# TECH ENHANCED AI INTERVIEW LEARNING PLATFORM

## **INTRODUCTION**

This project introduces an advanced machine learning model designed to generate a wide range of interview questions tailored to specific topics based on an individual's resume, all while maintaining a deep level of conversation. Furthermore, we incorporate cutting-edge natural language processing (NLP) algorithms to analyse spoken responses, pinpointing grammatical errors, and providing precise corrections post-interview. Additionally, we employ state-of-the-art speech processing techniques, including Automatic Speech Recognition (ASR), to assess learners' speaking pace, detect variations, and provide timely feedback for improvement. The goal of our model is to boost its adaptability and effectiveness across diverse topics and communication styles.

- RESUME SCANNING and CLASSIFICATION
- QUESTION GENERATION
- ASR and SPEECH ANALYSES
- GRAMMAR CORRECTION(using API)
- EVALUATION(BLEU SCORES)

### **Model Overview**

Our model consists of several components:

 Resume Scanning: It involves scanning the resume and then building the model to classify the resume according to various job profiles.

```
from sklearn.feature_extraction.text import TfidfVectorizer
tfidf = TfidfVectorizer(stop_words='english')
tfidf.fit(df['Resume'])
requredTaxt = tfidf.transform(df['Resume'])
from sklearn.neighbors import KNeighborsClassifier
from sklearn.multiclass import OneVsRestClassifier
from sklearn.metrics import accuracy_score
clf = OneVsRestClassifier(KNeighborsClassifier())
clf.fit(X_train,y_train)
ypred = clf.predict(X test)
```

 Question Generation Module: This involves use of techniques such as t5 transformer encodes the data using the training of model from dataset and then decode the data in the form of question and answer.

### Requirements

```
import torch
import json
from tqdm import tqdm
import torch.nn as nn
from torch.optim import Adam
import nltk
import spacy
import string
import evaluate # Bleu
from torch.utils.data import Dataset, DataLoader, RandomSampler
import pandas as pd
import numpy as np
import transformers
from sklearn.model selection import train test split
import matplotlib.pyplot as plt
from transformers import T5Tokenizer, T5Model,
   T5ForConditionalGeneration, T5TokenizerFast
TOKENIZER = T5TokenizerFast.from pretrained("t5-base")
MODEL = T5ForConditionalGeneration.from pretrained("t5-base",
   return dict=True)
```

```
OPTIMIZER = Adam(MODEL.parameters(), lr=0.00001)
Q_LEN = 256  # Question Length
T_LEN = 32  # Target Length
BATCH_SIZE = 4
DEVICE = "cuda:0"
```

- Automatic Speech Recognition: ASR converts spoken words into written text. It involves pre-processing audio, extracting features, mapping them to phonemes and words, and selecting the most likely sequence of words based on statistical models. The output is the transcribed text.
- Speech Processing Module: Incorporates state-of-the-art speech processing techniques to assess learners' speaking pace and detect variations. It offers real-time feedback for improvement in speaking skills.

```
import speech recognition as sr
import os
r = sr.Recognizer()
r.energy threshold = 10000
with sr.Microphone() as source:
   print('Say something')
   audio = r.listen(source)
   print("Done")
    text = r.recognize google(audio)
    f=open("output.txt","w")
    f.truncate(0)
    f.write(text)
    f.write("\n"
    f.close()
    print(text)
    os.system("say '"+'I think you said, '+text+'!'+"'")
except Exc
```

Evaluation Metrics: Evaluation Metrics assess the model's performance in analysing the responses of the candidate. Key metrics include accuracy, precision, recall, BLEU SCORE, AUC-ROC, mAP, and user feedback. They provide insights into the model's accuracy, ability to identify relevant replies, and overall effectiveness in the task.

### **Dataset Creation**

We curated comprehensive datasets consisting of interview transcripts, spoken responses, and grammatical error annotations. Additionally, we generated **custom datasets by crowdsourcing** interview questions and responses aligned with specific topics. This ensures the diversity and relevance of the generated questions and responses.



# CONTAIN 4 COLUMNS: QUESTIONS, ANSWERS, JOB PROFILE, CONTEXT(for T5 TRANSFORMER)

### **Model Training**

We conducted training of our AI model using the curated datasets. We employed techniques like **transfer learning and fine-tuning** to enhance adaptability and performance across various topics and speaking styles. **Hyperparameter tuning** is performed to optimize model performance. Hyperparameter tuning optimizes model performance by adjusting parameters not learned from data, enhancing effectiveness and generalization.

### **Evaluation Metrics**

We evaluate our model based on various metrics:

 Question Relevance (BLEU Score): Measures the similarity between generated questions and human-generated questions. A higher BLEU score indicates better question relevance.

# Score Calculation in BLEU. Unigram precision $\mathbf{P} = \frac{m}{w_t}$ Brevity penalty $p = \left\{ \begin{array}{l} 1 & \text{if } c > r \\ e^{\left(1 - \frac{r}{c}\right)} & \text{if } c \leq r \end{array} \right.$ BLEU $= p \cdot e^{\sum_{n=1}^{N} \left(\frac{1}{N} * \log Pn\right)}$

- **Error Analysis (F1 Score)**: Evaluates the accuracy of grammatical error detection and correction. A higher F1 score indicates better error analysis performance.
- System Usability (SUS Score): Assesses the usability of our system through user feedback. A higher SUS score indicates better usability.

### **Challenges Faced**

During the project, we encountered several challenge

- Custom Dataset Creation
- . Integrating different pipelines of code together
- . Fine tuning the model to make it more accurate

- . GPU limitations
- . Training the transformers part

### **Limitations & Future Improvements**

Despite our model's advancements, it has certain limitations:

- Limited Domain Coverage: Our model's performance may vary across different domains due to data availability and domain-specific language.
- Speech Recognition Accuracy: The accuracy of speech recognition algorithms may impact the performance of our speech processing module.
- **Scalability**: Scaling our model to handle a large volume of users simultaneously may pose challenges.
- . Training issue

In the future, we aim to address these limitations and improve our model by:

- Expanding the dataset to cover a wider range of topics and speaking styles.
- Enhancing speech recognition accuracy through advanced algorithms and techniques.
- · Implementing scalability measures to support a larger user base.

### Conclusion

Our advanced machine learning model for interview question generation and analysis demonstrates promising results in generating diverse interview questions, analysing spoken responses, and providing feedback for improvement. Despite certain limitations, our model shows potential for real-world applications in interview preparation and language learning. Further research and development can lead to significant advancements in this field.

### Team members involved:

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