Overview

The code is designed to use a microphone to collect audio data, process it through a pre-trained machine learning model (likely trained on Edge Impulse's platform), and perform actions based on the inference results. It includes configuration for hardware setup, audio data collection, signal processing, model inference, and output handling based on predictions.

Key Components

1. **Edge Impulse Ingestion SDK:** This is a software development kit from Edge Impulse that allows the embedded devices to capture and send data to the Edge Impulse platform for model training or to run inference locally on the device.

2. Memory Management:

- o EIDSP_QUANTIZE_FILTERBANK: A preprocessor directive used to control memory usage related to filter bank quantization, affecting RAM consumption.
- Static and dynamic memory management strategies are suggested to resolve issues related to TensorFlow Lite model arena allocation, useful for constrained environments.

3. Audio Processing with PDM (Pulse Density Modulation) Library:

- Audio data is captured using the PDM library, which is suitable for processing high-quality audio signals with microcontrollers.
- o The microphone_inference_start function configures the microphone settings, including the buffer size and sampling rate (16 kHz in mono mode).
- o A callback function (pdm_data_ready_inference_callback) is used to fill the audio buffer as new data becomes available.

4. Inference Data Structure:

o inference_t struct holds the audio buffer and status flags to manage data readiness and buffer count, facilitating the communication between the data collection callback and the main loop.

5. Model Inference:

- The loop function is the core of the operation where audio recording, signal processing, and running the inference happen.
- o Audio recording is triggered, and once the buffer is filled, it's used to create a signal_t struct that acts as an input to the machine learning model.
- o The Edge Impulse model is invoked using run_classifier, which processes the signal and outputs predictions.
- Based on the prediction results, certain actions (like turning LEDs on/off) are performed.

6. Output Handling:

 Predictions are logged to the serial console, and based on these predictions, different LEDs are controlled. For example, if a certain class's prediction score exceeds a threshold, an LED might flash.

7. Hardware Specifics:

 Pins for LEDs (RED, BLUE, GREEN) are defined, but their control is commented out except for the GREEN LED. This part of the code seems to be used for signaling the results of the inference visibly.

Functional Flow

- **Setup:** Initializes serial communication, configures LED pins, and starts the microphone with specific settings.
- Inferencing Loop:
 - Begins by delaying the start of recording, then records a specified duration of audio.
 - o Performs inference on the recorded audio.
 - Based on the inference results, executes conditional actions (like lighting up an LED).

Conclusion

This project is an embedded system application focusing on audio processing and real-time machine learning inference. It is structured to be modular and configurable, making it suitable for applications needing on-device audio analysis, such as environmental sound classification, voice commands recognition, or anomaly detection in sounds.