED ON CUSTOM DATASET

• ENGLISH-TO-ARABIC MODEL:

PRETRAINED HELSINKI-NLP MODEL

SAME ARCHITECTURE AS ABOVE

**USED WITHOUT FINE-TUNING** 

### TRAINING DETAILS

• FRAMEWORK: HUGGINGFACE TRANSFORMERS

• TRAINING ARGS:

BATCH SIZE: 8 (PER DEVICE)

**LEARNING RATE: 5E-5** 

**EPOCHS: 3** 

MIXED PRECISION (FP16)

• HARDWARE: GPU-ACCELERATED (GOOGLE COLAB)

### DATASET INFORMATION

• Source Data

BILINGUAL SENTENCE PAIRS FROM OPEN-SOURCE PARALLEL CORPUS

CONTAINS 10,000 ARABIC-ENGLISH PAIRS

BALANCED ACROSS DOMAINS (NEWS, CONVERSATIONS, GENERAL TEXT)

• LINK: <u>NMT-WITH-ATTENTION-FOR-AR-TO-EN</u>

### DATASET INFORMATION

• PREPROCESSING PIPELINE

**UNICODE NORMALIZATION** 

**CASE NORMALIZATION** 

**SPECIAL TOKEN HANDLING** 

**PUNCTUATION STANDARDIZATION** 

**TEXT CLEANING REGEX PATTERNS** 

SEQUENCE BOUNDARY MARKING (<START>, <END>)

• DATA SPLITS

**TRAINING: 90% (8,000 PAIRS)** 

TEST: 10% (2,000 PAIRS)

# PERFORMANCE METRICS

### QUANTITATIVE EVALUATION

• BLEU Score: 0.72 (ARABIC→ENGLISH)

**0-1 SCALE WHERE HIGHER IS BETTER** 

**COMPETITIVE WITH BASELINE MODELS** 

**HUMAN TRANSLATION TYPICALLY SCORES 0.6-0.7** 

### QUALITATIVE ASSESSMENT

- HANDLES COMMON PHRASES WELL
- MAINTAINS REASONABLE GRAMMAR
- Preserves most semantic meaning

### SYSTEM LIMITATIONS

• TECHNICAL CONSTRAINTS:

MAX SEQUENCE LENGTH: 128 TOKENS

LIMITED DOMAIN ADAPTATION

SUBOPTIMAL HANDLING OF:

PROPER NOUNS

**IDIOMATIC EXPRESSIONS** 

HIGHLY TECHNICAL TERMS

• LINGUISTIC CHALLENGES:

**ARABIC MORPHOLOGICAL COMPLEXITY** 

DIALECTAL VARIATIONS NOT HANDLED

GENDER/CASE AGREEMENT ISSUES

FORMALITY LEVELS NOT PRESERVED

#### CONCLUSION

In this project, we developed a bidirectional Arabic-English neural machine translation system using MarianMT transformer models from Hugging Face. Through fine-tuning on a curated parallel corpus, we achieved a competitive BLEU score of 0.72 for Arabic-to-English translation while leveraging a pretrained model for English-to-Arabic translation.

This project highlights the potential of open-source NMT models for bridging language barriers. While challenges remain in handling nuances like idiomatic expressions and dialects, the system provides a strong foundation for future improvements. By refining the model and expanding its training data, we can move closer to human-like translation quality in real-world applications.

### REFERENCES

- MarianNMT: <a href="https://marian-nmt.github.io/">https://marian-nmt.github.io/</a>
- Hugging Face Transformers: <a href="https://huggingface.co/docs/transformers/index">https://huggingface.co/docs/transformers/index</a>
- Helsinki-NLP Models: <a href="https://huggingface.co/Helsinki-NLP">https://huggingface.co/Helsinki-NLP</a>
- Dataset Link: <u>nmt-with-attention-for-ar-to-en</u>

## THANK YOU

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