



Real-Time Stress Detection System For Humans: An Image Processing And Machine Learning Approach

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Abstract: In the rapidly evolving and demanding realm of technology, individuals frequently face significant stress due to substantial workloads, stringent deadlines, and ever-changing work settings. Extended exposure to such stress can adversely affect both personal health and professional efficiency. Traditional stress detection methods, like self-reported surveys and physiological evaluations, tend to be intrusive and lack real-time functionality. Consequently, there is an increasing demand for more efficient and less invasive stress monitoring solutions. This research seeks to tackle this issue by creating an innovative stress detection system specifically designed for individuals. By utilizing advancements in image processing and machine learning, the system examines facial expressions and visual indicators that suggest stress. Through the use of advanced algorithms, the system can accurately assess and interpret stress levels in real-time. Additionally, the system incorporates continuous feedback mechanisms by consistently gathering employee surveys to monitor stress levels over time and assess the success of interventions. With its real-time capabilities and non-intrusive nature, this system represents a significant advancement over traditional stress detection methods, offering employees and organizations valuable insights to promote healthier work environments.

1.INTRODUCTION

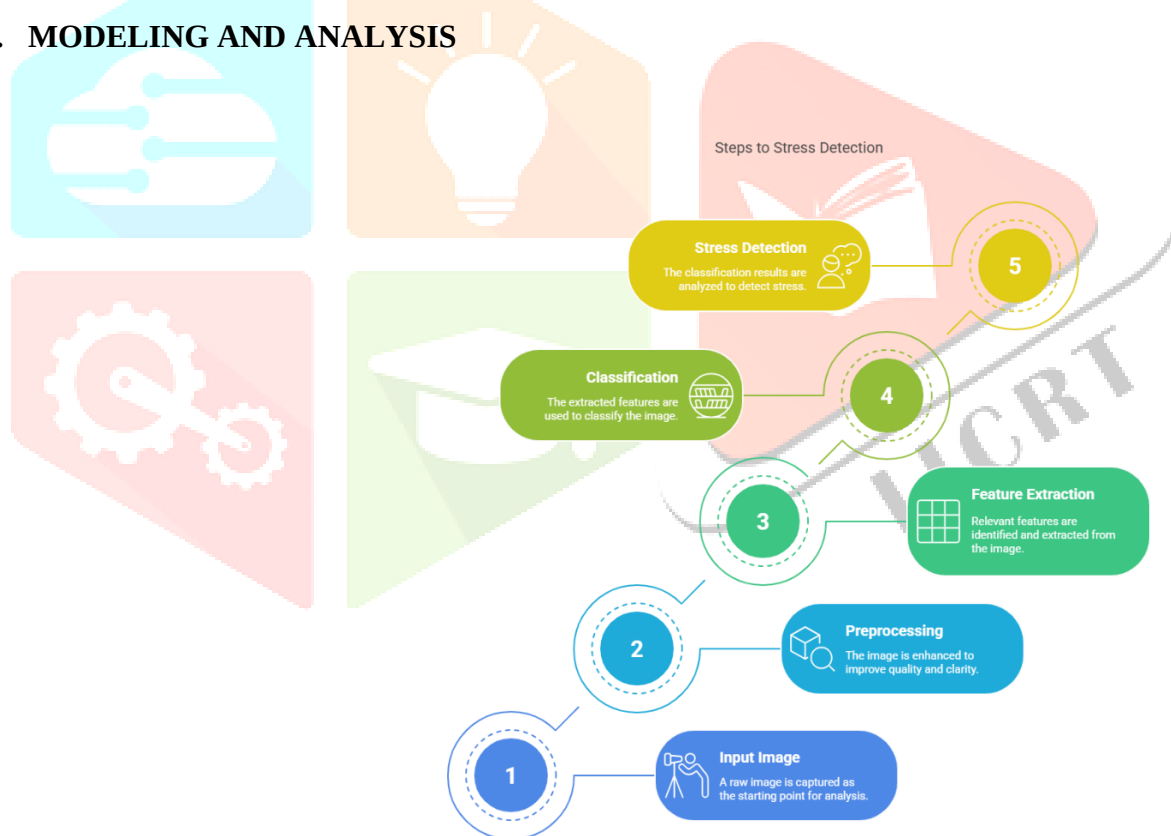
Employees in the swiftly evolving IT sector often face demanding work environments and substantial workloads, which can lead to increased stress levels. If left unmanaged, this stress can negatively impact both job performance and personal health. Traditional methods of stress detection, such as self-reports or physiological tests, have limitations in terms of real-time application and intrusiveness. As a result, there is a growing need for more innovative and effective approaches to stress monitoring and management. Our research aims to address this need by developing a comprehensive stress detection system specifically tailored for IT professionals, leveraging the latest advancements in image processing and machine learning technologies. The primary function of our system is to analyze facial expressions and other visual indicators of stress using image processing techniques. Machine learning algorithms are employed to classify and interpret these signals, providing accurate and timely stress evaluations. Additionally, the technology facilitates the regular collection of employee feedback through surveys, enabling continuous assessment of stress levels and the effectiveness of interventions.

For this study secondary data has been collected. From the website of KSE the monthly stock prices for the sample firms are obtained from Jan 2010 to Dec 2014. And from the website of SBP the data for the macroeconomic variables are collected for the period of five years. The time series monthly data is collected on stock prices for sample firms and relative macroeconomic variables for the period of 5 years. The data collection period is ranging from January 2010 to Dec 2014. Monthly prices of KSE -100 Index is taken from yahoo finance.

1. RESEARCH METHODOLOGY

- Convolution Layer:** The initial phase in feature extraction from an image involves executing a convolution. This process entails identifying the image's features through small data squares. By utilizing different filters, such as identity, edge detection, box blur, and Gaussian blur, it can carry out various functions like sharpening, detecting edges, and blurring.
- Pooling Layer:** When dealing with large images, pooling layers help decrease the number of parameters through a method known as spatial pooling. This approach also aids in reducing the dimensionality of the map.
- Fully Connected Layer:** The feature map matrix is converted into a vector format. By linking the layers, we have developed a model that leverages this feature.
- Softmax Classifier:** At the conclusion of the process, a classification function such as softmax or sigmoid is used to sort the results.

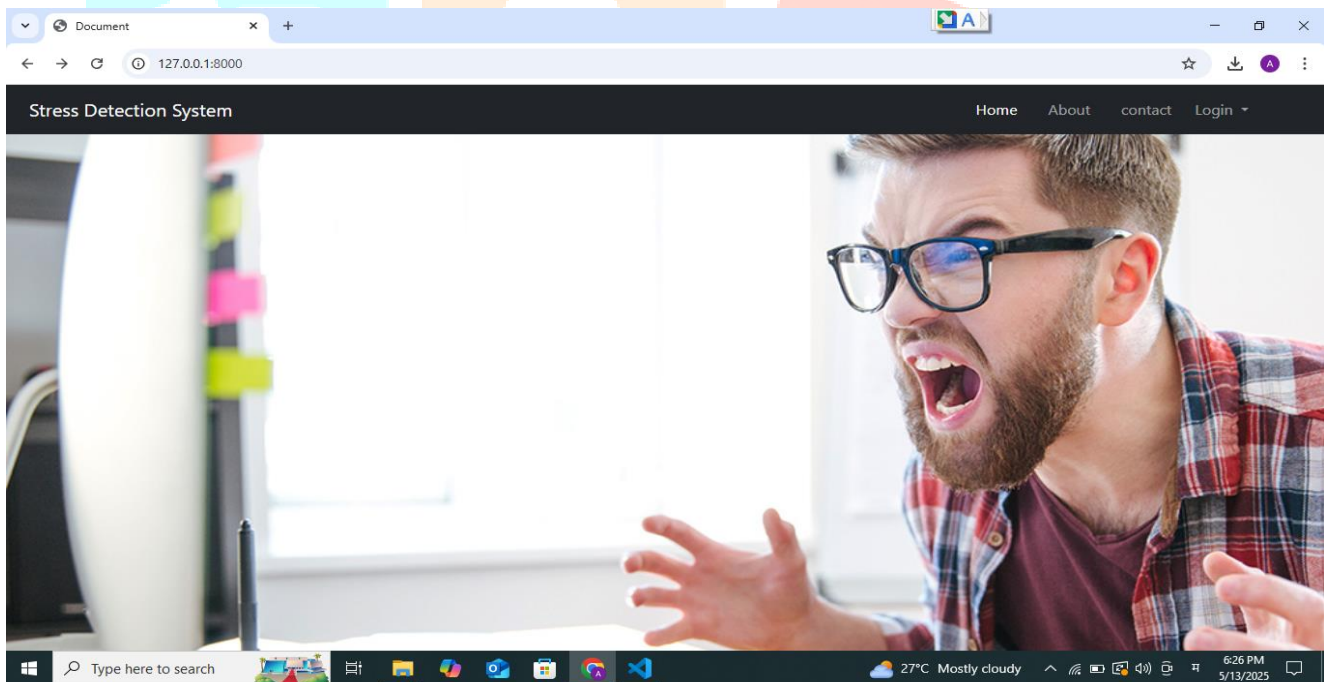
2. MODELING AND ANALYSIS



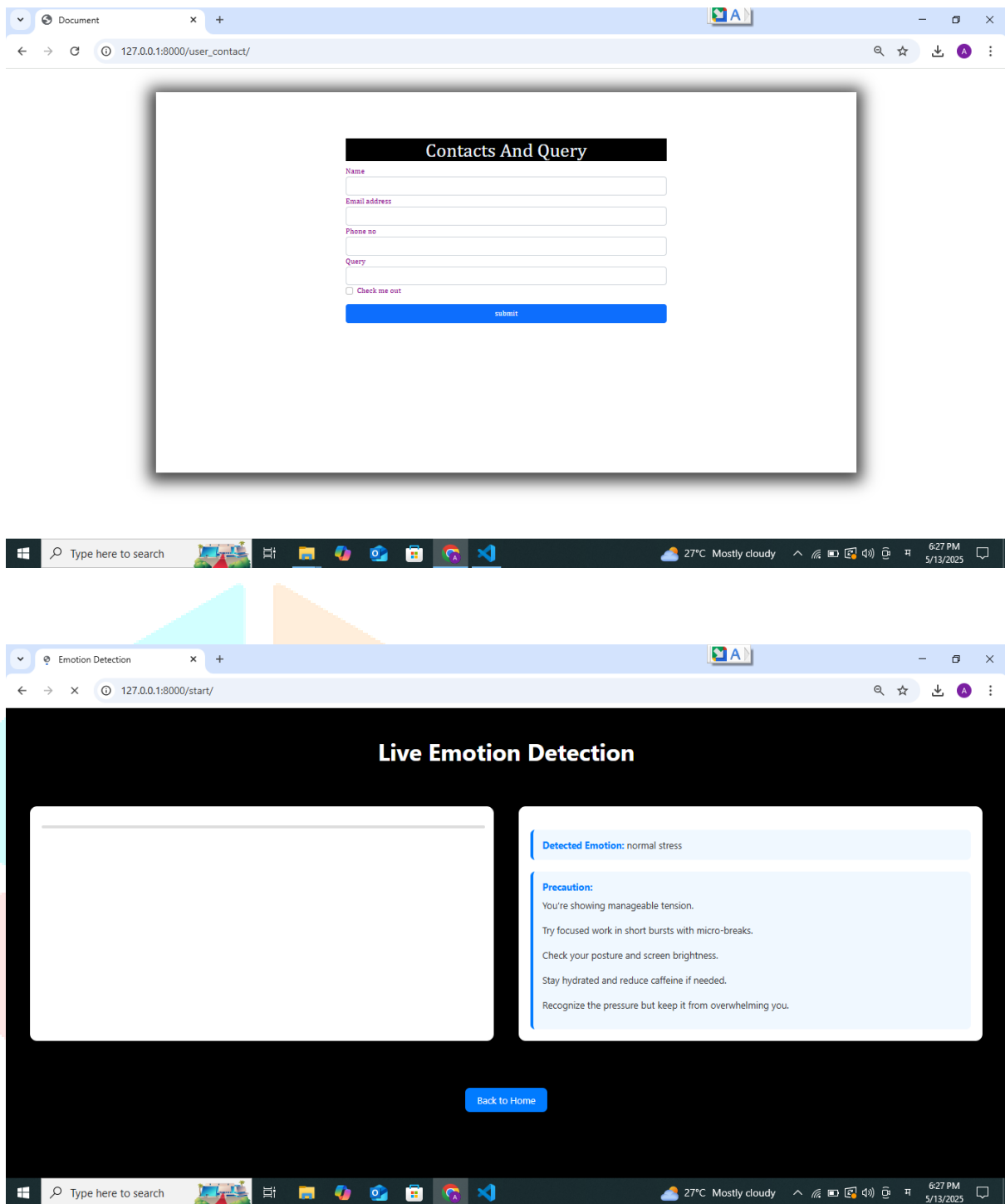
- Input Image:** The initial step involves an input image, serving as the foundational data required for stress analysis.
- Preprocessing:** This stage involves preparing the image through various steps to make it suitable for feature extraction. The preprocessing steps include:
 - Grayscale Conversion:** The image is transformed into grayscale, simplifying the data by reducing it to a single channel (intensity) instead of three (red, green, and blue). This reduction in complexity aids in focusing on structural details rather than color.
 - Resized Image:** The image is adjusted to a standard size, ensuring consistent processing, which allows the system to handle images of varying sizes uniformly.
 - Median Filtering:** A median filter is applied to the image to diminish noise. Reducing noise helps in obtaining clearer and more distinct features, enhancing the accuracy of feature extraction in the subsequent phase.

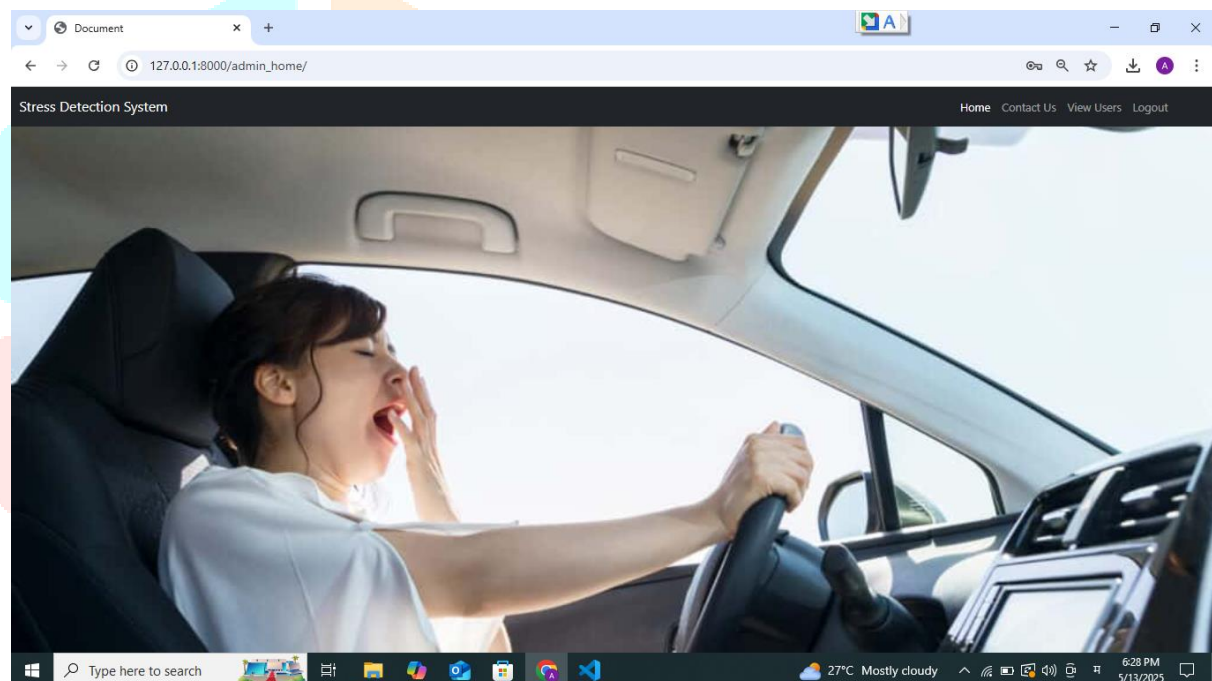
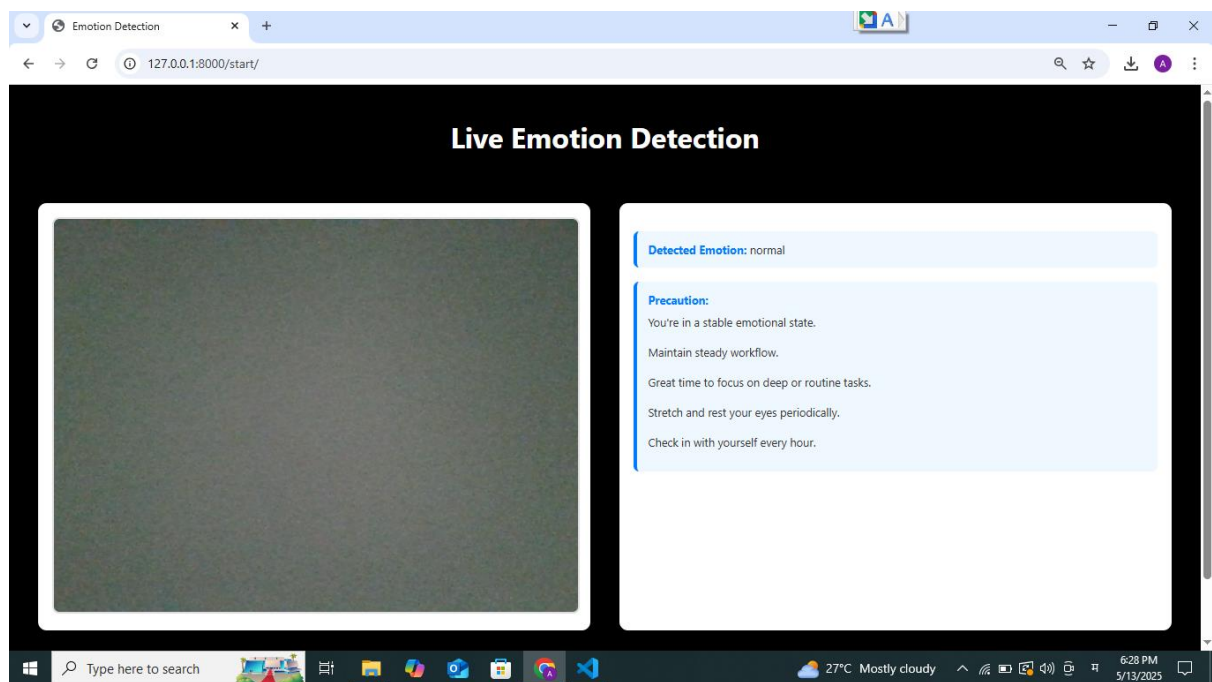
4. **Feature Extraction:** Following preprocessing, the system extracts essential features from the image. These features are vital for identifying patterns associated with stress. The types of features extracted include **Color Features:** These capture the color information in the image, which may encompass different shades and tones that could indicate stress (e.g., redness or paleness). **Texture Features:** These analyze the patterns in the image, such as smoothness, roughness, or repetitive patterns. Texture analysis may provide insights into stress indicators, as certain textures or patterns could be linked to stress responses. **Canny Edge Features:** The Canny Edge Detection algorithm is employed to identify edges within the image. Edge features help highlight key structures or boundaries, assisting in identifying significant areas or patterns in the image related to stress.
5. **Classification:** The extracted features are then input into a classification model. This model utilizes machine learning or statistical techniques to categorize the data based on predefined categories (e.g., stressed vs. not stressed). The classifier is trained to recognize patterns in the features that are likely associated with stress.
6. **Stress Detection:** The final output of the classification process is used to detect stress. Based on the classification results, the system determines if the input image shows signs of stress.
7. **Overall Flow:** The input image undergoes preprocessing to enhance quality, followed by the extraction of relevant features, which are used for classification. The classification results aid in detecting stress, completing the process from raw image input to stress analysis output.

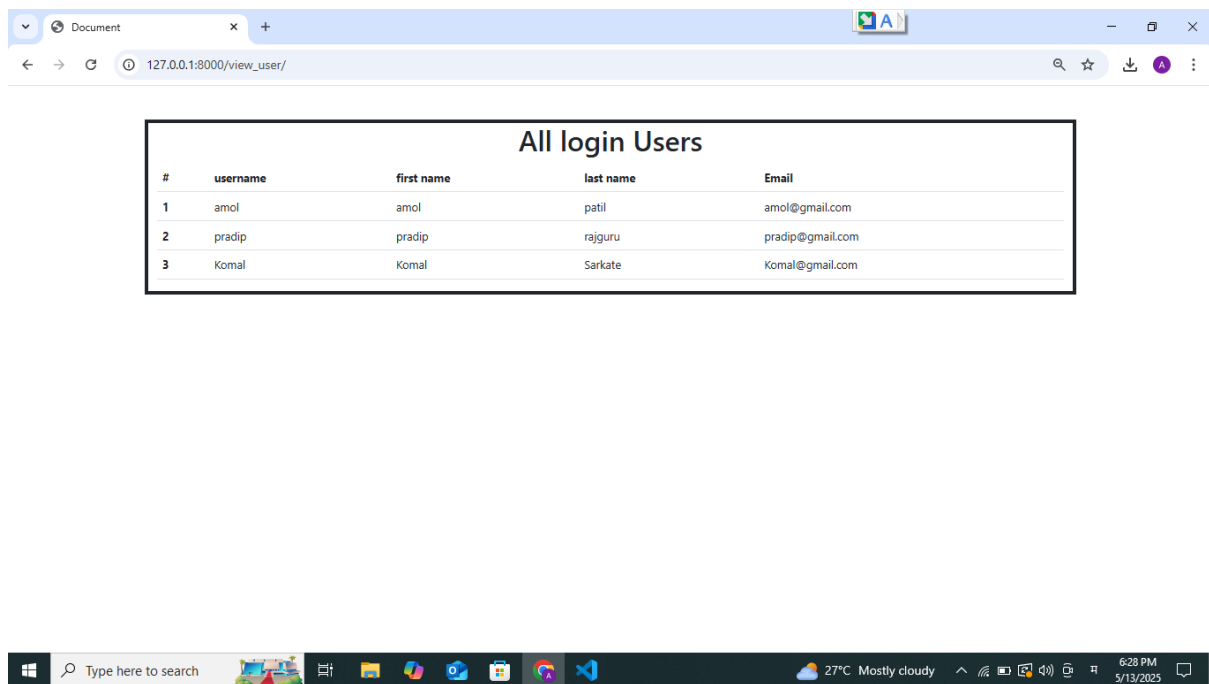
3. RESULTS AND DISCUSSION











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1	amol	amol	patil	amol@gmail.com
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4. CONCLUSION

This initiative marks a crucial advancement in merging machine learning with image processing technologies to tackle workplace stress among employees and individuals. By facilitating real-time stress detection and offering tailored interventions, the system aspires to create a healthier and more efficient work setting. Although promising, the project encounters several hurdles, such as concerns over data privacy, the precision of stress detection, and the ethical issues related to continuous monitoring. Nevertheless, with meticulous execution and ongoing improvements, the system holds the potential to significantly benefit both technology and society. It not only deepens our comprehension of stress management in high-pressure settings but also lays the groundwork for future innovations aimed at enhancing employee well-being and organizational performance.

5. REFERENCES

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