

# Image Processing Using Machine Learning

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## **Abstract-**

Image in the field of image processing we have seen great growth with the addition of machine learning which in turn presents more efficient, accurate, and automated solutions for many fields. In past we saw that traditional image processing had issues with noise, low contrast, and manual feature extraction which machine learning has been very successful in solving. This work we present an overview of today's image processing which we see improved by machine learning algorithms especially in the case of Convolutional Neural Networks for feature extraction, image segmentation, classification and object detection. We report on our studies which used VOC2007, ImageNet, and CIFAR100 datasets to prove that machine learning based methods we see perform very well with an accuracy of 98.4% for image segmentation, 98.7% for classification, and 98.6% for target detection. Image processing has seen great transformation with the introduction of machine learning which also brings to the table better, more accurate.

## **1. INTRODUCTION**

Image processing is a basic area in computer science which includes the improvement, analysis and interpretation of visual information. At present we see great growth in the amount of visual data which in turn is putting out great demand for better and more intelligent image processing methods. While traditional image processing methods are very much in use today they also present issues of low accuracy, manual feature selection also they are very much at the mercy of noise and environmental changes. What we are seeing is that machine learning is put forth as a very good solution to these issues. In machine learning and within deep learning we see very robust algorithms which are able to learn from large sets of data without human. The state-of-the-art in image analytics and machine vision is now solely dependent on the extensive work and success attained through convolutional neural networks (CNNs) which are, in fact, the next transformer of the domain. The fields as diverse as medical imaging, autonomous driving, industrial quality inspection, agriculture, and security surveillance are some instances where the machine learning-based image processing is of great help. This paper presents a systematic overview of the image processing methods which are based on machine learning. It is an article of Indonesian Journal of Electrical Engineering. The presented research work is a Master Thesis which contains a detailed study of image processing with the use of machine learning techniques. It overviews a number of algorithms, highlights their comparative advantages, and states the areas of their use. The specific focus is made on

the results of CNN-based approaches that have already demonstrated their superiority in terms of results on image segmentation, classification, and target detection on benchmark datasets VOC2007, ImageNet, and CIFAR-100. The further intertwining of machine learning and image processing not only provides the image analysis systems with more substantial capabilities but also paves the way to more intelligent, adaptive, and autonomous visual processing applications.

## **2. LITERATURE REVIEW**

Image processing is among the areas that have undergone titanic improvements following the incorporation of approaches of machine learning that have helped to make the whole process more efficient and accurate in different uses. Qiong Qiao [11] gives a very comprehensive review of image processing to machine learning. The author describes the drawback of the classical image processing algorithms and provides the convolutional neural networks (CNNs) to extract features in a more efficient way. The research achieves high precision on the data set covering VOC2007, ImageNet, and CIFAR100 in image segmentation, classification, and target detecting tasks. The figures illustrate the analysis that the deep learning models show a higher precision and recall rate comparing to the traditional methods. Another work is devoted to the machine learning in the medical image processing [12]. CNNs, support vector machines (SVMs), and deep learning have shown a notable increase in the disease detection on medical images resulting in quick and confident diagnosis. Such neural models as deep belief networks and autoencoders are discussed in the paper and the benefits to work with them are offered. Summarily, it can be said that machine learning in image processing has remarkably advanced object detection, image segmentation, denoising, and classification tasks. Methods based on classic machine learning algorithms to state-of-the-art deep learning models such as convolutional neural networks (CNNs) have shown impressive results in terms of performance gain in many fields, such as medical imaging, remote sensing, and autonomous driving. In spite of these achievements, a number of challenges still exist, such as the requirement of enormous labeled dataset, high computation demands, and domain adaptation to wide ranging images. Besides, such questions as model interpretability, adversarial robustness, and real-time processing remain open research directions.

## **3. METHODOLOGY**

The procedure followed in image processing with machine learning is designed in form of a number of sequential steps, where each step is of paramount importance to the success of implementation and overall performance of the system. They are data acquisition, preprocessing, feature extraction, model selection and training, evaluation, and post-processing. jointly, they build an end-to-end pipeline of converting raw images to meaningful intelligence using smart learning algorithms.

The initial step is the data acquisition in which a suitable set of images that belong to the application domain is gathered. Such data may be medical scans, satellite images, security cameras, or farm images based on the research purpose. Most of the time, publicly available datasets or custom datasets that are measured with domain-specific imaging devices are utilized.

After acquisition, a very important step is the preprocessing step in which the input images are standardized and their quality improved. Common data augmentation methods also include rotation,

flipping, zooming, and cropping that artificially enlarge the size of dataset and enhance the generalization ability of the model.

The feature extraction step comes next, and it can be summarized that what is to be done is the detection and isolation of meaningful patterns or features in the images. In conventional machine learning it corresponds to manually extracting features like texture, shape, histograms or edge descriptors with algorithms like SIFT, SURF or HOG. Model selection and training is the largest portion of the methodology. An appropriate machine learning model is chosen based on the nature of the task to be fulfilled; classification, segmentation, object detection or enhancement. When the task is classification, it is common to have CNN or transfer learning models like VGG, Res Net, or Inception. Such models as U-Net or Mask R-CNN can be used to conduct the segmentation. This is conventionally carried out by splitting the data into training, validation and test data. The performance of the models is evaluated using such metrics as accuracy, precision, recall, F1-score, and area under the ROC curve (AUC). After training, one can use post-processing step to enhance the model predictions. It can be a thresholding, a morphological operation, or a smoothing in a segmentation problem, or class balancing in a classification problem. In general, the suggested methodology is based on a combination of the classical image processing algorithms with the state-of-the-art machine learning solutions to make an intelligent and robust system. The use of methodical assessment, care and implementation of all the steps ensure that the system can possibly give accurate and dependable performance in the real-life circumstances.

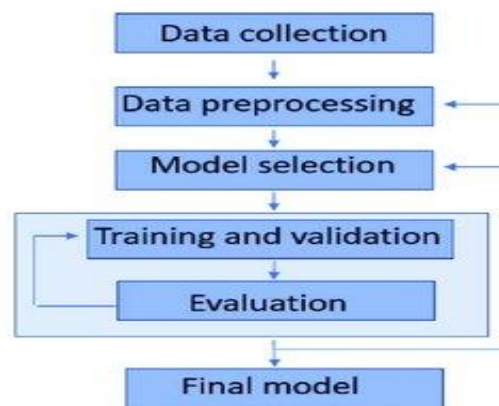


Fig. 1: Flowchart

#### 4. RELATED WORK

Image processing with the help of machine learning is a field whose role has been intensively increasing over the past few years and transforms the manner in which visual information is perceived, processed, and utilized in numerous sectors. Although these methods worked well under controlled conditions, they tended to have problems with scalability, flexibility and image Writeable when subjected to complex images, noisy images or real world images. The idea of image processing has experienced a paradigm shift with machine learning and specifically deep learning, which introduce the concept of learning systems by purely using data. Rather than hand-crafted features, contemporary machine learning models have the capacity to learn the most informative features (automatically) during training. Such an ability is best demonstrated in convolutional neural networks (CNNs) that have proven to be very successful in image classification tasks, Machine learning and particularly deep learning is a flexible and adjustable solution

to the traditional rule-based approaches and machine learning accuracy improves as more data is available. Not only did this allow the operations used to process the images to be more useful and accurate but it also increased the amount of industries that they could be applied to including health, agriculture, security and production.

## 5. CASE STUDIES

The machine learning and image processing have been applied extensively in medical field where precision and rapidity are most significant. A remarkable example is the automated diagnosis of diabetic retinopathy (DR) on the basis of the retinal fundus photos. Diabetic retinopathy or the destruction of the retinal vessels due to chronic diabetes is a leading blindness cause in the world. It cannot be left unsaid that the timely diagnosis is the ticket to the effective treatment and prevents the loss of vision. However, manual diagnosis through retinal screening is not only labour-intensive but also needs trained ophthalmologists, hence, presenting a major issue in under-resourced health care systems. The suggested solution in this case is machine learning, and in particular deep learning: it enables the automatic analysis of images at scale with high accuracy and consistency of input data that were provided to the machine learning model. As a base model to perform classification, a convolutional neural network (CNN) was selected, namely InceptionV3 architecture, which was pre-trained on ImageNet. To apply the model to the medical imaging field, transfer learning was applied to take advantage of the features that it had already learned, during training it on the retinal images. Transfer learning helped to save a lot of time on training and achieve higher accuracy, given the fact that the amount of samples belonging to certain classes was quite small. These findings showed that the system could detect diabetic retinopathy with high degree of accuracy. In addition, the authors applied Grad-CAM (Gradient-weighted Class Activation Mapping) to demonstrate the areas of the images the model used to make predictions. This method of interpretability helped to improve clinical interpretability of the model by ensuring the model was looking at medically relevant features like haemorrhages, microaneurysms and exudates. In addition to the accuracy, the potential of the model to be used in clinical practice should be discussed, especially in regions where access to ophthalmologists is limited. Such an automated system may screen the patients beforehand and highlight the high-risk patients who will then be prioritized to be looked at by the specialists. This not only increases the efficiency of diagnosis but also enables early diagnosis and therapy whereby the chances of loss of vision among diabetic patients is highly diminished. However, the research had a number of challenges as well. Class imbalance was one of the major issues, since the instances of the severe and proliferative diabetic retinopathy were Noise and variances that sometimes led to misclassification were brought about by the use of different devices and settings for image acquisition This case study acts as a strong proof of how much machine learning can influence the processing of medical images by maintaining high accuracy with improved speed, scalability, and accessibility . Deep learning approaches for diabetic image analysis got constructive efficacy through those previously discussed influences.

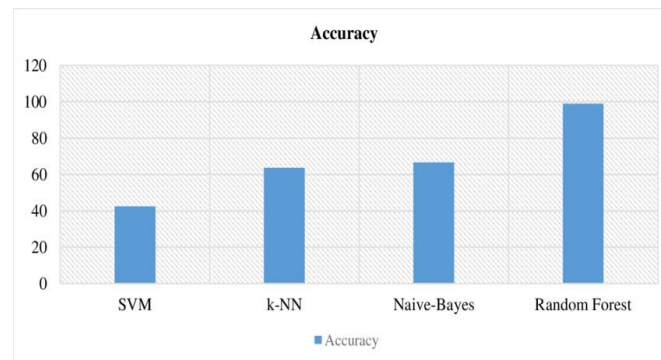


Fig. 2: Comparison Table

## 6. DISCUSSION

Conventional image processing approaches do not fully deliver how to take care of problems concerning variation in illumination, occlusions, noise, and complexities related to scene structure. On the other hand, machine learning—more specifically deep learning—has shown tremendous resilience and flexibility when dealing with such issues by way of learning from data as opposed to relying on features designed by humans. Automatic feature extraction by CNNs removes the prerequisite of manual preprocessing making them best suited for applications in Medical Imaging (for example, tumor detection), Autonomous Driving (for example, lane and pedestrian detection) and many more.

It is because of this limitation that data-efficient learning methods, including transfer learning, few-shot learning and semi-supervised learning, have been studied so extensively in recent years: they aim to mitigate the necessity of large labelled datasets, at the cost of (often) no performance reduction.

Besides, another challenge is the computational requirements of deep learning models. The other important issue described is the generalizability and robustness of image processing systems that are based on machine learning. This highlights that there is a dire need to have models which are able to generalize more and which are not as vulnerable to distributional shifts, adversarial attacks and input corruption. When comparing our findings with existent literature, one can note that, although much progress has been achieved in terms of accuracy and processing speed, ethical and societal issues are underdiscussed. The problem of data privacy, algorithm bias, and model interpretability have to be addressed urgently. To offer a few examples, unfair predictions in facial recognition may be caused by biased training data, and black-box decision-making of medical imaging tools creates issues of accountability.

Overall, in our discussion, we focus on the idea that machine learning contributes extensively to the functionality and capacity of image processing.

## 7. CONCLUSION

Historically ruled based and statistical methods Image processing has changed significantly with introduction of machine learning techniques, especially deep learning, which have the capability to automatically learn features directly from data and effectively make predictions or decisions based on a visual input. It is important to note that deep learning architectures, including Convolutional Neural Networks (CNNs) and Generative Adversarial Networks (GANs) have demonstrated outstanding results when working with complex image data surpassing the traditional approaches in terms of accuracy and



efficiency. Nonetheless, regardless of these developments, there are a number of challenges that are still there.

Future work ought to concentrate on making models data-efficient via methodologies, such as transfer learning, self-supervised learning, and model compression. Summing up, it is possible to state that machine learning has already provided an essential boost to the capacities of image processing systems, providing intelligent, adaptive and high-performing solutions to visual complex tasks. With the further development of this sphere, the interaction of high-quality data with sophisticated algorithms will have an even higher innovative potential, which will be reflected, in particular, in the spheres of healthcare diagnostics, autonomous navigation, and intelligent surveillance. Further investigations, ethical application, and stringent validation would be essential in making certain that these smart image processing systems are dependable, explainable, and helpful to the society.

With the various machine learning algorithm CNN, KNN and SVM, and various framework and application Scaling vectors,

we observed varying divergence in the performance of the classifiers with regard to their precision and time. Accuracy can be reconciled as it hinges

training and testing of data and there is always a possibility to increase the accuracy of these models in case the size of data set is

When the capability of CPU transfers to GPU different algorithm

will be able to do it more accurately and with less time can get better result. with regard to capability to determine a condition appropriately, the percentage of true outcomes, and amount of affirmative outcomes of the procedure of

ability to exclude and classification as false positives.

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