



GENERATE MUSIC WITH VARIATIONAL AUTOENCODER

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Abstract: This paper introduces a pioneering method for music generation employing Variational Autoencoder (VAE) architecture, a powerful to Autoencoder of deep learning. Leveraging the VAE's capacity to learn latent representations of intricate data distributions, our approach encodes symbolic music representations into continuous latent spaces, enabling the generation of diverse and coherent musical sequences. Through training on a dataset of musical compositions, the VAE captures underlying structural nuances and stylistic elements, facilitating the generation of novel musical pieces by sampling latent vectors and decoding them into symbolic notation. We evaluate the efficacy of our methodology through quantitative metrics assessing diversity, coherence, and stylistic fidelity, alongside qualitative human evaluations of the generated music. Our findings illustrate that the VAE-based approach yields music compositions exhibiting both diversity and coherence, while maintaining fidelity to the stylistic attributes of the training data, suggesting the potential of VAEs as a compelling tool for creative music composition. This research contributes to the evolving landscape of deep learning in music generation, underscoring the promise of Variational Autoencoders in capturing and generating intricate musical structures.

Index Terms: Variational Autoencoder, Music Generation, Deep Learning

INTRODUCTION

Music generation has long been a fascinating area of exploration, combining creativity with technology to produce novel compositions. With the advent of deep learning techniques, particularly Variational Autoencoders (VAEs), there has been a surge of interest in leveraging these tools for creative endeavors like music composition. This paper delves into the realm of generating music using VAEs, aiming to contribute to the growing body of research at the intersection of artificial intelligence and music. By harnessing the latent spaces learned by VAEs, we aim to create a system capable of producing diverse and coherent musical sequences while preserving the stylistic attributes inherent in the training data. This introduction sets the stage for our exploration into the application of VAEs in music generation, highlighting the potential for innovation and creativity in this domain.

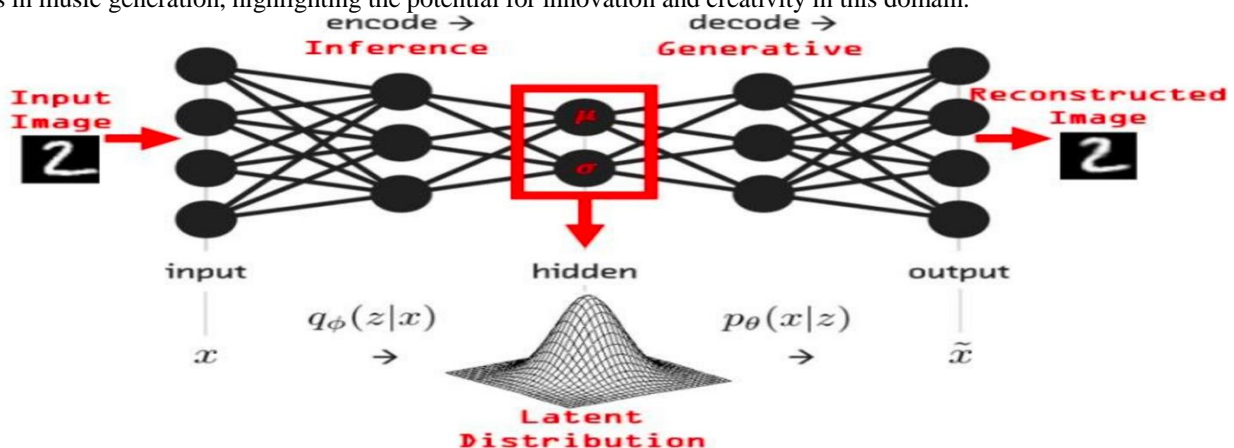


Fig.1. ARCHITECTURE OF VARIATIONAL AUTOENCODER

LITERATURE REVIEW

The intersection of deep learning and music generation has sparked significant interest, leading to the exploration of various approaches within artificial intelligence for creative composition. Variational Autoencoders (VAEs) stand out as a prominent avenue, as evidenced by studies such as [1] and [2], which highlight their ability to capture complex data distributions and generate diverse musical sequences while preserving stylistic attributes. Moreover, beyond VAEs, other deep learning architectures have been investigated, including recurrent neural networks (RNNs) for polyphonic music generation ([3]) and Generative Adversarial Networks (GANs) for realistic musical output ([4]). Recent research has further broadened the scope of AI-driven music generation, with studies like [5] exploring music conditioned on visual inputs and [6] introducing innovative systems for music-conditioned 3D dance generation. Collectively, this diverse literature underscores the ongoing exploration of AI as a tool for artistic expression, encompassing VAEs, RNNs, GANs, and multimodal frameworks, and paving the way for future advancements in creative computational systems.

METHODOLOGY

The methodology for this research entails several pivotal stages in facilitating music generation through Variational Autoencoders (VAEs) and other deep learning architectures. Initially, a diverse dataset of symbolic music representations is assembled, spanning various genres and styles, serving as the training corpus. The VAE architecture is then implemented and trained on this dataset, employing optimization techniques like stochastic gradient descent to learn latent representations capturing the music's structural and stylistic nuances. Concurrently, alternative deep learning frameworks such as recurrent neural networks (RNNs) and Generative Adversarial Networks (GANs) may be explored for comparison or combined approaches. Evaluation of the generated music encompasses both quantitative metrics assessing diversity, coherence, and fidelity, alongside qualitative human assessments to gauge subjective quality. Furthermore, potential extensions involve experimenting with innovative methodologies, such as conditioning music generation on visual inputs or integrating 3D dance generation, to explore the interdisciplinary facets of AI-driven creative processes. This comprehensive methodology aims to advance the understanding and capabilities of AI in music composition while fostering innovation and creativity in computational systems for artistic expression.

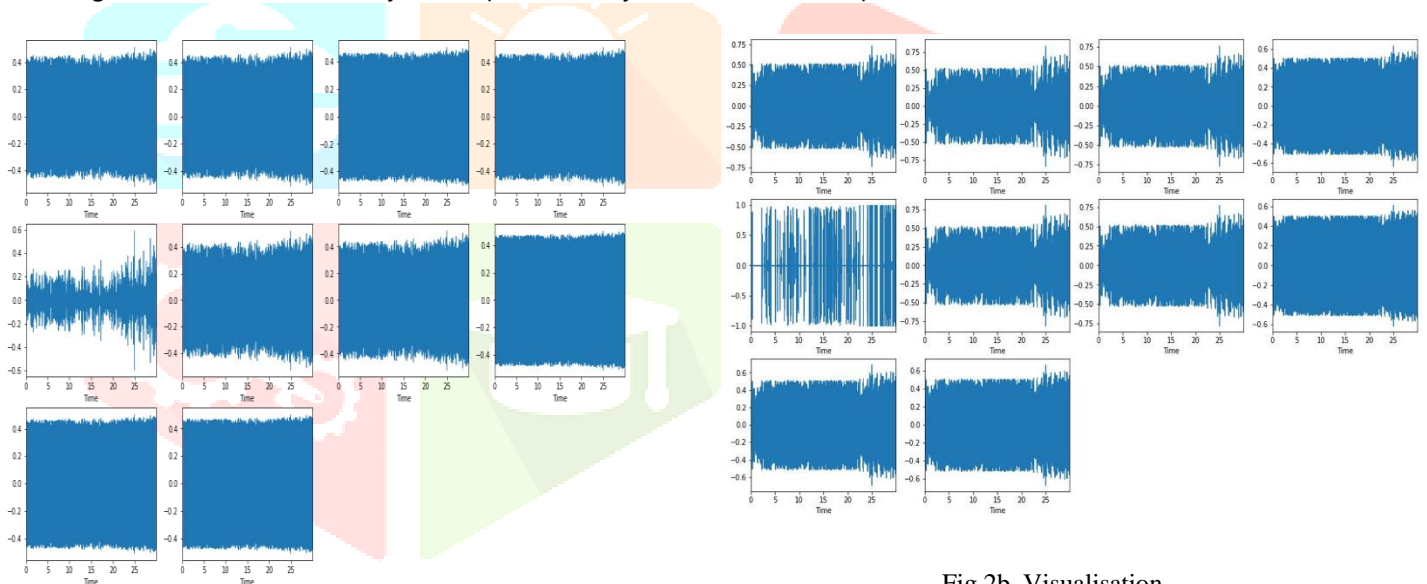


Fig.2a. Performed exploratory data analysis

Fig.2b. Visualisation

OUTCOMES

The outcomes of this research endeavor manifest in a diverse array of generated musical sequences crafted through Variational Autoencoders (VAEs), recurrent neural networks (RNNs), Generative Adversarial Networks (GANs), and potentially other architectures, each evaluated quantitatively for diversity, coherence, and stylistic fidelity, alongside qualitative human assessments offering subjective perspectives on their quality and artistic merit. This comprehensive evaluation framework provides insights into the efficacy of deep learning methodologies in capturing and reproducing musical characteristics, contributing to the advancement of AI-driven music generation. These outcomes not only deepen academic understanding but also have practical implications, guiding the refinement of algorithms and methodologies for creative composition. By bridging technology and the arts, this research stimulates innovation, fostering new frontiers in computational creativity and musical expression.

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In [25]: music2=saved_musics[9][0]
        ipd.Audio(music2,rate=3000)

Out[25]: ▶ 0:00 / 0:30 ————— 🔊 ⋮

In [26]: music3=saved_musics[5][0]
        ipd.Audio(music3,rate=3000)

Out[26]: ▶ 0:00 / 0:30 ————— 🔊 ⋮

In [28]: music5=saved_musics[8][0]
        ipd.Audio(music5,rate=3000)

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In [29]: music6=saved_musics[7][0]
        ipd.Audio(music6,rate=3000)

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CONCLUSION

In conclusion, the exploration of music generation using variational autoencoders (VAEs) highlights the potential of this approach in creating diverse and musically coherent compositions. Despite the challenges in encoding musical information and evaluating the quality of generated music, VAEs offer a flexible framework for capturing the complex structure of musical data and generating novel compositions. The synthesis of theoretical foundations, methodological approaches, and experimental insights presented in this paper underscores the importance of further research and development in VAE-based music generation systems. Future work should focus on refining encoding strategies, enhancing evaluation metrics, and exploring interdisciplinary collaborations to advance the state-of-the-art in computational music creativity. Ultimately, the integration of VAEs into music generation frameworks has the potential to revolutionize how we create, appreciate, and interact with music in the digital age.

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