

Dress Code Monitoring System Using YOLOv4

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ABSTRACT - The Dress Code Monitoring System using YOLOv4 aims to enhance compliance with dress code policies in educational institutions and workplaces through an innovative application of computer vision and deep learning technologies. This system leverages the capabilities of the YOLOv4 (You Only Look Once) algorithm, renowned for its real-time object detection and classification efficiency. By integrating high-resolution cameras and the YOLOv4 model, the system can accurately identify and analyze individuals' attire as they enter designated areas. The primary objective is to ensure adherence to specified dress codes, which can range from formal to casual or specific uniforms. The system operates by continuously monitoring live video feeds, detecting individuals, and classifying their attire in real time. If a mismatch with the dress code is detected, the system can trigger alerts or notifications to relevant authorities, such as administrators or supervisors. Furthermore, the application is designed to maintain privacy by anonymizing detected individuals and ensuring that data storage complies with relevant regulations. This innovative approach not only fosters a sense of professionalism and unity within the institution but also streamlines the enforcement of dress code policies, reducing the need for manual monitoring. The Dress Code Monitoring System using YOLOv4 represents a significant advancement in leveraging artificial intelligence for practical applications, demonstrating the potential for technology to enhance operational efficiency and compliance in various settings.

Keywords: Computer Vision, Deep Learning, Object Detection, Real-Time

1. INTRODUCTION

In professional and educational settings, dress codes play a significant role in fostering discipline, professionalism, and a sense of community. They serve to establish a cohesive environment, reflect institutional values, and enhance the overall ambiance of workplaces and academic institutions. However, ensuring consistent compliance with these standards poses a challenge, especially in large organizations where manual monitoring is labor-intensive and prone to errors.

Traditional methods of dress code enforcement often rely on human supervisors, leading to subjectivity, inconsistencies, and inefficiencies. With the rapid advancements in artificial intelligence and machine learning, these challenges can now be addressed using automated systems. Deep learning models, particularly those focused on object detection, have demonstrated remarkable potential in identifying patterns and features within images, making them ideal for this application.

YOLOv4 (You Only Look Once, version 4) is a cutting-edge object detection framework that stands out due to its high accuracy and real-time processing capabilities. By breaking down visual input into discernible objects, YOLOv4 is capable of swiftly detecting clothing items and classifying them into categories such as "formal," "casual," or "non-compliant." Its robust architecture ensures that it can handle complex environments with multiple individuals and overlapping objects, making it well-suited for real-world scenarios.

The primary objective of this project is to integrate YOLOv4 into a comprehensive Dress Code Monitoring System. This system not only automates the detection and classification process but also provides real-time alerts and analytics, enabling institutions to enforce dress codes efficiently. Additionally, the system incorporates advanced features such as scalability, adaptability to custom dress codes, and a user-friendly interface for administrators.

This paper explores the design, development, and deployment of the proposed system, emphasizing its technical underpinnings, challenges faced during implementation, and its impact on institutional management. By leveraging technologies like YOLOv4, Convolutional Neural Networks (CNNs), OpenCV, and Roboflow, this project sets a benchmark for AI-driven compliance monitoring in professional settings.

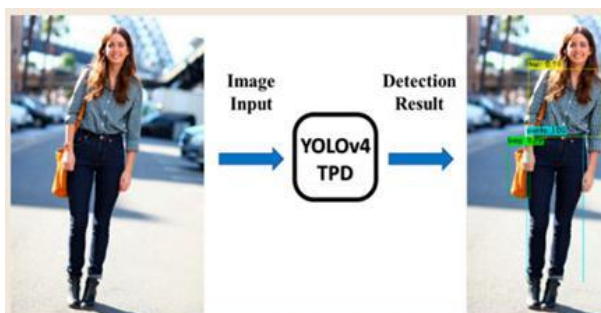
2. MATERIALS AND METHODS

2.1 Hardware Components

- Cameras: High-resolution cameras capable of 1080p video capture were used for real-time monitoring.
- Processing Unit: NVIDIA Jetson Nano was employed for local processing of video feeds and running the YOLOv4 algorithm.

2.2 Software Components

- YOLOv4 Algorithm: The pre-trained YOLOv4 model was fine-tuned using a custom dataset.



- Frameworks and Libraries: TensorFlow, OpenCV, and Robo flow were utilized for model implementation, image processing, and dataset preparation.

2.3 Data Preparation

A dataset of annotated images depicting compliant and non-compliant dress codes was curated using Robo flow for efficient dataset management and annotation. The dataset included a diverse range of clothing styles, colors, and environments to ensure robustness and adaptability of the model.

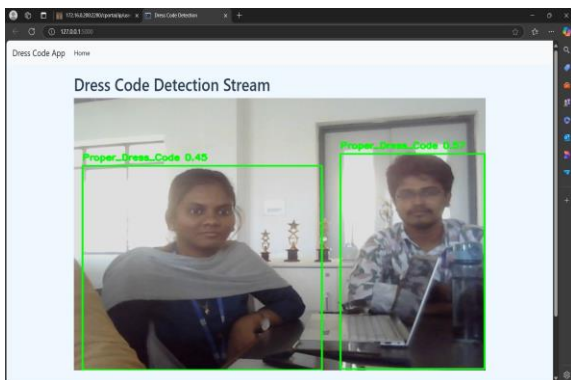
2.4 Methodology

The methodology followed for developing the Dress Code Monitoring System involved multiple phases that ensured comprehensive system design and implementation:

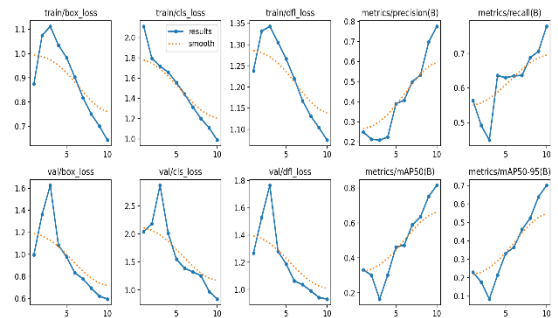
1. Data Collection and Annotation: A robust dataset was curated by collecting images from public repositories and institutional archives of more the 2500 images. These images were annotated with categories such as "formal," "casual," and "non-compliant" using Roboflow's intuitive platform. The annotations included bounding boxes around clothing items to train the YOLOv4 model efficiently. Data augmentation techniques, such as flipping, rotation, and brightness adjustments, were applied to increase dataset diversity and reduce overfitting.
2. Model Training and CNN Optimization: The YOLOv4 model was fine-tuned on the annotated dataset using a high-performance computing system. Training involved multiple epochs with iterative tuning of hyperparameters, such as learning rate, batch size, and momentum. Convolutional Neural Networks (CNNs) formed the backbone of the YOLOv4 model, enabling feature extraction from complex image patterns. Techniques like

dropout regularization and batch normalization were applied to improve the model's generalization and reduce overfitting.

3. **System Development:** A real-time monitoring system was designed using Python and OpenCV. OpenCV facilitated image preprocessing steps such as resizing, normalization, and noise reduction, ensuring optimal input quality for the YOLOv4 model. The system also included modules for video stream integration, real-time object detection, and automatic alert generation.
4. **User Interface Design:** A user-friendly interface was developed to enable administrators to monitor real-time outputs and access reports. The UI included functionalities to configure dress code parameters, view detection results, and receive alerts for non-compliance. The interface was built using Flask, ensuring compatibility with web and desktop applications.



5. **Testing and Validation:** The system was tested extensively under various conditions, including different lighting environments, camera angles, and clothing styles. Metrics such as precision, recall, and mean Average Precision (mAP) were used to evaluate the model's performance. Cross-validation techniques ensured that the system maintained high accuracy across unseen data.



6. **Deployment and Field Implementation:** The final system was deployed in a controlled environment, such as an educational institution. Cameras were strategically placed at entrances and common areas to capture video feeds. Feedback from users and administrators was collected to identify areas for improvement and ensure smooth operation.



3. RESULTS AND DISCUSSION

3.1 Performance Metrics

- **Precision:** 0.7759
- **Recall:** 0.7788
- **mAP50:** 0.8155

The system demonstrated high accuracy in detecting and classifying clothing items. Alerts for non-compliance were generated within 0.5 seconds of detection, ensuring real-time responsiveness.

3.2 Observations

The integration of CNNs significantly enhanced feature extraction, enabling the system to adapt to diverse clothing styles and lighting conditions. OpenCV facilitated efficient image preprocessing and analysis, while Roboflow streamlined dataset preparation and augmentation and live videos.

Moreover, the framework exhibited versatility and flexibility to institutional prerequisites by supporting unique updates to clothing standard boundaries through the UI. The utilization of live video takes care of permitted managers to screen consistence continuously, diminishing manual oversight prerequisites and functional failures. Besides, the definite investigation given by the framework assisted establishments with distinguishing normal clothing standard infringement and find proactive ways to address them.

Notwithstanding, during testing, the framework once in a while attempted to precisely characterize clothing things under outrageous lighting conditions or with covering articles of clothing. This featured the significance of assorted preparing information to deal with edge cases successfully. The perceptions underline the framework's capacity to work on functional effectiveness while additionally distinguishing regions for additional improvement.

3.3 Limitations

While the framework performed well under controlled conditions, certain constraints were noticed:

Lighting Conditions: Critical varieties in lighting, like shadows or faintly lit regions, sporadically affected the exactness of dress arrangement. The model's dependence on clear and steady visual sources of info stays a test.

Impediments: Covering pieces of clothing or extras, for example, scarves or packs, once in a while

prompted misclassification or inability to recognize infringement. Improving the model's capacity to separate between covering objects is basic.

Constant Handling: While the framework worked effectively in conditions with restricted pedestrian activity, high-thickness regions introduced difficulties to continuous location because of computational burden.

Flexibility: Albeit the framework was vigorous for predefined clothing standards, presenting completely new classes required retraining the model, which may not be achievable progressively applications.

Addressing these limitations involves incorporating advanced augmentation techniques, such as simulating various lighting conditions, and leveraging additional training data for improved robustness. Future iterations could also integrate edge computing devices with higher processing power to handle high-density environments effectively.

4. CONCLUSION

The Dress Code Monitoring System, which uses YOLOv4, represents a substantial leap in automated compliance monitoring. By utilizing deep learning for real-time object detection, the system lowers manual oversight while increasing operational efficiency. The addition of OpenCV for preprocessing and Robo flow for dataset management accelerated development. Future improvements will involve increasing the dataset, including more powerful CNN architectures, and creating enhanced user-friendly UI elements. These enhancements will improve the system's precision and adaptability, assuring consistent operation across a wide range of real-world settings. As a result, the Dress Code Monitoring System not only promotes adherence to established guidelines but also fosters a professional and respectful environment. Overall, this innovative approach represents a promising step forward in utilizing artificial intelligence for social regulation, offering a practical solution that enhances compliance while respecting individual expression.

5. REFERENCES

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6.BIOGRAPHIES



DHARSHINI V

1. Training YOLOv4 model
2. Testing and Evaluation
3. UserInterface Developmet



DEEKSHA G

1. Model training
2. OpenCV Implementation
3. Bounding Boxes in Real-Time



IRFAN BASHA I

1. Data Collection
2. Dataset Annotation
3. Labelling



ABDUL AHAMED B

1. Data Collection
2. Labelling
3. Report writing