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Data Augmentation for Medical Image Analysis

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ABSTRACT: In the field of medical research, a significant challenge often arises when it comes to acquiring a substantial collection of images for training computers to comprehend various health issues. Take, for instance, thyroid problems; you might think of it as trying to piece together a jigsaw puzzle, but some crucial puzzle pieces are missing. To address this issue, scientists turn to Generative Adversarial Networks (GANs), sophisticated computer tools that act like talented digital artists. These GANs are designed to generate new images that closely resemble real medical images, despite not being authentic. They fill in the missing puzzle pieces, thus aiding in the study of medical conditions like thyroid problems, especially when there's a scarcity of genuine medical images.

In assessing the effectiveness of this approach, researchers employ a method known as the Inception Score. This evaluation process begins by collecting genuine medical images used by healthcare professionals, such as ultrasound images of the thyroid. These real images demonstrate an average Inception Score of 1.4298, signifying that they exhibit high quality and come in a range of different variations, akin to the diverse pieces of a complex puzzle. However, when the computer-generated images are subjected to the same evaluation, their average score is slightly lower, measuring at around 1.1263. This discrepancy suggests that, while the computer-generated images serve a valuable purpose, they do not quite match the quality and diversity seen in authentic medical images.

KEYWORDS : Generative Adversarial Networks (GANs), Ultrasound Thyroid Images, Inception Score (IS), Generative models.

I. INTRODUCTION

In the realm of medical research, the availability of diverse and extensive collections of high-quality medical images is essential for training advanced machine learning algorithms and developing innovative diagnostic tools. However, a persistent challenge in this field is the scarcity of such images, hindering progress like missing pieces in a crucial puzzle. To address this issue, a team of researchers and data scientists has turned to Generative Adversarial Networks (GANs), AI systems that act like virtual artists, generating synthetic medical images that closely resemble real ones.

GANs play a crucial role in overcoming the data scarcity hurdle by crafting artificial images that enrich datasets. These images fill the gaps left by the lack of genuine data, offering a promising solution to create comprehensive, diverse, and representative medical image collections. To assess the quality and diversity of these synthetic images, researchers employ a sophisticated evaluation method called the Inception Score. This score is crucial for shaping the future of data augmentation in medical image analysis, providing insights into the realism and diversity of the generated images.

The Inception Score works by subjecting synthetic images to evaluation by a pre-trained neural network model, such as Google's Inception model. This process assigns scores reflecting the images' resemblance to real medical images and their diversity. Fine-tuning GAN models based on Inception Scores holds the potential for significant improvements in the generation of synthetic medical images, aiming for a future where these images are indistinguishable from real ones and



cover a wide variety of medical scenarios. With a robust foundation of diverse and representative data, the future of medical research promises transformative advancements, leading to novel diagnostic tools, more effective treatments, and ultimately, healthier lives through the synergy of artificial intelligence and rigorous evaluation.

II. LITERATURE SURVEY

Le et al.[1] concentrate their study on the application of GANs for medical image fusion, as presented in their work titled "A Generative Adversarial Network for Medical Image Fusion". The authors specifically delve into the potential of GANs as a tool for merging medical images, with a dedicated focus on improving the quality and information content of these images. By honing in on the fusion process, their research aims to contribute valuable insights and advancements to the field of medical imaging, potentially enhancing diagnostic capabilities and paving the way for innovative applications in healthcare. In doing so, Le et al. address the pressing need for sophisticated techniques that can improve the utility of medical images in the diagnostic and research domains. The research by Hoo-Chang Shin and team[2] focuses on using smart computer systems called GANs to create new medical images. They explore how GANs can be helpful in two important ways: first, by generating more images to add variety and richness to medical datasets, which are like collections of pictures used by doctors and researchers. Secondly, the new images generated did not contain any personal information thus ensuring patient privacy. This work is crucial because having diverse medical images helps to train computer programs to better understand and analyze different health conditions. The researchers are likely to discuss how GANs can play a key role in both improving the quality of medical data and protecting patient privacy in the field of medical imaging. The literature on "Semi-Supervised Attention-Guided GANs(SAG-GANs) for Data Augmentation on Medical Images" by Chang Qi and his collaborators[3] explores an innovative approach to improving the training of computer models. They introduce a technique called SAG-GANs which combines aspects of both semi-supervised learning and attention guidance. The goal is to generate additional medical images for training purposes, enhancing the model's ability to understand and analyze medical data more effectively. By incorporating attention guidance, the authors aim to ensure that the generated images focus on the most relevant regions for accurate analysis. This method is particularly relevant in the medical field, where datasets are often limited.. Their study discusses the details of SAG-GANs and their potential impact on the field of medical image analysis [3]. Nripendra Kumar Singh and Khalid Raza [4] focus on the generation of medical images to create synthetic medical images by emphasizing the significance of these techniques in overcoming challenges related to limited data availability and enhancing the diversity of existing datasets. The authors concentrate on the implications of GAN-based image generation within the medical domain, specifically discussing its impact on the improvement of diagnostic tools, the training of machine learning models, and the advancement of medical research.

III. METHODOLOGY

The present research focuses on generating medical images using deep learning approaches as there is a problem of data scarcity in the field of medical imaging. We mainly focus on thyroid cancer to increase the dataset of thyroid images for better analysis of thyroid images and produce an effective model to tackle the thyroid disease. The images are generated based on GAN with the help of python libraries such as NumPy,OpenCV (cv2), pandas, Matplotlib, Seaborn, and TensorFlow. The general architecture of the present system is shown in figure 1.

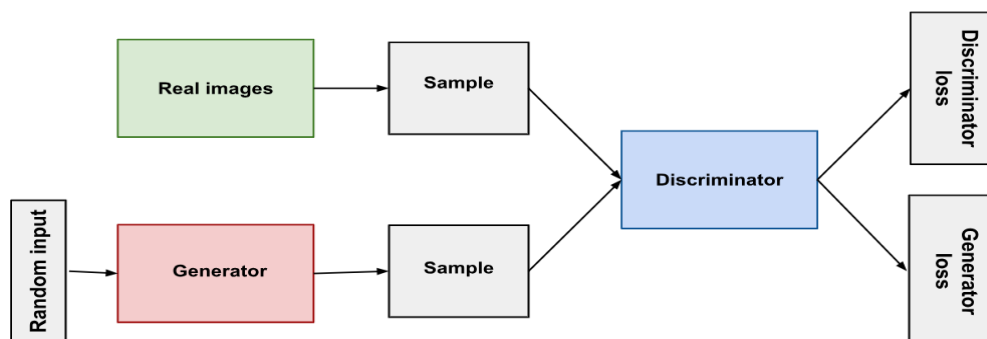


Figure 1: Architecture of GAN



To generate thyroid images we have collected 46 ultrasound thyroid images from Kaggle[5]. The collected images are of different sizes so by applying preprocessing methods we have resized all the images into 64x64. In general the GANs contain Generator model and Discriminator model where the generator model takes random noise as input and generates the new images. On the other hand, the discriminator will take these new images as input and classify whether the image is fake or real. To do this the generator calculates the loss in every generation cycle and it tries to reduce the loss. In the same way, the discriminator also tries to reduce the discriminator loss for ensuring efficient classification. The generated model is evaluated based on the Inception Score metric. The Inception Score(IS) is an objective performance metric, used to evaluate the quality of generated images or synthetic images, generated by Generative Adversarial Networks(GANs). It measures how realistic and diverse the output images are[6]. We have calculated the inception score for the real images dataset and for the generated images dataset. Inception score for real images is 1.4298 and for generated images is 1.1263 which is almost near to the inception score of the real images.

IV. RESULTS

We have developed a model for generating thyroid images in the field of medical imaging. Collected 46 images from kaggle and generated 550 thyroid images which resemble actual images. The sample input images and generated images are shown in figure 2 and figure 3 respectively.

Figure 2: Input Image

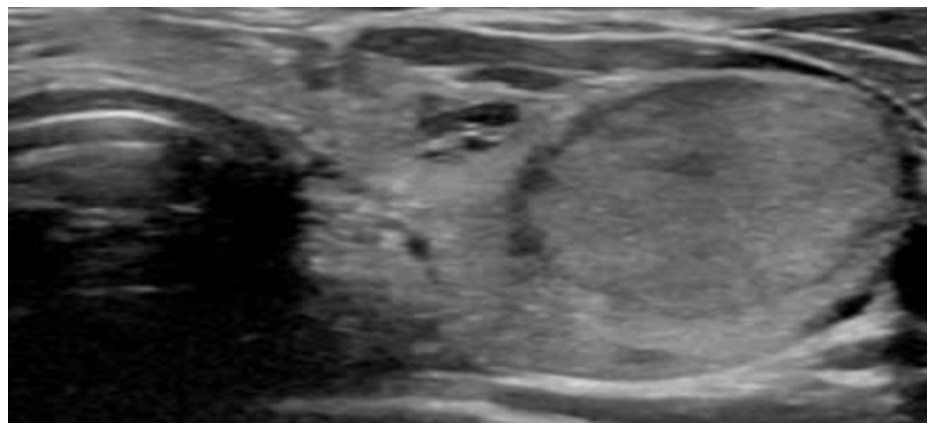
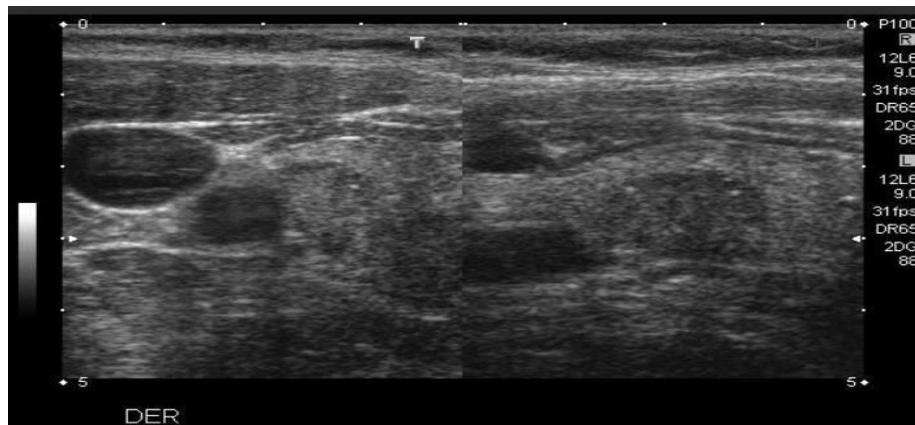


Figure 3: Generated Image



V. CONCLUSION

The GAN architecture employed in the experiment exhibits potential in the generation of medical images. The use of the Inception Score offers a valuable quantitative metric to evaluate the quality and diversity of the generated images, providing insights into the effectiveness of the GAN's performance. This experiment, as elucidated in the provided code, serves as an initial step and lays the foundation for future enhancements. There is ample room for extension and optimization to achieve more robust image generation capabilities. The outcomes suggest a favorable trajectory for further research and development, presenting an opportunity to enhance and broaden the current methodology in medical image synthesis for increased sophistication and impact.

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