## **LLM Fine-Tuning Training Guide for AI Interview Practice Platform**

#### Overview

This document provides detailed guidelines for fine-tuning a Large Language Model (LLM) to power the Al-driven interview practice platform. The goal is to train the LLM to analyze user facial expressions, voice characteristics, and interview responses to generate precise feedback and coaching suggestions.

## 1. Training Objectives

- Detect and interpret facial expressions such as stress, confidence, hesitation, and distractions (e.g., looking away, fidgeting).
- Analyze voice data for breath patterns, speech fluency, pauses, and confidence levels.
- Evaluate answers for relevance, completeness, and communication skills.
- Generate detailed, context-aware feedback reports highlighting strengths and improvement areas.
- Support interactive chatbot conversations to guide users through their reports and practice plans.

## 2. Data Requirements

- \*\*Visual Data:\*\* Annotated video recordings showing facial expressions, eye movements, and hand gestures during interviews.
- \*\*Audio Data:\*\* High-quality speech recordings with annotations for pauses, breath, tone, and confidence markers.
- \*\*Textual Data:\*\* Transcripts of interview sessions including questions, answers, and Al-generated feedback.
- \*\*Behavioral Labels:\*\* Stress levels, confidence ratings, distraction flags, and timing metrics.
- \*\*User Profiles:\*\* Resume data and interview question mappings.

### 3. Features to Track and Model

- \*\*Facial Expression Detection:\*\* Capture micro-expressions, smiles, frowns, eye contact, blinking frequency.

- \*\*Gesture Recognition:\*\* Identify hand movements like fidgeting, pointing, and nervous gestures.
- \*\*Voice Analysis:\*\* Detect speech rate, pauses, breath control, and pitch variation.
- \*\*Answer Evaluation:\*\* Semantic relevance, completeness, hesitation detection, filler words.
- \*\*Timing Metrics:\*\* Time spent per question, delays before answering.

### 4. Output Expectations

- \*\*Per Question Analysis:\*\* Stress/confidence score, fluency score, answer quality rating.
- \*\*Overall Report:\*\* Summary of strengths, weaknesses, time management, and suggestions.
- \*\*Improvement Suggestions:\*\* Personalized coaching tips based on detected patterns.
- \*\*Interactive Dialogue:\*\* Allow chatbot to explain feedback and recommend practice exercises.

## 5. Chatbot Integration

- Design LLM to support two-way conversations with users based on their interview reports.
- The chatbot should be able to answer user queries, clarify feedback, and guide practice routines.
- Maintain user context and learning history to personalize advice.

# 6. Training Strategies

- \*\*Data Augmentation:\*\* Use techniques to diversify facial and voice data (e.g., varied lighting, noise).
- \*\*Multi-Modal Training:\*\* Combine video, audio, and text data for holistic model training.
- \*\*Validation:\*\* Regularly evaluate model accuracy on unseen data and real user sessions.
- \*\*Continuous Learning:\*\* Update model iteratively with new user data and feedback.

## 7. Training Scenarios

### Scenario 1: Basic Facial and Voice Recognition

- Focus on training the model to recognize fundamental expressions and voice features.

- Use labeled datasets with clear expression and speech annotations.

### ### Scenario 2: Advanced Behavioral Analysis

- Train on complex patterns like micro-expressions, subtle stress cues, and speech hesitations.
- Incorporate context awareness to relate expressions with answer content.

#### ### Scenario 3: Real-Time Feedback and Coaching

- Develop model components for real-time signal processing and immediate feedback.
- Integrate with chatbot for dynamic user interaction during practice sessions.

### 8. Technical Details

- Use state-of-the-art deep learning frameworks such as PyTorch or TensorFlow.
- Employ pretrained models for facial expression recognition (e.g., OpenFace, Affectiva).
- Use OpenAI GPT-4.5 or similar LLM for natural language understanding and generation.
- Utilize Gemini API or equivalent for emotion and gesture detection integration.
- Implement voice analysis using specialized audio processing libraries (e.g., librosa, pyAudioAnalysis).

## 9. Summary

This guide ensures that the LLM is trained comprehensively across multiple modalities to provide accurate, contextual, and actionable interview feedback. Combining vision, audio, and NLP components will result in a powerful AI coach that helps users improve their interview performance effectively.

End of LLM Fine-Tuning Training Guide