



Beat Maker Using AI and ML

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Abstract—

This project is about a Beat Generator. It uses Pygame for the front end and TensorFlow with Magenta for the back end. Users can create beat patterns on a grid-like interface, and the AI generates different variations based on what users make. Pygame, TensorFlow, and Magenta work together to make a fun and flexible platform for users to explore and express their musical ideas. There's also a possibility of adding more music styles and collaborative features in the future

In the ever-evolving landscape of music creation, the fusion of Artificial Intelligence (AI) and machine learning has ushered in a new era of innovation for musicians and enthusiasts alike. This project presents a sophisticated Beat Generator, serving as a harmonious amalgamation of musical artistry and advanced technologies. At its core, this platform boasts a Pygame and Pygame.mixer front end, complemented by a robust back end powered by TensorFlow and the Magenta library. This union creates a nexus where user interaction seamlessly intertwines with AI-driven pattern creation.

The essence of the project lies in providing users with a creative platform to effortlessly craft unique and dynamic beat patterns. The Pygame front end offers an intuitive grid-based interface, facilitating real-time interaction with various beats and rhythm elements. The auditory dimension is enriched by the Pygame.mixer module, orchestrating a symphony of diverse audio samples associated with each beat.

The true magic unfolds in the back end, where TensorFlow and the Magenta library take center stage. Through the prowess of machine learning models, the back end meticulously analyzes and comprehends user-provided patterns, subsequently generating beat patterns that are both reminiscent of the user's input and creatively distinct. This AI-driven approach ensures an ongoing, dynamic interplay between user creativity and the machine's capacity for musical exploration.

Looking ahead, the project envisions not just a tool but a comprehensive musical experience. Users are empowered to explore and express their musical ideas in a fun and flexible environment. The horizon holds promises of further enriching this experience by introducing additional music styles and collaborative features. The trajectory of the Beat Generator is one of continual growth, where the synergy of Pygame, TensorFlow, and Magenta opens avenues for endless creative possibilities.

In this symbiotic convergence of musical expression and technological ingenuity, the Beat Generator stands as a testament to the evolving landscape of music composition. It is not merely a tool but a gateway for musicians of all levels to embark on a journey of exploration, creativity, and collaboration. The Beat Generator beckons users to delve into the intricate tapestry of AI-driven musicality, paving the way for a future where the boundaries of musical expression are defined only by the breadth of imagination.

I.INTRODUCTION:

In the field of music creation, the combination of Artificial Intelligence (AI) and machine learning has led to the development of innovative tools for musicians and enthusiasts. This project aims to blend the art of music composition with advanced technologies. The Beat Generator, featuring a Pygame and Pygame.mixer front end, along with a robust back end powered by TensorFlow and the Magenta library, sits at the intersection of user interaction and AI-driven pattern creation.

At its essence, the project serves as a creative platform, allowing users to easily craft unique and dynamic beat patterns. The Pygame front end offers an intuitive interface with a grid-based layout, enabling real-time interaction with different beats and rhythm elements. The visual representation is complemented by the Pygame.mixer module, managing the auditory aspect by playing diverse audio samples associated with each beat.

The real magic happens in the back end, where TensorFlow and the Magenta library play a crucial role. Using machine learning models, the backend analyzes and comprehends user-provided patterns, generating similar yet varied beat patterns. This AI-driven approach ensures a continuous and dynamic interplay between user creativity and machine-generated musical exploration.

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Furthermore, as this visionary project unfolds, it stands not just as a tool for musical creation but as a transformative journey where human ingenuity and artificial intelligence converge. The Beat Generator is not confined to the limitations of traditional composition tools; instead, it beckons users to explore uncharted realms of musical creativity, inviting them into a collaborative dance with the machine. As we delve deeper into the intricacies of this project, we discover not just a sequence of beats but a narrative where the user's musical ideas intertwine harmoniously with the AI-driven orchestrations, ushering in a new era of collaborative and limitless musical possibilities.

II. OBJECTIVES:

Facilitate Musical Creativity:

Provide users with a versatile platform for crafting unique and dynamic beat patterns.
Enable real-time interaction with beats and rhythm elements through an intuitive Pygame-based grid interface.

Utilize Advanced Technologies:

Blend the art of music composition with cutting-edge technologies, specifically Pygame, Pygame.mixer, TensorFlow, and the Magenta library.
Leverage machine learning models in the back end to analyze user-created patterns and generate varied beat patterns.

Enhance Music Production Options:

Assist users in producing a diverse range of beats, expanding their options for musical compositions.
Develop a comprehensive package suitable for both beginners and professionals engaged in music production.

Create a User-Friendly Interface:

Design a user interface that is visually appealing, user-friendly, and accessible to musicians of all skill levels.
Provide a seamless experience for users to visually and audibly experience the ongoing creation of beats.

Integrate TensorFlow and Magenta Efficiently:

Install the necessary dependencies for TensorFlow and Magenta, establishing a virtual environment for efficient project management.
Implement Magenta's pre-trained model architecture for beat generation, extracting pertinent features from musical data.

Collect and Preprocess Dataset:

Collect a diverse dataset comprising drum patterns and beats to train the chosen model.
Preprocess the dataset into MIDI file format, preparing it for effective training and pattern generation.

Train and Adjust Model:

Utilize TensorFlow to partition the dataset into training sets and train the selected model.
Make necessary adjustments to parameters for optimal model performance and pattern generation.

Enable User Customization:

Create a user interface that facilitates interaction with the beat generation model. Empower users to customize beats according to their preferences, fostering creativity and exploration.

In summary, this project aims to harmonize user creativity with AI-driven music composition, providing a comprehensive and accessible tool for music producers while laying the groundwork for future expansions in music styles and collaborative features.

This project aids users in producing a diverse range of beats, expanding their options and enhancing their selection of musical compositions

To accomplish this, a front end will be developed for the user, allowing them to create their beats while visually and audibly experiencing the ongoing creation. The backend component will be incorporated to assist in delivering comparable beat patterns to the user.

This project will serve as a comprehensive package suitable for both beginners and professionals engaged in music production.

III. LITERATURE REVIEW

Evolution of Music and AI Technology [1]

Dillon Ranwala explores the evolving intersection of music and AI technology. The article discusses the historical context and the impact of AI on music creation, providing a backdrop for understanding the significance of AI in the proposed Beat Generator project.

LSTM Neural Networks in Music Generation [2]

Sigurður Skúli's article delves into the use of Long Short-Term Memory (LSTM) neural networks in music generation. Understanding LSTM networks is crucial for comprehending the neural network approach used in the project's back end. The article provides insights into how these networks can capture long-term dependencies, which is relevant for generating coherent and dynamic beat patterns.

Recurrent Neural Networks for Music Composition [3]

Belousov, Phuycharoen, and Milosevic's work on recurrent neural networks (RNNs) for music composition sheds light on how these networks can be employed to create music. Since the proposed methodology involves the use of Magenta, which

utilizes RNNs, understanding the principles from this work is fundamental.

Deep Learning Techniques for Music Generation [4] The book by Briot, Hadjeres, and Pachet provides a comprehensive overview of deep learning techniques specifically applied to music generation. This resource is valuable for gaining a broader understanding of the various models and methodologies in the domain of AI-driven music creation.

Subjective Evaluation of Deep Learning Models for Symbolic Music Composition [5]

Hernandez-Olivan, Puyuelo, and Beltran conduct a subjective evaluation of deep learning models for symbolic music composition. This research can provide insights into the potential challenges and considerations when evaluating the effectiveness of AI-generated music, guiding the assessment of the proposed Beat Generator's output

IV. PROPOSED METHODOLOGY

Here, we outline the proposed methodology for our project:

Step 1: Install the dependencies for Tensorflow and Magenta, and establish a virtual environment to efficiently manage the project's required dependencies.

Step 2: Collect a varied dataset comprising drum patterns and beats, then preprocess the dataset into MIDI file format to prepare it for training.

Step 3: Utilize Magenta's pre-trained model architecture for beat generation. Encode the musical data for the model and extract pertinent features like note duration, velocity, etc.

Step 4: Partition the dataset into training sets and proceed to train the chosen model with TensorFlow, making necessary adjustments to parameters as required.

Step 5: Create a user interface that facilitates interaction with the beat generation model, enabling users to customize the beat according to their preferences.

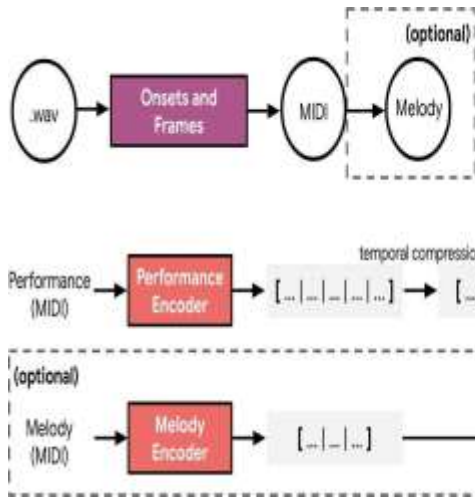


Fig.1: Proposed Approach

IV. EXPECTED OUTCOMES

Diverse Beat Patterns: The project aims to deliver a tool capable of generating a diverse range of beat patterns. The AI-driven backend, leveraging TensorFlow and Magenta, is expected to produce variations that align with user inputs, thereby enhancing the creative possibilities for music producers.

User-Friendly Interface: The front end, built with Pygame, is expected to offer an intuitive and visually appealing interface. Users, whether beginners or professionals, should find it easy to interact with the grid-like layout, allowing for real-time creation and customization of beats.

Flexibility and Customization: Users will have the ability to customize beats according to their preferences. The Beat Generator should provide a platform for users to experiment with different styles and elements, fostering a sense of creativity and exploration.

Integration of TensorFlow and Magenta: The successful integration of TensorFlow and Magenta into the project's architecture is crucial. The expected outcome includes a well-trained model capable of understanding and generating musical patterns that align with the input dataset and user interactions.

Potential for Future Expansion: The project sets the stage for future enhancements, such as the addition of more music styles and collaborative features. The modular design should facilitate the integration of new functionalities, ensuring the longevity and adaptability of the Beat Generator.

In summary, the expected outcome is a user-friendly, flexible, and AI-driven Beat Generator that not only meets the immediate goal of diverse beat pattern generation but also lays the foundation for future advancements in AI-based music composition and collaboration.

The expected outcome of the Beat Generator project encompasses both technical achievements and user-oriented benefits:

V. REFERENCES

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