

A Survey on Face Generation Using Deep Learning

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ABSTRACT

The field of deep learning has witnessed remarkable advancements in the generation of realistic and high-quality human faces. This abstract provides an overview of the techniques and methodologies employed in face generation using deep learning models. Face generation using deep learning primarily relies on Generative Adversarial Networks (GANs) and Variational Autoencoders (VAEs). These models have demonstrated the ability to synthesize lifelike faces, presenting a wide range of applications in computer vision, entertainment, and beyond.

Keywords — Deep learning frameworks (e.g., TensorFlow, PyTorch), Image-to-image translation, Image synthesis. Unsupervised learning.

I. INTRODUCTION

Is an exciting and rapidly advancing field within artificial intelligence (AI) and computer vision. It revolves around the creation of realistic and high-quality human faces from scratch, often indistinguishable from actual photographs of real individuals. This technology has a wide range of applications, including in entertainment, virtual reality, facial recognition, and even addressing privacy concerns.

Deep learning, particularly Generative Adversarial Networks (GANs) and Variational Auto encoders (VAEs), has played a pivotal role in revolutionizing the way we generate faces. These neural network architectures have the ability to learn complex patterns and features from large datasets of real faces, allowing them to generate new and diverse faces that closely resemble human subjects.

The field of computer vision and deep learning has witnessed remarkable advancements, particularly in the realm of image-to-image translation. Image-to-image translation refers to the process of converting an input image from one domain to another, while preserving the essential characteristics and details. One of the most intriguing applications of this technology is in the generation of highly realistic and detailed human faces, an area known as face generation.

Face generation using deep learning has gained widespread attention due to its potential applications in various domains, including entertainment, biometrics, and virtual reality. This technology enables the synthesis of

entirely new faces, which can be indistinguishable from real human faces, or the transformation of existing faces into different expressions, ages, or styles.



Learning GAN for Pose-Invariant Face Recognition

By amalgamating vast datasets of facial images, these networks have the capability to generate entirely new and diverse faces with a remarkable level of detail, including facial expressions, age, gender, and ethnicity. The applications of this technology are far-reaching, spanning from computer graphics, gaming, and entertainment to biometrics, facial recognition, and even medical research, illustrating its potential to revolutionize the way we perceive and utilize artificial intelligence in the realm of visual representation and human-computer interaction.

II. LITERATURE SURVEY

Due to the huge availability of data, it is difficult to classify/process images at a higher speed and accuracy. The first technique was in the field of computer vision and it used image data for face recognition and detection of an object from the image but later Convolutional Neural Networks (CNNs) took place. It uses a Vanilla Gans and Conditional GAN. The proposed solution seeks to generate artificial images using Deep Convolutional Generative Adversarial Networks (DCGAN). Having performed well with unlabeled (unsupervised) data in the past, DCGANs make a perfect option to any scenario.

“Face Recognition using Deep Learning” technique proposed by Banumalar Koodalsamy, Manikandan Bairavan Veerayan, and Vanaja Narayanasamy in year 2023. The problem solved in that technique is to Identifying a person primarily relies on their facial features, which even distinguish identical twins. As a result, facial recognition and identification become crucial for distinguishing individuals. Also, Face recognition using deep learning is a powerful technology that has made great strides in recent years. Future work in this area could be to automatizes the attendance system in schools and colleges.

“3D Face Recognition in Deep Learning Era: A Surve”. Proposed by Sahil Sharmal, Vijay Kumar inn year 2022. This paper provides an in depth analysis of 3D face reconstruction using deep learning techniques. The performance analysis of different face reconstruction techniques has discussed in terms of software, hardware,

pros and cons. Also it help to Most of the researchers have worked on RGBD images. With deep learning, working on a mesh image or the voxel image has hardware constraints.

“Text-to-Face Generation via Attribute Disentanglement “ Proposed by Prof. Tianren Wang, Teng Zhang, Brian Lovell in year 2021. In this paper consists s Text-to-face model that not only produces images in high resolution (1024X1024) with text-to-image consistency, but also outputs multiple diverse faces to cover a wide range of unspecified facial features in a natural way. This uses a GAN model which includes a novel TTF frame work for high quality image- generation More specifically, it propose a pre-trained BERT(Bidirectional Encoder Representations from Transformers) multi-label model for natural language processing.

“International Journal of Information Management Data Insights” Proposed by AlankritaAggarwal, Mamta Mittal, GopiBattineni in year 2021.To generate a wide range of data types, including images, music and throught deep learning associated Generated Adversarial Network (GAN). GANs have so far shown very impressive results on tasks that were difficult to perform using conventional methods. Transforming low-resolution images, for example, was previously quite a challenging task and was generally carried out using CNNs.GAN architectures, such as SRGANs or pix to pix, have shown the potential of GANs for this application.

“Review of deep learning: concepts, CNN architectures, challenges, applications, future direction” Proposed by LaithAlzubaidi 1, Jinglan Zhang1, Amjad J. Humaidi, Ayad Al-Dujaili, Ye Duan, Omran Al-Shamma, J. Santamaria, Mohammed A. Fadhel, Muthana Al-Amidie and LaithFarhan in year 2021. Recently machine learning (ML) has become very widespread in research and has been in incorporated in a variety of applications, including text mining, spam detection, video recommendation, image classification, and multimedia concept retrieval. Among the different ML algorithms, deep learning (DL) is very commonly employed in these applications.

“Thermal Face Generation Using Style GAN” Proposed by Gabriel Hermosilla, Diego Ignacio Henriquez Tapia, Hector Allende, Gonzalo Farias Castro and Esteban Vera in year 2021. The technique is used to Generative adversarial networks, StyleGAN2, thermal face recognition, deep learning. This process use generative adversarial networks using StyleGAN2 to create high quality synthetic thermal images and obtain training data to build face recognition models in deep learning. As a result, a database of 39,000 synthetic thermal images generated using different variants of StyleGAN2 was obtained. The growth of Generative Adversarial Networks (GANs) is rapid.

“UI Code Generation using Deep learning” Proposed by AditiDeolekar, KimayaDhanawade, Rahul Chavan, SayaliBhambure inn year 2021. A full functional automated UI code generation System would be useful in GUI code generation and software UI prototyping. A system having deep neural networks can learn latent variables which describe the objects in an image and their relationships with the corresponding variable-length textual descriptions.

“A Realistic Image Generation of face from Text Description Using the Fully Trained Generative Adversarial Networks” Proposed by Muhammad Zeeshan Khan, SairaJabeen, Muhammad UsmanGhani Khan, TanzilaSaba, AsimRehmat, AmjanRehman , (Senior Member, IEEE), and UsmanTari in year 2021. He problem solved is the proposed work trained the text encoder as well as the image decoder at the same time to generate more accurate and efficient results. In addition to the proposed methodology, another contribution is to generate the dataset by the amalgamation. The technique is used to the conditional GAN to build the twoend to end network for text to image generation.

“Face Generation based on Race using Generative Adversarial Networks” Proposed by P. N. Siva Jyothi , K. Kranthi Kumar, PratishtMathur,Mohammad Junaid, T. Vinay Reddy in year 2020. The problem statement is the Deep Convolutional Generative Adversarial Networks, that use convolutional neural networks for the generator,

and the discriminator, have been utilized in order to accomplish the task of the generation due to advantages over conventional GANS such as less noise, greater stability of results. Deep Convolutional Generative Adversarial Networks was used. It is an extension of the classic GAN, wherein the generator and the discriminator are Convolutional Neural Networks instead of a Multi-Layered Perceptron.

III. LIMITATIONS OF EXISTING WORK

- **Realism and Quality Variability:** Generated faces may vary in terms of realism and quality. Some generated faces might look highly realistic, while others could appear distorted or unnatural, depending on the model and training data.
- **Limited Control:** Users often have limited control over the specific attributes of generated faces, such as fine-grained features or expressions. Customization can be challenging.
- **Training Time:** Training deep generative models can be time-consuming, especially when using large datasets, and may require expertise in machine learning and access to substantial computational resources.
- **Generalization:** Deep learning models may struggle to generalize well to diverse facial attributes, ethnicities, or age groups not well-represented in the training data, leading to biases and inaccuracies.

IV. CONCLUSION

Image to image translation is a powerful deep learning technique that has numerous applications in computer vision. In this project, we aim to implement image to image translation for face image synthesis from sketches. The project methodology involves various phases, including research and planning, data collection and preparation, model selection and training, evaluation and optimization, software development, and deployment and maintenance. The expected outcomes of the project are a software product that can generate realistic face images from sketches, improved performance of the image-to-image translation model, and features like face morphing and face copy-paste. Overall, this project will provide a valuable contribution to the field of image to-image translation and have numerous practical applications.

V. REFERENCES

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