

**THE UNIVERSITY OF BAMENDA**

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**DEPARTMENT OF  
COMPUTER  
ENGINEERING**

**REAL-TIME AI MUSIC ACCOMPANIMENT VIA  
PREDICTIVE LATENCY COMPENSATION**

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## **1. BACKGROUND**

Generative Artificial Intelligence has revolutionized audio synthesis, with models like Google's MusicLM and OpenAI's Jukebox demonstrating the ability to generate complex musical compositions. These advancements utilize large-scale deep learning architectures to model long-term musical structure. However, most current research focuses on offline generation—creating music asynchronously after a prompt is received. This latency is acceptable for composition tools but catastrophic for live performance, where musicians rely on millisecond-level interaction loops to maintain synchronization.

## **2. PROBLEM DEFINITION**

The central software engineering challenge in interactive music systems is the Synchronization Gap. In a live setting, the time taken for audio input processing, neural inference, and audio output creates a perceptible delay. Existing systems react after the musician plays, guaranteeing that the accompaniment is always slightly behind the beat. There is a lack of intelligent agents capable of quantifying this system latency and predictively generating audio for a future timestamp, effectively "fast-forwarding" the output to align perfectly with the live performer.

## **3. PROPOSED APPROACH**

This project proposes a Predictive-Adaptive Architecture that treats music generation as a time-critical control system. Instead of minimizing latency, the system anticipates it. The approach consists of three core components:

1. **Dynamic Latency Estimation:** The system will continuously calculate the "Round-Trip Latency" of the hardware pipeline. This variable serves as a dynamic offset for the generation engine.
2. **Predictive Sequence Generation:** The Deep Learning model (optimized LSTM or Transformer) will be trained for Next-Step Anticipation. It analyzes the musician's

current trajectory to generate the musical sequence required for the future moment when the audio will actually be heard.

3. Tempo-Locked Synchronization: The system will generate symbolic music (MIDI) time-shifted by the exact duration of the computed latency. This ensures that despite processing delays, the audible output aligns perfectly with the live transient of the human performer.

#### **4. EXPECTED OUTCOMES**

The primary deliverable is a functional software prototype capable of accompanying a live musician without perceptible lag. Key success metrics include:

- Synchronization Accuracy: The phase difference between the human beat and the AI beat should be negligible ( $< 10\text{ms}$ ), achieved via the predictive compensation algorithm.
- Adaptability: The system dynamically adjusts its "fast-forward" offset if CPU load changes.
- Musical Coherence: The generated accompaniment maintains harmonic consistency with the input melody.

## 5. REFERENCES

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