# Emotion Detection Project Documentation

# Overview

This project involves building an emotion detection web service. The process includes training a machine learning model using the dair-ai/emotion dataset, deploying the model as an inference endpoint using AWS SageMaker, and setting up a web server on an EC2 instance to interact with the SageMaker endpoint.

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# **Development Process**

Here's a breakdown of the steps taken to set up and deploy the project:

## 1. Data Preparation and Model Training

- **Dataset**: The dair-ai/emotion dataset from Hugging Face was used, containing labeled text data for various emotions.
- **Model**: The architecture chosen was DistilBERT, a lightweight version of BERT, suitable for text classification tasks.
- Training Platform: AWS SageMaker was used for training the model.

### o Steps:

- Dataset Preparation: The dataset was uploaded to an S3 bucket.
- Training Job Configuration: A training job was configured in SageMaker with the following settings:
  - Algorithm: Hugging Face's Transformers framework with DistilBERT.
  - **Hyperparameters**: Tuned for learning rate, batch size, and the number of epochs.
  - **Instance Type**: ml.p3.2xlarge for fast training using GPU acceleration.
- Fine-Tuning the Model: The model was fine-tuned on the emotion dataset, achieving a high accuracy on the validation set.

## 2. Setting Up the SageMaker Inference Endpoint

 After training, the model was deployed as an endpoint on AWS SageMaker.

### o Steps:

- Create Endpoint Configuration: The trained model artifacts were used to create an endpoint configuration.
- Deploy the Endpoint: The model was deployed to an ml.m5.large instance for inference.

 Testing the Endpoint: The endpoint was tested using sample inputs to ensure it returned the expected emotion predictions.

### 3. Developing the Web Server

 Platform: A Flask web server was developed to serve as a front-end interface for users to interact with the SageMaker endpoint.

### Configuration:

 The server was configured to receive text input via a POST request, forward the request to the SageMaker endpoint using the boto3 library, and return the predicted emotion and confidence score.

### • Error Handling:

 Implemented error handling to manage cases where the SageMaker endpoint is unreachable or the input is invalid.

# 4. Hosting the Web Server on an EC2 Instance

### EC2 Instance Setup:

- Launched an EC2 instance using an Amazon Linux 2 AMI.
- Configured the security group to allow inbound HTTP and SSH traffic.

### • Environment Setup:

- Installed Python, Git, and other required dependencies on the instance.
- o Cloned the project repository from GitHub.
- Installed the necessary Python packages from requirements.txt.

### • Running the Flask Server:

- Configured the Flask application to listen on all network interfaces (0.0.0.0) to make it accessible externally.
- Used screen or tmux to keep the server running in the background.

### • Testing:

 The server was tested using tools like curl and Postman to ensure that it correctly communicated with the SageMaker endpoint and returned the predicted emotion.