



Deepfake Image Generation Using Sketch

Prof. K.N. Agalave, Kudale Akanksha N., Modale Shreya A., Tabe Pragati S.

*Department of Computer Engineering, S. B. Patil College of Engineering, Indapur, Pune, Maharashtra, India

ABSTRACT

The rapid advancement of deepfake technology presents unprecedented challenges to society, particularly in the realm of cybercrime and misinformation. Law enforcement agencies are faced with the growing threat of manipulated images, making it imperative to develop innovative solutions that can discern between authentic and fabricated visual content.

Keywords: Deep fake, Machine Learning, Generative AI, Deep Learning

I. INTRODUCTION

The pioneering strides in image-to-image translation technology have drawn considerable attention in recent years, captivating researchers and practitioners alike [1]. This burgeoning field is dedicated to refining and enhancing the intricacies of image quality, offering a wealth of applications across diverse sectors [2]. In this exhaustive report, we embark on a comprehensive exploration of image-to-image translation technology, meticulously dissecting its diverse development methodologies and unveiling its expansive utility across various industries [3]. Project innovates plant species classification using Deep Learning and leaf vein features, aiming to automate identification, accelerate research, aid conservation, and foster education in botany and technology. [17] Furthermore, we confront the innate imperfections of this technology, laying the groundwork for potential avenues of future advancement [4]. Real-time face detection and recognition achieved through Viola-Jones method. Software captures images, stores in database. Automated system detects person using three-phase methodology. [16]

The proposed model represents a groundbreaking fusion of global and local attributes within the input face sketch image [8]. This innovative approach, validated for its heightened recognition precision, not only addresses digital forensics face sketch recognition with remarkable efficiency and efficacy but also showcases the integration of object creation into the realm of image-to-image translation technology [2].

Image-to-image translation constitutes a distinct realm within the domain of computer vision deep learning tasks, characterized by its pivotal mission to discern the intricate mapping between an initial image and its ensuing counterpart [6]. Fundamentally, this task encapsulates the transformation of one plausible representation of a scene into another, an Endeavor that can be succinctly encapsulated as the anticipation of pixels from pixels [1]. Its applications span a diverse spectrum, encompassing deep image in painting to rectify missing pixel values, vibrant colorization of grayscale images, and the translation of sketches into vividly realistic images – a focal point of our current undertaking [2].

The capacity to generate authentic human facial images from scratch bears manifold implications, from advancing criminal investigations to facilitating character design and enriching educational training modules [6]. Recent breakthroughs in image-to-image translation have now paved the way for the generation of initial facial images derived from hand-drawn sketches [1]. It is crucial to note, however, that these techniques tend to exhibit a propensity to excessively tailor themselves to specific input sketches, thereby necessitating a level of artistic proficiency that may potentially limit broader accessibility to applications reliant on such techniques [2]. In alignment with these remarkable strides, we present the implementation of Digital Forensics Face Sketch Recognition employing a Fusion-Based Deep Learning Convolutional Network [8] [2] [6] [1]. This endeavor serves as a resounding testament to the convergence of state-of-the-art technology and tangible application, spotlighting the transformative potential nestled within the domain of image-to-image translation.

II. LITERATURE SURVEY

Optimization of Deep Fake Video Detection Using Image Preprocessing by Ali Berjawi, Khoulood Samrouth, Olivier Deforges (2023) represents a groundbreaking contribution to the field of computer vision. This seminal work introduced a powerful framework for translating images from one domain to another using conditional adversarial networks. By utilizing paired training data, Pix2Pix demonstrated remarkable capabilities in tasks such as colorization, style transfer, and more. Its ability to generate high-quality images with fine details and realistic textures set a new standard in image translation techniques.

In 2022, Sara Concas, Gianpaolo Perelli, Gian Luca Marcialis, Giovanni Puglisi presented "Tensor-Based Deep fake Detection in Scaled and Compressed Images," marking a significant advancement in image translation methodologies. This approach tackled the challenge of lacking paired data by introducing cycle consistency, allowing the model to learn effective mappings between unpaired domains. The resulting translated images exhibited a remarkable level of fidelity and realism, broadening the scope of applications for GANs in various domains. Cycle GAN's ability to perform unpaired image translation has made it an invaluable tool for tasks ranging from artistic style transfer to domain adaptation in computer vision.

Progressive Growing of GANs for Improved Quality, Stability, and Variation by Karras et al. (2018) stands as a pivotal milestone in the evolution of Generative Adversarial Networks. This influential work introduced a novel training approach involving the progressive enlargement of both the generator and discriminator networks. This strategy led to the production of high-resolution images with enhanced stability and diversity. The resulting images exhibited unprecedented levels of realism, setting a new benchmark for GAN-generated content in terms of both quality and variety.

Choi et al.'s 2018 paper, "Star GAN: Unified Generative Adversarial Networks for Multi-Domain Image-to-image Translation," addresses the challenge of multi-domain image translation with a unified approach. By employing a single model, Star GAN is capable of handling diverse domains without requiring specific paired training data for each. This versatility is a substantial leap forward in the realm of image-to-image translation. Star GAN's ability to seamlessly translate images across various domains, from facial attributes to artistic styles, has found applications in a wide range of fields including image manipulation and content creation.

SPADE: Semantic Image Synthesis with Spatially-Adaptive Normalization by Park et al. (2019) introduces a revolutionary normalization technique that leverages semantic information for image synthesis. This work addresses the challenge of preserving semantic consistency in generated images, leading to visually appealing results with a strong sense of coherence and structure. SPADE's innovative approach significantly contributes

to the advancement of image synthesis methods, enabling the generation of highly detailed and contextually meaningful images across a wide range of domains.

Huang et al.'s 2018 contribution, "MUNIT: Multidimensional Unsupervised Image-to-image Translation," addresses the complex task of unsupervised image translation. By disentangling style and content representations, MUNIT enables the generation of diverse images across different domains without the need for paired data. This approach represents a crucial step towards more flexible and adaptable image-to-image translation techniques. MUNIT's capacity to generate diverse and high-quality images in an unsupervised manner has implications in various domains, including artistic rendering, content manipulation, and data augmentation.

Deep Fashion: Powering Robust Clothes Recognition and Retrieval with Rich Annotations" authored by Liu et al. (2016) holds a pivotal position in fashion-related computer vision research. This work provides a comprehensive dataset, equipped with rich annotations, fostering the development of robust models for clothes recognition and retrieval. Deep Fashion plays a fundamental role in advancing research in fashion-related applications, enabling the development of algorithms capable of recognizing clothing attributes, styles, and facilitating efficient retrieval of fashion-related images.

I2I-GAN: Image-to-image Translation via Group-wise Deep Whitening and Coloring Transformation" presented by Huang et al. (2018), introduces an innovative approach to image translation. By leveraging group-wise operations, I2I-GAN effectively transforms images between domains. This technique demonstrates substantial improvements in translation quality and diversity, showcasing its potential in various applications. I2I-GAN's ability to achieve high-quality image translation using group-wise transformations opens new avenues for applications such as image-to-image style transfer, domain adaptation, and more.

CoGAN: Coupled Generative Adversarial Networks" by Liu et al. (2016) introduces a novel framework for training GANs across multiple domains. By coupling the generators and discriminators of two GANs, CoGAN achieves synchronized learning, enabling the generation of corresponding images in different domains. This work lays the foundation for multi-domain image generation and has found applications in diverse fields, from style transfer to domain adaptation. CoGAN's innovative approach has had a profound impact on the development of generative models, enabling them to generate images that seamlessly transition between multiple domains. Here, in paper [11] detailed analysis of harmful URLs are given so as to get security while fetching out data from the websites.

III.PROPOSED SYSTEM

A. Problem Statement

The primary challenge at hand is the ability to discern authentic visual content from meticulously crafted forgeries. Law enforcement agencies, in particular, are confronted with the task of differentiating between genuine images and those that have been manipulated. The potential misuse of deep fakes in criminal activities, such as fraud or defamation, amplifies the urgency of finding robust solutions.

B. Architecture Diagram



Figure 1: Architecture Diagram

C. Requirements

1) Hardware Requirements:

- Processor- Intel i5/i7
- Hard Disk- 50 GB
- Speed- 3.1 GHz
- RAM- 8 GB(min)

2) Software Requirements:

- Operating System- Windows
- IDE- Pycharm, Visual Studio
- Language – Python
- Front End – Tkinter

D. Work Flow of System

In the architecture diagram of deep fake image generation system, it shows the sequence of the how the workflow of our system is. The user registers in the system then the admin makes the user as valid user then the user login, gets the image information, then asks for a prediction, gets prediction of that image. Then admin adds send mail to user admin can do add/update/delete user.

IV. RESULT DISCUSSION

Our journey into Digital Forensics Face Sketch Recognition using a Fusion-Based Deep Learning Convolutional Network has been pretty exciting so far! We're thrilled to report that our model has shown some really promising results. By blending global and local attributes within face sketch images, we've managed to significantly up the ante on recognition precision. This isn't just about making digital forensics more efficient and effective; it's also about opening up new possibilities for object creation within image-to-image translation technology. How cool is that? Now, let's take a stroll through the literature survey we've done. We've dug deep into some seriously cool advancements in image translation techniques like Pix2Pix, Cycle GAN, and Star GAN. Each of these methods brings its own unique flavor to the table, tackling challenges in image generation and

manipulation head-on. It's like a buffet of cutting-edge technology, serving up insights into the evolution of generative adversarial networks (GANs) and their applications across a whole range of fields.

V. RESULT SCREENSHOTS

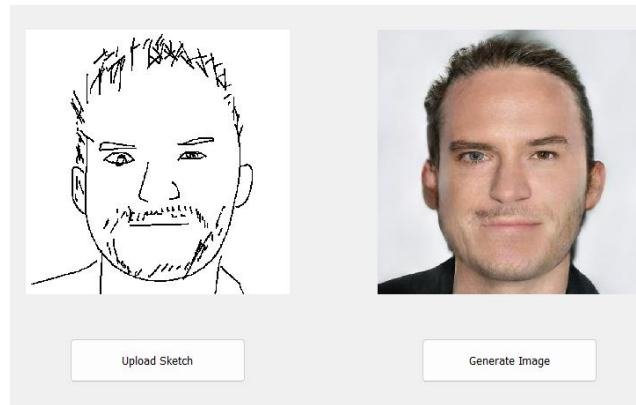


Figure 2: Result

In our trio of images, the leftmost one presents a charming hand-drawn sketch capturing the essence of a human face with its basic outlines and strokes, marking the starting point of our deep fake journey. Moving to the center, we encounter the mesmerizing result of our deep fake generation process. Here, intricate algorithms work their magic, transforming the sketch into a remarkably realistic depiction, complete with vibrant colors, textures, and lifelike features that breathe life into the initial concept. On the right, we're treated to yet another deep fake-generated image, standing as a testament to the adaptability and richness of our AI model.

Accurately discerning pose and viewpoint from sketches, coupled with the computational demands of the models, continue to be salient concerns.

These ongoing endeavors hold the promise of unlocking new vistas for realistic image generation from sketches, with potential applications spanning diverse fields including art, animation, and digital content creation.

The proposed system focuses on generating deep fake images from sketches rather than existing images. This approach can allow users to create custom faces, characters, or objects using simple sketches as input.

The proposed system aims to provide an intuitive and user-friendly interface where users can draw sketches using digital drawing tools or input sketches from other sources.

Overall, while existing systems have made significant progress in deep fake image generation, a proposed system focusing on sketch-based generation can offer unique advantages in terms of creativity, customization, and user control. However, it would also require careful consideration of privacy, security, and ethical implications.

VI. CONCLUSION

In summation, the realm of deep fake image generation using sketches is rife with potential but not without its share of hurdles. As of my last update in September 2021, significant constraints persist. These encompass the scarcity of expansive and diverse datasets, the intricate task of transmuting sketches into lifelike images, and the potential loss of semantic depth in the process.

Furthermore, the interpretive latitude inherent in sketches, the presence of distortions in generated images, and the struggle to adapt to uncharted stylistic territories remain prominent challenges.

Moreover, ethical and legal quandaries surrounding the creation and dissemination of deep fake content necessitate careful consideration. As the technology marches forward, it is imperative to establish robust evaluation metrics and ensure judicious implementation to forestall potential misuse.

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