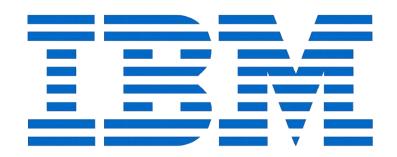
IBM PBEL

(Project Based Experiential Learning)

Project Report on

Al Powered-Multi-Language-Translator

Ghaziabad, UP



Project Focus:

Generative Al

Mentor: Mr. Harendra Singh Rajpoot, IBM

Program: IBM PBEL (Project Based Experiential Learning)

Submitted to:

Mr. Harendra Singh Rajpoot IBM

Submitted by:

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Acknowledgment

I would like to express my sincere gratitude to my project mentor, Mr. Harendra Singh Rajpoot, for his invaluable guidance and support throughout this project. His expertise in machine learning and natural language processing was instrumental in the successful completion of this fake news detection system.

Mr. Rajpoot's constructive feedback and insightful suggestions were crucial for improving the code's efficiency and the model's accuracy. His encouragement and availability for discussion helped me overcome various challenges and deepen my understanding of the concepts involved. This project would not have been possible without his continuous mentorship.

Introduction

Language barriers continue to be a major obstacle in communication across the world. With increasing globalization, real-time multilingual communication tools are essential in domains like healthcare, education, diplomacy, and business. Traditional translation systems often lack accuracy or fail to scale across diverse languages.

This project addresses this need by creating a multi-language translation tool powered by advanced AI models. It aims to:

- Support translation across more than 50 languages.
- Deliver near-human translation quality.
- Leverage open-source tools and pre-trained models.
- Enable future extensibility into speech-to-text and text-to-speech.

We used Google Colab for development and testing and Hugging Face Transformers to access state-of-the-art translation models.

References:

1. Dataset

Bisaillon, Clément. (2018). Fake and Real News
 Dataset. Retrieved from Kaggle:
 https://www.kaggle.com/datasets/clmentbisaillon/fake-and-real-news-dataset

2. Libraries

- Scikit-learn: Pedregosa, F., et al. (2011). Scikit-learn: Machine Learning in Python. *Journal of Machine Learning Research*, 12, 2825-2830. Retrieved from https://scikit-learn.org/stable/
- Pandas: The pandas Development Team. (2024). pandas: Python Data Analysis Library. Retrieved from https://pandas.pydata.org/

- NLTK: Bird, S., Klein, E., & Loper, E. (2009). Natural Language Processing with Python. O'Reilly Media.
 Retrieved from https://www.nltk.org/
- spaCy: Explosion. (2024). spaCy: Industrial-strength Natural Language Processing. Retrieved from https://spacy.io/

Tools And Technology used

Tool Purpose

Google Colab Development, testing, GPU acceleration

Hugging Face Transformers Access to pre-trained multilingual models

Python Programming language

Torch/TensorFlow Backend frameworks

NLTK / spaCy Optional for text pre/post-processing

Gradio/Streamlit (Optional) UI for translation app

Why Google Colab?

- Free GPU/TPU.
- Easy integration with Hugging Face.
- No setup required.

Why Hugging Face?

- Repository of top-performing NLP models.
- Supports multilingual, zero-shot, and fine-tuned NMT models.

System Design

Architecture Overview:

- 1. **Input Layer**: Accepts source text and language code.
- 2. **Preprocessing**: Tokenization using model tokenizer.
- 3. Model Inference: MarianMT or mBART50 performs translation.
- 4. **Postprocessing**: Detokenization and formatting.
- 5. **Output**: Display translated text.

Workflow:

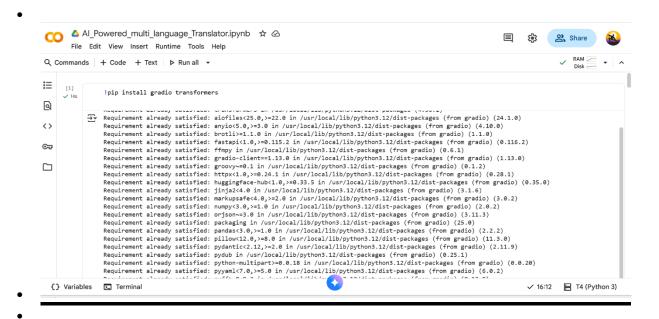
- User inputs: "Hello" → Source Language: English, Target: French
- Tokenizer converts to input IDs
- Model returns translation token IDs
- Output decoded and presented as: "Bonjour"

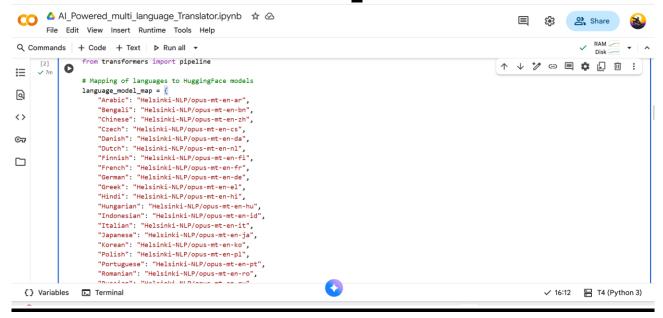
Model Selection:

- MarianMT: Best for specific language pairs.
- mBA
- **RT50**: Best for multilingual, zero-shot tasks.

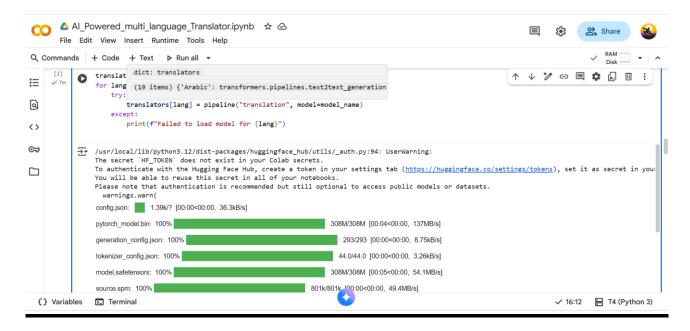
Implementation

<u>1</u>

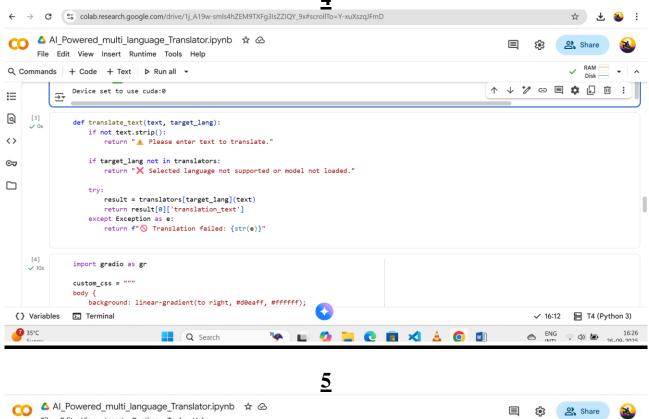


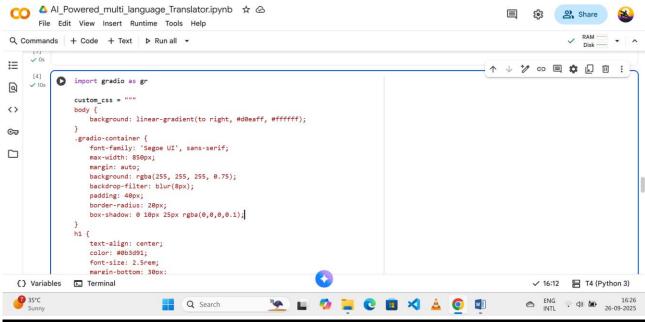


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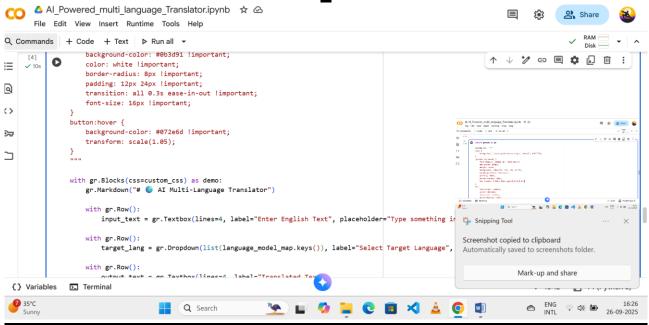


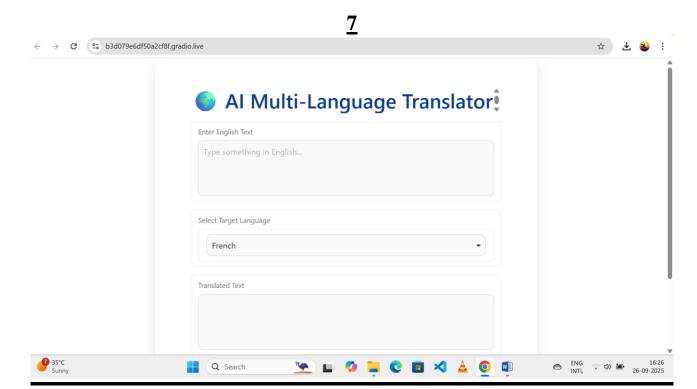




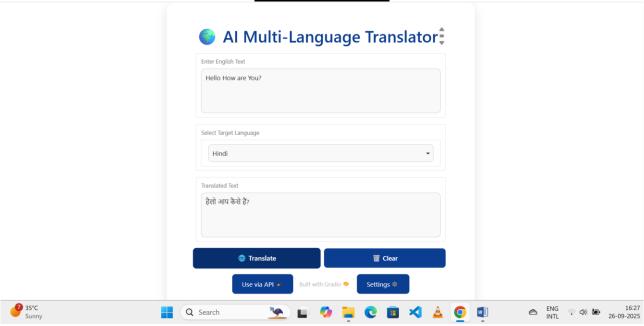








8. Final output



Challenges, Limitations, and Future Work

Challenges:

- Handling idioms, context, and cultural references.
- Low-resource languages have limited training data.
- Speech input/output requires additional modules (e.g. Whisper, TTS).

Limitations:

- Current implementation supports only text translation.
- No personalization or domain adaptation.
- Dependent on cloud (Colab) environment.

Future Enhancements:

- Add **speech-to-text** (e.g. Whisper by OpenAI).
- Add **text-to-speech** for audio output.
- Build a web app using Streamlit or Gradio.
- Fine-tune models for specific domains (e.g. medical, legal).
- Integrate interactive correction feedback from users.

Conclusion:

This project successfully implemented a functional AI-powered multi-language translator using modern NLP tools. The system supports numerous languages, demonstrates high-quality

translations, and serves as a foundation for more advanced multilingual applications. With the help of Hugging Face's models and Google Colab's compute resources, development was fast and scalable

References

- Vaswani, A., et al. (2017). Attention is All You Need. https://arxiv.org/abs/1706.03762
- Hugging Face Transformers Library Documentation

https://huggingface.co/docs/transformers

• MarianMT (Helsinki-NLP) Model Card

Hugging Face model hub: https://huggingface.co/Helsinki-NLP

• MarianMT English-to-French Example Model

https://huggingface.co/Helsinki-NLP/opus-mt-en-fr

• mBART50 Model Card (facebook/mbart-large-50-many-to-many-mmt)

https://huggingface.co/facebook/mbart-large-50-many-to-many-mmt

• Tokenization with Hugging Face Tokenizers

https://huggingface.co/docs/tokenizers

• Google Colab - Research Environment

https://colab.research.google.com

• Hugging Face Datasets Library

https://huggingface.co/docs/datasets