AdaptLingo: An Adaptive Speech-to-Speech Conversational Agent Lolumbia University Tailored to Users' English Proficiency Levels

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Language students have varying levels of proficiency.

How can an English conversation partner agent adapt to different student proficiency levels?

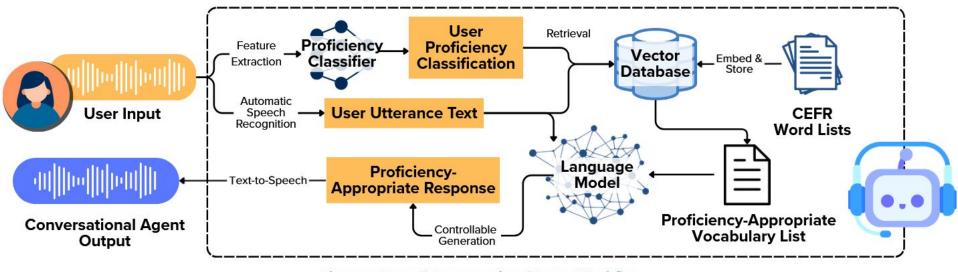


Figure 1. Overall Conversational Agent Workflow.

Step 1: Scan QR Code

Step 2: Click "Visit Site"

Step 3: Scroll down.

Step 4: Use the chatbot!



Proficiency Classifier

Data Collection:

- We sampled data from COREFL [1], the Qatar Corpus [2], the EDEN ASR Dataset [3], and LibriSpeech [4].
- Four native English speakers labeled the perceived proficiency levels according to the Common European Framework of Reference for Languages (CEFR) for 301 audio clips.

LLM-as-a-judge Annotation:

- Used gpt-4o-mini, a large language model (LLM) to label proficiency levels using features extracted from audio clips; validated annotation alignment with humans; and labeled the rest of the dataset.
- Random Forest Classifier trained on the annotated dataset yields
 0.95 Macro F1 score on the test set.

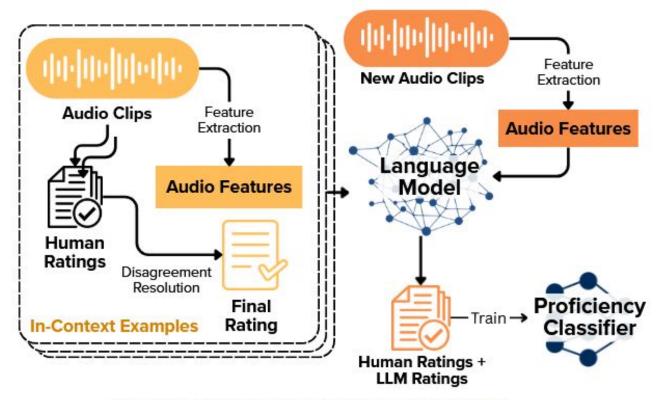


Figure 2. Training the Proficiency Classifier.

Classifier	Test Accuracy	Test Macro FI	Cross- Validation Accuracy Mean	Cross- Validation Macro FI Mean
SVM	0.93	0.92	0.90	0.89
XGBoost	0.95	0.94	0.93	0.92
Random Forest	0.95	0.95	0.92	0.92

Retrieval-augmented Generation (RAG)

- We use the all-mpnet-base-v2 Sentence Transformers model to embed individual words in CEFR vocabulary lists.
- We use Chroma for vector storage and k-nearest-neighbors (KNN) search for retrieval.
- KNN search finds the top k words that have the most similar meaning to the full user's utterance.
- 40 words are retrieved to construct a proficiency-appropriate vocabulary for generation.

Language Model Adaptation

- Upon selecting the retrieved proficiency-appropriate vocabulary that is also relevant to the conversation context:
 - The **logits** corresponding to the vocabulary are boosted, so those words have a higher probability of appearing in the assistant's response.
 - The proficiency-appropriate vocabulary is also passed into the language model's prompt.
 - The language model generates two sentences: the first addressing the user's utterance, and the second asking the user back a relevant question to continue the conversation.
- We leverage the Qwen2.5-14B instruct LLM.



I want to practice English conversation.

We can chat anytime. What topics interest you?





I'm interested in cooking.

Cooking sounds fun. Do you like baking too?





Yes, I like baking too!

Great, baking is enjoyable. What kind of cakes do you bake?





I bake chocolate lava cakes.

Get in Touch

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