

IMAGE FORTIFY: RESILIENT MULTI-TASK LEARNING FOR SECURE AND SEAMLESS IMAGE RESTORATION

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ABSTRACT:

Digital images shared on social networks are vulnerable to security and privacy threats, especially from tampering attacks aimed at altering their content. Traditional detection methods often struggle with compressed or low-resolution images and lack self-recovery capabilities. This project introduces an Image Protector utilizing a novel approach combining a Vaccinator and an Invertible Neural Network. It employs multi-task learning across four modules: image vaccination, consistency maintenance, tampered pixel classification, and self-recovery. During the forward pass, an immunized image and edge map are generated, while the backward pass utilizes Run-Length Encoding to restore the original image, ensuring integrity and authenticity. Real-world tests demonstrate effective tamper localization and content recovery.

Keywords:

Digital images, social networks, Security threats, Privacy threats, Tampering attacks, Traditional detection methods, Compressed images

INTRODUCTION

Social networking involves using internet-based platforms such as Facebook, Twitter, Instagram, and Pinterest to maintain connections with friends, family, colleagues, or customers. These platforms serve both social and business purposes and provide significant opportunities for marketers. Facebook, boasting 2 billion daily users, stands as the largest social network. Other popular platforms include Instagram, Twitter, WhatsApp, TikTok, and Pinterest. Social networks are characterized by user-generated content, profile creation, and the establishment of lasting connections through features like friending or following. While social networks focus on individual connections, social media emphasizes sharing content with a broader audience. Most social networks also function as social media sites..

OBJECTIVES

- To effectively train the network and enhance its resilience against real-world
- threats
- To develop the Cyber Vaccinator for Image Immunization.
- To implement the Vaccine Validator for Media Distinction.
- To establish binary masks for object contours.
- To design Forward Pass for Tamper Detection.
- To develop a Localizer for Tampered Area Detection.
- To create Backward Pass for Image Self-Recovery.
- To simulate Adversarial Attacks for Effective Network Training.
- To encourage Image Self-Recovery Mechanisms.
- To achieve Proactive Image Immunization and Restoration.



LITERATURE REVIEW

| Toolniques | | | | | | | | | | |
|---|---|--|--|---|--|--|--|--|--|--|
| Authors | Title | Techniques Used | Advantages | Disadvantages | | | | | | |
| Dong et al. | MVSS-Net for enhanced detection accuracy | Deep learning, neural networks | Comprehensive Coverage of Recent Advances | Lack of Critique on Methodologies | | | | | | |
| Liang et al. | Robust hashing techniques for image copy detection | Hashing algorithms | Clear Identification of Key Researchers | Limited Detail on Implementation | | | | | | |
| Lin et al. | Method for identifying subtle manipulation cues | Image processing, feature extraction | Highlights Diverse Methodologies | Scope Limited to Specific Algorithms | | | | | | |
| Zhang et al. | Dual-branch approach for improved detection robustness | Dual-branch neural networks | Mentions Practical Resources | Potential Publication Bias | | | | | | |
| Liu et al., Wu et al., Lin et al. | Novel network architectures for precise manipulation localization | Various neural network architectures | Provides Insight into Methodological Approaches | Missing Comparative Analysis | | | | | | |
| Wang et al. | Object Former | Object detection, computer vision | Advances state- of-the-art in object detection | Implementation complexity may be high | | | | | | |
| Chen et al. | Multi-view multi-scale supervision | Multi-view learning, deep learning | Enhanced learning from multiple perspectives | Requires significant computational resources | | | | | | |
| Matthes | "Python Crash Course" | Python programming | Easy-to-follow introduction to Python | Focuses primarily on basics, may not cover advanced topics in depth | | | | | | |
| Sweigart | "Automate the Boring Stuff with Python" | Web scraping, Automating GUI Applications | Simplifies automation tasks with practical examples | Limited coverage of advanced Python topics | | | | | | |

METHODS AND MATERIALS

- **Techniques Used**: This column outlines the primary techniques or methodologies employed in each referenced work, such as deep learning, hashing algorithms, image processing, etc.
- Advantages: Summarizes the strengths or positive aspects of the literature review based on the authors and their titles.
- **Disadvantages**: Highlights the limitations or potential drawbacks associated with the literature review, offering a critical perspective on the discussed research.

This format provides a comprehensive overview, integrating information on techniques used along with advantages and disadvantages, which can aid in understanding the context and impact of each referenced work in image manipulation detection.

Recent advancements in image manipulation detection have catalyzed significant developments in algorithms and methodologies [1]. Noteworthy contributions include Dong et al.'s MVSS-Net, which enhances detection accuracy, Liang et al.'s robust hashing techniques for image copy detection [2][3], and Lin et al.'s method for identifying subtle manipulation cues [4]. Zhang et al. improved detection robustness with a dual-branch approach [5], while Liu et al., Wu et al., and Lin et al. introduced novel network architectures for precise manipulation localization [6][7][8]. Wang et al.'s Object Former and Chen et al.'s multi-view multi-scale supervision have further advanced the field. Additionally, practical resources such as Matthes' "Python Crash Course" and Sweigart's "Automate the Boring Stuff with Python" support real-world algorithm implementation.

Technical Description of components

| S.N | Field | Data Type | Field size | Constraint | Description |
|-----|----------|------------|------------|-------------|----------------|
| , | id | int | 11 | Primary Key | User id |
| 2 | Name | varchar | 20 | Null | User Name |
| 3 | Gender | Varchar | 10 | Null | User Gender |
| 4 | Dob | Varchar | 40 | Null | User dob |
| 5 | mobile | Big int | 20 | Foreign Key | User Mobile |
| 6 | email | Varchar | 20 | Null | User Email |
| 7 | adhaar | Varchar | 11 | Null | User aadhar |
| 8 | Usernam | Varchar | 20 | Null | Post username |
| 9 | password | varchar | 20 | Null | Post password |
| 10 | Date_tim | Time_stamp | Time_stamp | Null | Post Date time |

KEY COMPONENTS OF MULTI-TASK LEARNING (MTL)

- **Hard parameter sharing**: Shares hidden layers across tasks while maintaining task-specific output layers, which helps mitigate overfitting.
- **Soft parameter sharing**: Each task-specific model retains its own weights and biases, but these are regularized to encourage similarity across tasks, ensuring a balance between specialization and generalization.
- **Task clustering**: Groups similar tasks together to enhance knowledge transfer between related tasks, leveraging similarities for improved learning efficiency.
- **Shared layers**: Allows models to learn shared representations of data, fostering synergy between tasks and reducing redundancy in learned features.



- **Loss functions**: Customized for each task, accommodating varying task importance and optimizing performance accordingly.
- **Feature extraction**: Identifies task-specific features as well as shared patterns within the data, facilitating effective knowledge transfer across tasks.

SYSTEM SPECIFICATION

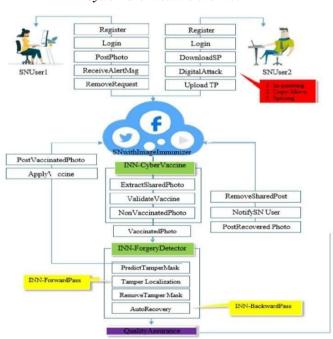
(6.1) Hardware Requirements

- **Processor:** Quad-core or higher for efficient parallel processing.
- RAM: 8 GB for seamless image processing and neural network tasks.
- Storage: 256 GB SSD for fast data access and storage.

(6.2) Software Requirements

- Operating System: Windows 10 or 11
- **Programming Language:** Python (version 3.6 or higher)
- Neural Network Framework: TensorFlow or PyTorch
- Image Processing Libraries: OpenCV and PIL (Pillow)
- Web Framework: Flask
- Database Integration: MySQL
- **IDE**: IDLE
- Web Technologies: HTML, CSS, and JavaScript

System architecture overview



RESULTS AND DISCUSSION



(8.1) SCREENSHOTS



Fig.8.1. Signup page

Social IV is the first app that is used to vaccinate the picture as Fig.5.1.1 Shows the login page and sign-up page as well.



Fig.8.2: Login page

show the username and the password after the sign-up page used for the login page.

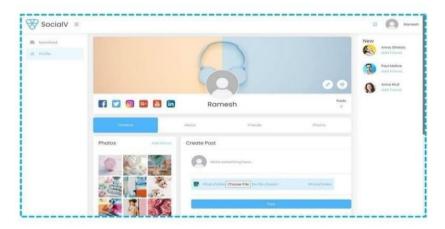


Fig. 8.3: Image immunizer dashboard

This picture shows the front page of the profile which is where the picture is uploaded and gets vaccinated.





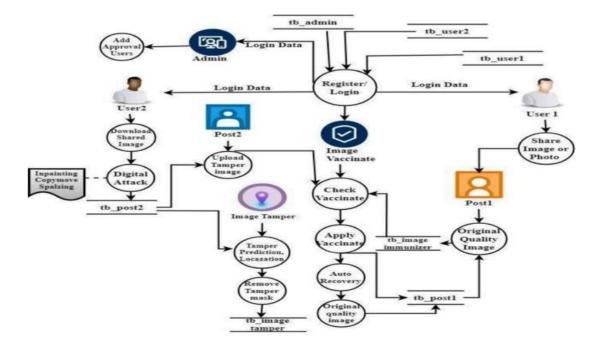
(8.4) Social media login page

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Invertible Neural Network:

Invertible Neural Networks (INNs) are potent tools for inverse design optimization, enabling both forward and inverse predictions at minimal additional cost. While the forward process predicts outputs from inputs, the inverse process determines potential input parameters for a given system response. INNs utilize affine coupling blocks and are adept at solving inverse problems; however, they currently encounter challenges in accurately quantifying uncertainty, particularly for regression tasks. The advancement of new loss functions aims to bolster the accuracy of the inverse process. INNs leverage latent variables to capture critical information and find applications in fields like computer vision and inverse design. Despite their utility, standard INN implementations typically provide prediction results without uncertainty estimation.





CONCLUSION

In conclusion, the project "Image Immunizer Middleware for Online Social Networks" presents an innovative solution to combat digital image attacks effectively. By leveraging Invertible Neural Networks and adversarial simulation, the system ensures the authenticity and integrity of images shared on social platforms. The Cyber Vaccinator Module processes images by applying imperceptible perturbations during vaccination, thereby safeguarding them against tampering. The Vaccine Validator distinguishes vaccinated media to enhance security measures. Through its Forward and Backward Passes, the system identifies and restores tampered areas, ensuring the recovered images align closely with their originals. Adversarial training further bolsters resilience against a wide range of attacks.

This middleware seamlessly integrates with social media platforms, providing users with real-time notifications regarding image status and offering the capability to restore tampered images. Ultimately, it aims to establish a secure and trustworthy environment within social media networks.

FUTURE ENHANCEMENT

Future enhancements for the Image Immunizer Middleware for Online Social Networks using Invertible Neural Networks (INN) aim to bolster its capabilities and adapt to evolving technology. Integrating blockchain technology can enhance transparency in image transactions, ensuring tamper-evident records throughout the lifecycle of shared images. Expanding the middleware to include multimodal content analysis, such as videos and audio, will provide a more comprehensive defense against digital manipulation within Online Social Networks (OSNs). These advancements underscore our commitment to robust security and holistic content protection.

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