



## **Advancements in digital image processing technology and practical applications**

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### **Abstract**

This literature-based study explores recent breakthroughs in digital image processing, highlighting technological innovations and their practical implications in several areas. The study utilised a conceptual and desk-based methodology, integrating current literature to offer a thorough understanding of the theoretical underpinnings and practical ramifications of contemporary image processing techniques. The research analyzed significant advancements, including improved picture-enhancing algorithms, real-time object detection, deep learning integration, and noise reduction strategies, utilizing a diverse array of scholarly sources. The study revealed the increasing impact of digital image processing across sectors like healthcare, security, remote sensing, entertainment, and autonomous systems. It provides significant insights but also recognizes limitations inherent in literature-based research, such as the lack of empirical confirmation and possible gaps in the latest findings. The study consolidates recent progress in digital image processing, connecting theoretical achievements with practical applications to facilitate continued study, innovation, and deployment across diverse industries. The study concludes by outlining new patterns and proposing avenues for future research, especially in artificial intelligence-based image analysis and transdisciplinary applications.

**Keywords:** Digital image processing technology, image acquisition, segmentation, feature extraction, and digital image processing applications

### **1. Introduction**

Digital image processing technology involves using computers to remove image noise, enhance image quality, segment parts of an image, and recover details. Its growth has been fuelled by advances in computer networks, the rise in mathematical capabilities, and increasing demand across various industries. These factors have opened new opportunities and pushed the technology forward. The first known use of digital image processing dates to the 1920s, when photos were transmitted via cables between the United Kingdom and the United States. However, the image quality was quite poor, highlighting the need for improvement and optimization (Gonzalez & Woods, 2018), and could lead to wrong diagnosis (Picone et al., 2025) Major progress did not occur until the 1970s, when the technology was significantly updated and pattern recognition systems began to be integrated into image analysis. As

technology continues to evolve and expand into new fields, the expectations for digital image processing are also growing, driving even more rapid development in this area.

Although many studies have investigated techniques and applications in digital image processing, such as enhancement, compression, segmentation, and pattern recognition, there is a conspicuous absence of thorough, integrative research that correlates contemporary technological advancements with their practical applications across various sectors such as healthcare, agriculture, autonomous vehicles, and surveillance (Gendy & Patel, 2024; Mahmood, 2024). Most existing studies have focused mainly on the creation of technical algorithms or industry-specific case analyses (Sarker, 2021; Steen et al., 2021; Mahmud et al., 2022). This study bridges the scientific advancements with cross-domain practical applications, providing a cohesive perspective that aids both scholars and practitioners aiming to understand the wider implications of developments in digital image processing.

This study aims to critically analyze recent breakthroughs in digital image processing technologies and investigate their use in real-world contexts across different disciplines. This study synthesizes current literature to offer a thorough overview of developing methodologies, emphasize their practical significance, and identify areas for future research and development.

## **2. Research methodology**

This study employed a qualitative, literature-based research methodology to explore recent developments in digital image processing (DIP) and their practical applications in industry and higher education. The process was based on desk research and conceptual analysis, focusing on the identification, evaluation, and synthesis of reliable scholarly and technical sources.

### **2.1 Research strategy**

The study employed a descriptive and exploratory strategy intended to deliver a thorough overview of the cutting-edge technology in Digital Image Processing (DIP). This encompasses an examination of both the theoretical foundations and practical applications of current advancements in the subject.

### **2.2 Data sources**

Data were exclusively gathered from secondary sources utilizing the following databases:

- IEEE Xplore;
- ScienceDirect;
- SpringerLink;
- Google Scholar;
- Academic books; and
- Conference proceedings centred on image processing and computer vision.

The following search words were used to gather pertinent literature for the study:

- Digital image processing;
- Computer vision;
- Image enhancement techniques;
- Machine learning in image processing;
- Deep learning for image analysis;
- Real-time image processing;
- Applications of image processing;
- Medical image analysis;

- Remote sensing and image processing;
- Pattern recognition;
- Edge detection algorithms;
- Image segmentation;
- Object recognition;
- Artificial intelligence in image processing; and
- Image filtering techniques.

The inclusion requirements for sources were that they must have been published between 2015 and 2025 to maintain relevance. Literature that failed to satisfy the criteria was omitted from the study. Seventeen references were utilised in this study.

### 2.3 Data analysis

A thematic content analysis was employed to examine and integrate the gathered data. Key themes that emerged from the data were used to present and discuss the findings.

### 2.4 Ethical considerations

This study utilized exclusively secondary data; hence, no direct ethical approval was necessary. Throughout the research process, proper citation and adherence to the values of academic honesty were upheld.

## 3. Overview of digital image processing

Digital image processing ((DIP), also known as computer image processing, refers to the use of computer algorithms to perform image manipulation and analysis. It involves converting an image signal into a digital signal, which is given to the computer system for further processing. This includes tasks like enhancing images, reducing noise, segmenting objects, restoring image details, encoding, compressing, and extracting features. Figure 1 shows the overall process of digital image processing.

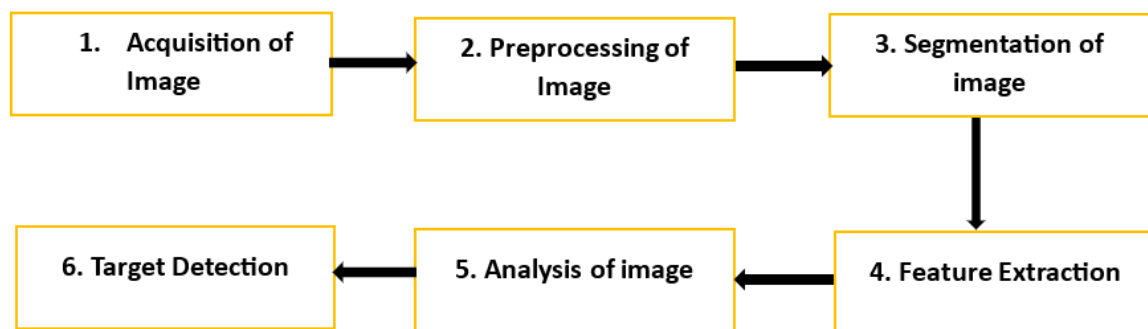


Figure 1. Schematic Representation of Digital Image Processing  
Source: Authors' model

DIP is a field that deals with the use of computers to process and manipulate digital images. It plays a key role in many modern applications, including medical imaging, remote sensing, computer vision, and robotics. At its core, DIP involves essential tasks like capturing, enhancing, restoring, compressing, segmenting, and recognizing images (Gonzalez & Woods, 2018).

A digital image is made up of a grid of pixels, where each pixel represents the intensity of light at a specific point. The quality and complexity of processing an image depend on factors like its resolution, color depth, and the model used to represent colors (Jain, 1989). To improve the visual quality or to extract useful information from images, enhancement techniques such as histogram equalization, filtering, and contrast adjustment are commonly applied (Castleman, 1996). Transformation methods like the Fourier and wavelet transforms are also widely used, especially for noise reduction and image compression, which are important for efficient storage and transmission (Pratt, 2007).

Segmentation, which involves dividing an image into meaningful parts, is another key process. Methods like edge detection, thresholding, and region-based techniques are typically used to identify objects and structures within images (Sonka et al., 2013). In recent years, advances in machine learning, particularly deep learning, have significantly improved traditional image processing. Convolutional Neural Networks (CNNs) are now used to automate complex tasks like feature extraction and image classification (Goodfellow et al., 2016). The combination of classic image processing techniques with AI has led to more accurate and efficient solutions across a range of fields, from facial recognition to autonomous driving and medical diagnostics (Alahmar, 2025).

#### **4. The key steps involved in digital image processing**

##### **4.1 Acquisition of images**

To begin with, in the process of image acquisition, that is, from an imaging standpoint, an image needs to be captured before any processing can take place. Depending on the type of imaging sensor, different methods are used. For example, standard TV cameras are used to capture visible light images, while infrared cameras are used to detect infrared images, which are especially valuable in military applications. Acoustic wave imaging is used to gather internal structural details of opaque objects by analyzing how sound waves travel through different materials. Similarly, X-ray imaging helps reveal internal shapes by making use of how various objects allow X-rays to pass through them.

In addition,  $\gamma$ -ray imaging is used to obtain information about how human organs function by detecting gamma particles emitted by isotopes, helpful in identifying both normal and abnormal conditions. MRI (magnetic resonance imaging) is used to monitor changes in organs based on how different substances respond to magnetic fields. These various techniques are often used to produce 2D images, which can then be turned into 2D or 3D representations through methods like tomography. The development and application of these imaging technologies have played a vital role in advancing fields such as medicine, the military, and industry (Luo et al., 2018).

##### **4.2 Image preprocessing**

Image pre-processing techniques are commonly applied to help better understand what an image contains, making it easier to analyze later. The main goal is for key points of interest in the image to be identified, which then supports more accurate feature extraction. At this stage, the work is done at the most basic level, where both the input and output are simple intensity images. During pre-processing, unwanted distortions are removed, and important image features are enhanced so that the image is clearer and more useful for the next steps in processing (Suguna & Devi, 2018).

### **4.3 Image segmentation**

Today, advanced technologies are being developed in the field of image processing, especially in image segmentation. This process is seen as both essential and complex. Image segmentation is often considered the most important stage of image analysis. Its main goal is to simplify the data, making it easier to analyze. (Garg, 2016). In this process, a digital image is divided into continuous, separate, and meaningful regions so that important features can be more easily identified and extracted (Abdulateef & Salma, 2021).

### **4.4 Feature extraction and analysis of image**

Feature extraction helps describe the important shape details within a pattern, making it easier to classify using a structured method. In image processing and pattern recognition, it's considered a type of dimensionality reduction. The main purpose is to pull out the most meaningful information from the original data and represent it in a simpler, lower-dimensional form. When an algorithm receives too much data, especially when much of it is repetitive or unnecessary, then the data is converted into a smaller set of key features, known as a feature vector. This transformation process is what we call feature extraction. If the right features are chosen, they can capture all the important information needed to complete the task, without relying on the full, complex data set. Today, pattern recognition, which relies heavily on feature extraction, is a rapidly growing area in image processing research (Kumar & Bhatia, 2014).

### **4.5 Target detection**

The main goal of remote sensing image target detection is to locate and classify important targets within these images, which plays a vital role in areas like intelligence, disaster relief, and industry. In recent years, advances in technology have shown deep learning algorithms to be a highly effective tool for image processing. These algorithms automatically extract important features from images, which greatly reduces the amount of manual work required. This is especially true for detecting targets in high-resolution, large-scale remote sensing images, where deep learning has replaced traditional methods that relied on manual algorithm design. As a result, the range of applications has expanded, and detection accuracy has improved significantly. With ongoing improvements in neural networks and the growing use of deep learning, processing remote sensing images has become faster and more precise. Researchers are now focusing on developing advanced deep convolutional neural networks to keep up with evolving environmental demands in target detection (Liu et al., 2023).

## **5. Key advantages of digital image processing**

The first known use of digital image processing dates back to the 1920s, when photos were transmitted via cables between the United Kingdom and the United States. However, the image quality was quite poor, highlighting the need for improvement and optimization. Major progress did not occur until the 1970s, when the technology was significantly updated and pattern recognition systems began to be integrated into image analysis.

As technology continues to evolve and expand into new fields, the expectations for digital image processing are also growing, driving even more rapid development in this area. The technology also supports high processing resolution, which depends on the number of quantization bits used, commonly 8, 12, or 16 bits, or even more, allowing images to be represented in detail.

Today, digital image processing has come a long way and offers several key advantages:

- Better reproducibility: Unlike traditional analogy methods, digital images do not lose quality during storage, copying, or transmission. (Chen et al., 2009)
- Wider frequency range: Compared to voice data, image data requires much more bandwidth, making it more complex to process. ( Kelly, 2024).
- Broad applicability: Digital images can be sourced from a wide variety of devices, from microscopes to telescopes, making the technology useful in many fields (Chen et al., 2011)
- High flexibility: If the image data can be expressed with mathematical formulas or logic, it can be processed digitally. (Chen et al., 2011)

## 6. Applications of digital image processing

Digital image processing technology has advanced remarkably and is now widely used across many fields, including remote sensing, aerospace, medicine, communication, industry, and agriculture (Desai et al., 2020). It helps enhance image quality, enables intelligent analysis, and reduces manual effort. In remote sensing, it's used to analyze satellite images; in medicine, it improves diagnostic imaging; and in agriculture, it supports identifying pests and diseases. It also plays a key role in multimedia communication, automated manufacturing, and facial recognition in public safety. Additionally, it aids in cultural heritage restoration, intelligent robotics, and environmental monitoring, offering precise and efficient data interpretation. Its continued evolution promises even broader applications (Liu et al., 2023).

Table 1: Applications of Digital Image Processing

Field	Application
Physics and Chemistry	Spectrum analysis
Biology and Medicine	Cell analysis, CT and X-Ray analysis
Environment Protection	Research of atmosphere, Ocean, Forest, etc..
Agriculture	Estimation of Crops and Plants
Irrigation works	Lake, River and Dam
Weather	Climate and weather report
Communication	Internet, Mobile, Telephone communication
Military	Artillery or missile detection and trainings

Source: Literature survey

## 7. Findings and Discussions

DIP has been recognized as a key technology across various fields, including medical imaging, remote sensing, computer vision, and robotics. Tasks that were once done manually, such as analyzing and interpreting images, are now being handled by computers, thanks to advancements in DIP. Through the use of computational techniques, digital images are being enhanced, refined, and transformed to reveal valuable insights and improve decision-making.

Its growing role is being driven by the increasing need for high-quality visual data in areas like healthcare (for diagnostic purposes), agriculture (for monitoring crops via satellite), and security (for facial recognition and surveillance). As the amount of image data continues to rise, DIP systems are being relied on more than ever as part of broader solutions in artificial intelligence and big data, where speed and accuracy are essential.



## 8. Limitations

This study is primarily based on secondary sources, which provides a broad overview of existing knowledge. However, incorporating original experimental data or case studies in future work could offer deeper validation of the findings. While the analysis explores key algorithms and models in digital image processing, there remains potential to expand the technical depth and further illustrate their real-world applications through practical implementation examples.

## 9. Recommendations

In the development of DIP, greater emphasis should be placed on interdisciplinary collaboration. The use of AI alongside traditional methods is encouraged, and real-time processing capabilities should be further improved. More support should be given to building quality datasets and updating educational programs. Ethical use and privacy must be safeguarded. Open-source contributions and knowledge sharing should also be encouraged to drive continued progress in the field.

## 10. Contribution to policy and practice

Digital Image Processing has made important contributions to policy and practice by enabling data-driven decisions in healthcare, environmental monitoring, and public safety. It has encouraged the development of standardized protocols for analyzing images, supported ethical guidelines for AI use, and helped improve the accuracy of digital records. Additionally, it has boosted efficiency, enhanced automation, and supported better planning across fields like medicine and urban development.

## 11. Conclusions

This paper has explored the current state of research and the key areas where digital image processing technology is being applied. Also, this paper discusses the key steps involved in Digital Image Processing. Today, this technology is widely used in everyday life, from the internet to mobile phones, showing just how closely it's connected to our daily experiences. As technology continues to evolve, digital image processing will also keep advancing. To support this growth, more research and exploration will be needed from those working in the field.

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