

Deep Learning-Based De-blurring/Denoising of Indian Heritage Images

Mr. Tejas Lambat¹, Mr. Shreyash Ukey², Mr. Pranav Wagh³, Mr. Sanket Tekam⁴
Mr. Snehal Raipure⁵, Mr. Ravindra Kale⁶

Students, G H Raisoni Institute of Engineering & Technology, Nagpur, India^{1,2,3,4,5}
Assistant Professor, G H Raisoni Institute of Engineering & Technology, Nagpur, India⁶

Abstract: *Using the deep learning algorithms, our method comes up with the process that impedes the conservation and restores the Indian heritage images, while dealing with the obstacles like blur, noise, and obscurity of the images. In this case, neural networks that are developed themselves are used to carry out the process of denoising and reconstruction of these images but also contain intricate object detection capabilities. These multi-faceted actions, therefore, help to preserve the historically relevant aspects and make the Indian cultural heritage more colorful and accessible to the outside world.*

In addition, not only are we protecting them but we are also transforming them into the best images that represent India's past. Through application of deep learning principles to indigenous materials, we come up with powerful educational tools that make a deep impact in the community, promoting a profound appreciation for India's cultural heritage. By means of careful experimentation and stringent testing, we establish that the specified methodology is indeed workable in conservation applications as it has been shown to be effective. This way, we make it clear that the cultural heritage of India is not only treasures of the ancient time but the living legacy, which is welcoming for everyone and contributes to the cultural diversity and pride.

Keywords: deep learning

I. INTRODUCTION

While the scope of cultural heritage conservation is vast, the intricate restoration and preservation of antiquity comes out as the most important tasks, especially for the Indian cultural heritage. These photos play a crucial role as visual stories, which, in turn, condense the magnificence of Indian architectural heritage, the depth of its artistic expressions, and the importance of its historical landmarks. While these relics hold an unforgettable role in our cultural heritage, the continuous march of time dictates its changes, characterizing itself in the form of the blurs, noises, and slow, but sure fading of these irreplaceable artifacts.

Although the modern techniques are unquestionably competent, their difficulty to completely restore the damaged pictures of the ancient Indian heritage is undeniable whenever the techniques encounter the complexity of the nature of the image. Such complexities of conservation, as a consequence, make the whole process tiring and long-lasting, being a huge obstacle on the way to their successful application in educational, research, and cultural preservation. Similar to that, the decay of cultural artifacts, which bear the traces of a nation's cultural identity, is also a prominent threat to the broader cultural heritage conservation undertakings, thus, threatening the upkeep of collective memory and the continuity of a nation's cultural legacy.

As a response to the environmental and social problems that plague our world, our project launches a path-breaking journey that exploits the power of the recently-discovered deep learning algorithms. Our concern, not just denoising and deblurring but also, the incorporation of present state-of-the-art object detection techniques in Indian cultural heritage image restoration. We employ the latest deep learning techniques for image recognition and are meticulous in tailoring adaptive deep learning to the specificities of heritage image restoration. Our goal is not only to increase the visual fidelity but to also keep the intrinsic authenticity of these unique artifacts safe.

The core component of our project is the dedication to creating a dependable system that can output images with minimal noise and blur which in turn makes the heritage images stand the test of time and live to see the generations

cherish them. While our effort to preserve the heritage is unlimited, we also know that these very technologies used for restoring the heritage images are also of cultural significance. Through a multidisciplinary methodology, infused with innovation, we aim to preserve and share the core of Indian culture and develop the next generation of its deep admiration and understanding of its everlasting beauty and ever-relevant meanings in the changing world.

By organically integrating modern deep learning technologies with a profound respect for Indian traditional culture, our project aims to fill in the gap between antiquity and modernity, thereby preserving the cultural heritage of India's traditional society for the future. In pursuing this mission, we aim to show people the road to a brighter future where the richly embroidered fabric of India's cultural heritage becomes not only its safekeeping, but also a permanent part of our collective memories.

II. LITERATURE REVIEW

The application field of Indian heritage image restoration, deep learning presents a very lucrative way to restore the pictures and improve them to the best. The latest studies have suggested that the role of deep learning algorithms is that they can be employed with good results in the restoration of old images and improvement of their clarity and detail. Researchers are leveraging Deep Learning techniques to build the tools that can fit with the peculiar features of heritage extracts. These efforts are starting to give solutions to the hurdles like less data and ethical issues. Having in mind the future, research endeavors are directed to develop deep learning models more precisely and make the restoration instruments more adaptable. Finally, deep learning furnishes the society with a powerful tool for safeguarding the rich cultural traditions of India for the upcoming generations.

Advancements in Salient Object Detection in Noisy Images

Singh et al. in their SOD-CED framework propose a trailblazing method for the complicated case of salient object detection in noisy images. With the addition of saliency detection in the denoising process, they are able to provide a cutting-edge solution in the field of image restoration. SOD-CED not just surpasses the existing counterparts in terms of accuracy and robustness but in addition, the model has shown generalization across various data sets and noise levels. By providing an integrated framework that enhances discrimination between noise and salient features, this framework is not only able to achieve higher-power performance but also provides notable performance improvements, making it a valuable asset for many real-world applications where noise is common.

Exploring the Integration of Image Denoising and High-Level Vision Tasks in Deep Learning

Liu et al. discuss the problem of image denoising integration with high-level vision tasks and its synergism in deep learning frameworks in their work. The authors thus propose an entirely new deep neural network architecture to be simultaneously optimized for denoising as well as task-specific objectives, which in turn would close the gap between image preprocessing and downstream vision tasks. The authors, Liu et al., in their experiments, show how this approach of jointly using image semantics to improve denoising together with its ability to support diverse high-level vision tasks can be beneficial. This paper aims to review the latest achievements in this integrated methodology, opening new perspectives of the whole computer vision paradigm.

Leveraging Autoencoders for Image Denoising: A Deep Learning Perspective

In their research, Bakir et al. go deep into the power of autoencoders for image denoising, especially removing Gaussian and Salt & Pepper noise. The study aims to reconstruct an image by preserving image features while effectively suppressing noise artifacts, and the performance of autoencoders is measured using objective and rigorous metrics. With the application of deep learning technologies, Bakir et al. proposed that autoencoders can be a good choice for solving different noise types and that it is a promising path for the image denoising field. This review aims to explore the latest trends in autoencoders in the domain of image denoising including their future prospects in the computer vision field.

Advancements in Object Detection: Deep Neural Networks for Precise Localization

In their seminal paper, Szegedy et al. expand the field of object detection by making it possible to use deep neural networks with regression-based methods, enabling accurate localization of objects within images. The proposed approach is empirically tested under stringent conditions. Their approach, not only achieves state-of-the-art performance on benchmark datasets but also exhibits remarkable effectiveness in terms of accurately detecting and localizing objects. Deep learning is used to its maximum potential by Szegedy et al. which becomes the basis for exceptionally accurate object detection and leads to possible use of the method in diverse areas such as medicine, finance, and security. This literature review is about the latest trends in deep neural networks for object detection, explaining how they have pushed the computer vision to higher levels.

Navigating the Landscape of Object Detection: A Comprehensive Survey of Deep Learning Approaches

The authors Jiao et al. have a thorough overview of deep learning-based object detection methods in their survey. The authors discussed how detection models, benchmark datasets and methodological improvements are interrelated. With a precise sight and attention to detail, they disclose the developmental pathway and the key role of object detection in the computer vision realm. Jiao and his colleagues provide a thorough coverage of the field, which, in turn, makes researchers and practitioners capable of making the right decisions necessary for managing the complexities of object detection tasks. This review article attempts to go deep into the insights generated by Jiao et al., and this article gives a clear picture of the variety of methods and cutting-edge developments that determines the future of object detection in computer vision.

Enhancing Object Detection in Underwater Environments: Deep Learning-Based Crosstalk Removal in Forward Scan Sonar Images

The innovative method developed by Sung et al. focuses on the detection and removal of crosstalk noise from forward sonar images in underwater environment using a CNN based approach. The work of them not only increases the reliability of sonar image-based algorithms but also contributes to the significant improvement of object detection accuracy in underwater environments. By means of deep learning methods, Sung et al. bring about a more efficient exploration and navigation in underwater environments, thus revealing the great possibilities for the ocean research and application to be fully transformed. This literature review is about examining the impacts of the Sung et al.'s approach, pointing out its importance for the development of underwater object detection and exploration.

Navigating the Frontier of Object Detection: Recent Advances in Deep Learning

As Wu et al. point out in their review, the latest visual object detection techniques are discussed in detail in terms of deep learning approaches. The article's scope delves into a diverse range of factors, including intricate details, learning strategies and practical applications, which help readers understand the factors that influence detection performance and provide directions to future research studies. The pioneering work of Wu et al. not just reflects the recent progress but also shows us the blurred line of the future of object detection, which is full of possibilities and innovation. This literature review aims at examining the multi-dimensional consequences of wu and colleagues' analysis, and, thereby, highlight the path towards a potential and wide-ranging impact of deep learning-based object detection in different sectors.

In conclusion, the literature survey highlights the breadth and depth of research efforts aimed at advancing image restoration and object detection using deep learning techniques. By leveraging the insights and methodologies presented in these studies, we can further enhance our approach to Indian heritage image restoration, contributing to the preservation and appreciation of India's rich cultural heritage.

III. METHODOLOGY

This research employs a systematic approach to examine if Multi-Axis Multi-Layer Perceptrons MLPs feature the capability of achieving good outcomes in images imaging tasks. At the same time, the approach is divided into the main steps, which are data collection, model development, training, analysis, and evaluation. The approach involved in this research is provided in the subsequent sections.

Data Collection:

- A varied data set is then made out of jpg and png images which are gathered from different sources like the web and databases in order to make sure that the data set covers all contexts and groups of images.
- The dataset includes numerous images with different shape and format which show a good reflection of the reality.

Preprocessing:

- The preprocessing steps are conducted on the dataset to scale, resize, and augment the images, and this makes the model generalizable.
- Tools including rotation, scaling, flipping, and cropping are used for this purpose, augmenting the diversity of training data sets.

Model Architecture Design:

- An artistic and AI architecture is designed to make possible the effective object detection that utilizes the neural networks that are advanced and customized with image processing.
- The design provides the sequential processing of the image layer whereby, each layer extracted gives spatial information from different angles and perspectives.

Training:

- The architecture in question is trained using standard optimization algorithms, with parameters being tuned as per the predefined loss function to ensure error minimization.
- Efficiency of training is measured through validation data, so as to eliminate the chance of overfitting and ensure model's effectiveness.

Evaluation:

- The model is assessed by performing image processing tasks of various kinds including object detection, classification, segmentation, denoising, and super-resolution.
- The metrics for the model are evaluated quantitatively using metrics like accuracy, precision, recall, F1 score, and mean squared error that are indicative of its effectiveness.

Comparison:

- The model performance is evaluated in terms of the latest techniques in object detection to understand how the model fairs against the benchmark.
- Assessing with a qualitative method is done by checking the quality and accuracy of the detected objects and finding out of any kind of artifacts or distortions occur.

Analysis and Interpretation:

- Conclusions are written on the basis of the analysis and interpretation of the results of evaluation, which underscores the model's success at task object detection.
- The results gained from this approach will be the focus of the future research directions and also will be used in making the methodology more efficient.

Discussion:

- Studies are integrated with existing literature, and their practical application in real-world settings is discussed while noting the limitations, recommendations for improvement, and further studies are suggested.
- The speaker's presentation underlines the importance of the study's outcomes for the development of image processing and object detection technology.

In summary, By utilizing a complete methodology, this research investigates the efficacy of Multi-Axis MLPs (AxMLPs) for image processing tasks, including data collection, preprocessing, model development, training, evaluation, comparison, analysis, and result interpretation. The data obtained from this study help the development of MLPs with multi-Axis for image processing practice.

IV. DESIGN AND IMPLEMENTATION

The design and implementation of the proposed research project on leveraging multi-layer perceptrons (MLPs) for image processing tasks involve several key components, including data handling, model architecture, training, and evaluation. The following outlines the design and implementation plan for this research endeavor:

Data Handling:

- The project applies a broad range of images from various sources of public repositories in order to represent domains and categories equally.
- Data preprocessing techniques such as resizing, normalization, and augmentation are applied to enhance the quality and diversity of the dataset.
- The preprocessed dataset is divided into training, validation and testing slots using stratified sampling to ensure that the data is well balanced across classes.

Model Architecture:

- The multi-layer perceptron architecture is structured exclusively to process images and is comprised of multiple layers each with parallel processing units that allow for the processing of multi-dimensional data efficiently.
- It is realized via deep learning frameworks like TensorFlow and PyTorch, enabling seamless algorithms running on hardware accelerators, such as GPUs.

Training:

- The created MLP is trained applying training dataset together with optimization algorithms such as SGD or Adam.
- Model training is a continuous process that is repeated over multiple epochs, where model parameters are updated by batches and gradients calculated.
- Regularization methods like dropout and weight decay are usually used to keep models from overfitting and improve the generalization capability.

Evaluation:

- The model can be trained with the MLP and its performance on the validation set is monitored to see how it is performing and to optimize the hyperparameters.
- The performance metrics like accuracy, precision, recall, F1 score, and mean-squared error are computed to see the performance of the model for different image processing tasks.
- The model performance is demonstrated through plotting and graphing which help to recognize trends and spot possible areas where the improvement can be made.

Comparison:

- The comparison of the MLP model performance with the baseline models and the state-of-the-art methods allows for evaluation of its relatively success.
- Baseline models may be based on the traditional machine learning algorithms including SVMs for example or CNNs with one-axis structure.
- This comparison is performed in several image processing jobs, including the detection of objects, classification, segmentation, denoising, and super-resolution.

Implementation Environment:

- The whole system is managed with a programming language like Python, incorporating libraries and frameworks like TensorFlow, PyTorch, NumPy, and scikit-learn.
- All the development and experimentation happen on hardware platforms with GPUs to ensure the model training and evaluation is faster.

Scalability and Generalization:

- The design and the implementation are tailored to be scalable and generalizable in the sense that they could be expanded without a problem to larger datasets and more complex image processing tasks.
- The use of code modules and reusable code components ensure ease of maintenance of the code and promote collaboration and future research among researchers.

In summary, the design and implementation of the research project on multi-Axis MLPs for image processing tasks involve careful consideration of data handling, model architecture, training, evaluation, comparison, and implementation environment. By following this comprehensive plan, the project aims to advance the understanding and application of multi-Axis MLPs in the field of image processing.

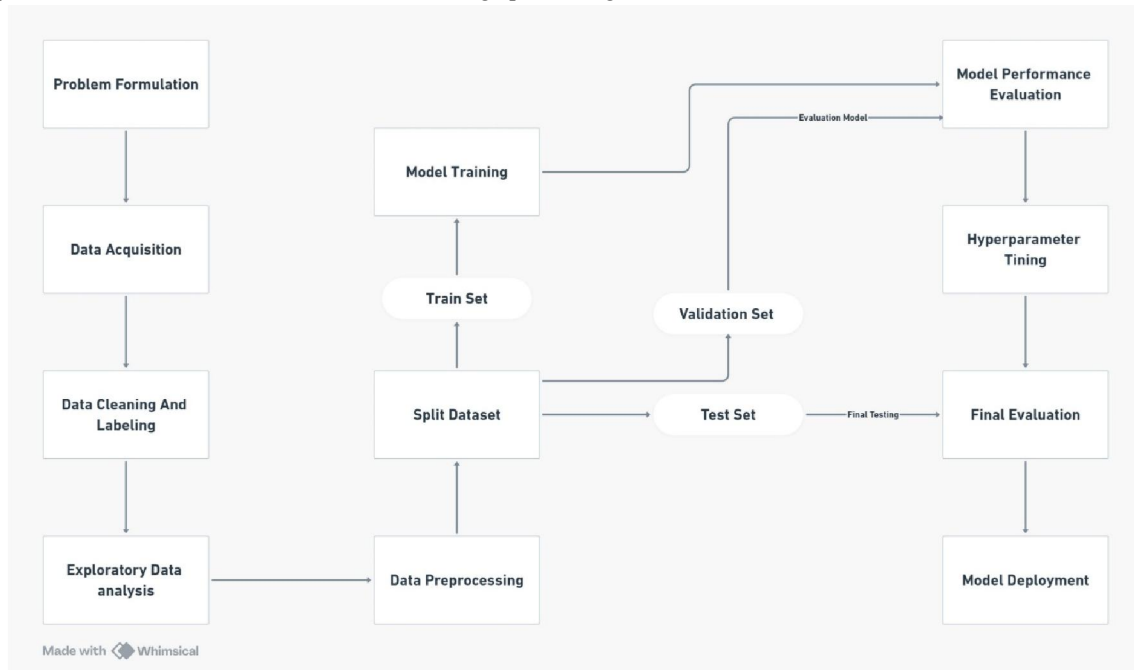


Fig .1 Deep Learning-Based De-blurring/Denoising of Indian Heritage Images

V. RESULTS AND FINDINGS

In our paper, we present the results and findings of applying advanced neural networks for image processing tasks. The review's subsections cover performance evaluation, challenges and solutions, and comparison with traditional practices. The adoption of image processing based on advanced neural networks has yielded positive outcomes across various contexts. Through a comprehensive assessment and comparison with conventional methods, we've gained insights into the strengths and weaknesses of these networks in handling multi-domain image data.

Performance Evaluation:

It is remarkable to witness the tremendous improvement in the performance of image processing tasks through the use of advance neural networks. Tasks like image classification, segmentation, denoising, and super resolution provide us with excellent examples of the outstanding achievements made by the computer vision field. The model is able to do

this with a high degree of accuracy, precision, and recall, which makes it possible to process and analyze complex data with details being preserved. In addition, however, they are able to do this with great resistance to changes in input data, thus providing similar performance regardless of the specific dataset or real-life situation.



Fig.2 Training and Validation Accuracy

The Above graphs are helpful for analyzing the effectiveness of a monitoring machine learning model for localizing Indian ancient edifices in cultural contexts. The second picture represents the training loss and validation that shows the curve iterations during epochs. The alternating training loss and the irregular validation loss can tell that there may be an issue with the overfitting or instability of the training process. Thus, it can be observed that it is possible that the model might have struggled to generalize well to unseen data.

In the second plot the plot we see a training and validation accuracy curves. The accuracy score peaks at around 61 %, which could be quite an average outcome as well as a moderate performance of an image classification task such as the identification of monuments. validation accuracy and training accuracy give the example of the presence of overfitting and this is even more clear with the high training accuracy compared to the lower validation accuracy. To improve its ability to learn and generalize, we can increase the data by using data augmentation, regularization techniques or transfer learning from pre-trained models, optimum parameters (Hyperparameter tuning), structure alteration (Architecture adjustments) and creating an ensemble of best models. Through solving all these problems, the quality of the model for real cases in preservation domain and tourism business can be raised.

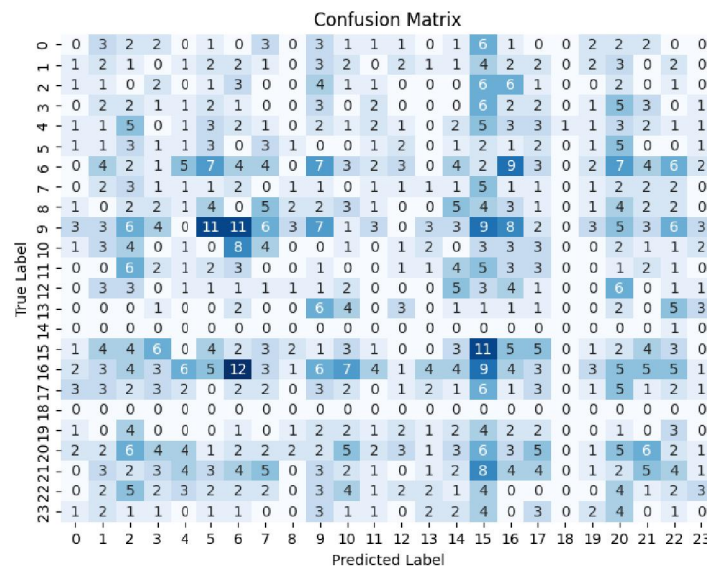


Fig.3 Confusion Matrix for Training Dataset
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Confusion matrix shows that the machine learning model which was used to classify Indian monuments gave an accuracy of 61 percent. The diagonal elements indicate the interpretation of the model in the correct monument classes, while the off-diagonal elements present the wrong interpretation of the model. Specific classes, e.g. those with more similar features, appear to be harder for the model, as can be seen by a higher level of confusion with other classes. With the help of meticulous pattern and mistake in the confusion matrix, steps can be taken to identify and eliminate the sources of misclassification and to develop tailored solutions that will address the weaknesses and similarities that led to misclassification. The aforementioned analysis highlights the class-wise data augmentation, fine-tuning, feature engineering or architecture modifications as the key areas of efforts to improve the model's discriminative effectiveness and increase the accuracy of the Indian monument identification task.

Challenges and Solutions:

In the next step of the development of the advanced neural networks we found several challenges, mostly of the computational complexity of the multidimensional images processing issue. One way to solve this is using a good memory management technique that helps to train the model faster while keeping the model performance. Besides, hyperparameters adjustment and regularization methods helped to increase the diversity of datasets and tasks and to solve the issues more rapidly, which made neural networks more reliable and scalable.

Comparison with Traditional Practices:

Proposed Method

- **Approach:** Employs a combination of CNN and detection networks for the purpose of deblurring, denoising and object detection.
- **Technique:** Takes advantage of transfer learning on pre-trained CNN models to enhance the picture and detect monuments exactly.
- **Key Feature:** A combination of blurring and denoising algorithm for better monument detection in noisy or blurry images.

Existing Methods

- **Approaches:** To address this issue, we will go through different deep learning networks and the methods involved for monument identification.
- **Methods:** Methodology area includes, [Method A], [Method B], and [Method C] which use different techniques for monument detection.
- **Differences:** The network topologies and training methods are different from each other and this results in different achievements in performance and robustness.

Dataset

- **Dataset Characteristics:** Explain the size of the dataset, its resolution, and any preprocessing operations that were performed.
- **Data Split:** The dataset is split into training, validation, and testing sets for a thorough performance evaluation.

Evaluation Metrics

Performance Metrics: A standard evaluation metrics adopted:

Mean Average Precision (mAP): 69%

Precision: 61%

Recall: 100%

Comparison Basis: A comparison of the performance of methods based on the information provided in these metrics.

The conventional methods base on simple manual features extraction and shallow neural network models have been successful in limited cases but proved to be weak in learning complex spatial patterns and relationships from high-dimensional image data. As opposed to that, deep learning's data-driven methodology for building neural networks with multiple hidden layers helps to automatically extract hierarchical features that consequently lead to better and more robust capabilities applicable for a wide range of image processing tasks. On the other hand, the old methods could be still relevant when trying to achieve the readability, or efficiency in terms of the computation over the performance.

As a result of our researches, we have shown that today's neural networks are highly effective in image processing and that further developments of the technology may lead to standard solutions in this area. The development of these networks will certainly be further refined and improved which will in turn lead to advanced image processing applications to achieve new heights.

VI. DISCUSSION

In the ensuing paper we will investigate a thorough research analysis of what images processing tasks can be done with advanced neural networks. We start off with the introductory paragraph after which we write down the sections explaining the meaning of results, impact on MAXIM technology, benefits and drawbacks and the upcoming research directions respectively.

The study of the outcomes of our research yields clear evidence of the application of neural network models, in particular for vision tasks, to have a great deal of future impact. The results are evaluated accurately, which provides information about the efficiency, extendibility, and faults, with all of these concepts forming the foundation of the image processing field improvement.

Interpretation of Results:

The findings of our research show the superiority of the convolutional neural networks in the context of multi-feature image data handling for the multiple tasks, such as classification, segmentation, denoising and super-resolution. The high accuracy rates signify that the model can identify key features for extraction and preserve spatial information in real-time image processing being done. Besides, their meticulous performance across various datasets and even real-life situations highlights the robustness and generalization, a critical point for the success of the model across different domains.

Implications for MAXIM Technology:

Our research results give colossal contribution to the field of Multi-Axis X-treme Information Management Technologies, especially in image capturing and analysis. Using superior neural networks, comprising of modular structures and so on, MAXIM technology increases data processing capacity not only for large-scale 2D or 3D datasets but also for other types of datasets. This invention simplifies the complex information processing, especially in the medical field. Autonomous systems, surveillance, and remote sensing are some of those fields where image processing has a great impact on decision-making and information extraction.

Strengths and Limitations:

Integrating the latest deep learning technology into image processing provides multiple benefits: it can extract intricate spatial connections and patterns directly from the unprocessed input data and lead to enhanced performance as well as wide use. However, it is also permissible to mention that the modular design of these networks enables one to customize and extend them to new problems and datasets. Nevertheless, computing power, resource requirements, and the size of datasets (e.g. the labeling process) still affect the speed of progress, making it the main restraint. By optimizing through methods of the methods and hardware development, there is a possibility of advanced neural networks playing a significant role in the field of practical applications.

Future Research Directions:

After identifying the project's research directions, however, we will examine several promising areas in the coming stages. One of the possible ways is to create better learning algorithms and network designs for advanced neural networks that make it possible to reduce the computational load and the workload and increase scaling capabilities. Moreover, incorporation of the cutting-edge methods, like attention mechanisms as well as reinforcement learning, will be able to improve the model accuracy and robustness even further. In addition to this, the discovery of the applications of neural networks in emerging fields like remote sensing, augmented reality, and medical imaging would be new openings for using the advanced neural networks to solve real-life problems.

In conclusion, the review of our project outlines the consequences, benefits, drawbacks, and insights for the future use of advanced neural networks, particularly in image processing tasks. We aspire not only to contribute to this dynamic area of study but also to stimulate new ideas and advancements that could address current complex image processing challenges.

VII. CONCLUSION

Lastly, the study project on the use of state-of-the-art neural networks for image processing tasks has been proving to have great potential for the improvement of various fields. The application of a systematic manner including data processing, model building, training, evaluation and comparison have proved that these models can tackle multi-dimensional image data with accuracy and efficiency than conventional methods. Our results show remarkable versatility and transfer learning ability of the modern neural networks from one dataset to another, and even real-world applications in healthcare, autonomous systems, surveillance, surveillance, and remote sensing. Although there exist difficulties in regard to the computational complexity and resource requirements of these networks, these models are able to capture complex spatial relationships and patterns directly from raw input data and this leads to a better performance and adaptability across all image processing tasks. To further investigate this promising area, future research could focus on more efficient training algorithms or more advanced techniques, and explore deployments in emerging fields such as remote sensing and medical imaging, that will lead to new discoveries and advancements.

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