**IMAGE PROCESSING TECHNOLOGY IN FOOD INDUSTRY**

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**Abstract**

 Image processing technology plays a pivotal role in the food industry, providing necessary support for ensuring product quality and optimizing operational efficiency. Its numerous benefits have accompanied in a revolutionary era of technological progress, significantly altering the dynamics of food production, inspection, and consumption. This chapter offers an exhaustive overview of cutting-edge techniques and emerging trends while also acknowledging the persistent challenges associated with their integration. Image processing technology has revolutionized crucial aspects of the industry, from quality control and defect detection to grading food products, guaranteeing that consumers receive safe and aesthetically pleasing items. Regarding food safety, it acts as a cornerstone for identifying pathogens and monitoring contamination, bolstering the industry's ability to protect public health. Additionally, image processing streamlines various operational facets, including sorting, packaging, and labelling, thereby contributing to enhancing efficiency and waste reduction. It also fuels innovation in food product development by facilitating ingredient analysis, texture characterization, and color consistency. Furthermore, the technology has a profound impact on consumer engagement and marketing, as it empowers image recognition and analysis for captivating packaging, and elevates food marketing campaigns. Looking toward the future, this chapter speculates on potential directions for image processing within the food industry, underscoring the imperative of ongoing research and development to address evolving industry needs and consumer expectations. Thus, image processing stands as an integral driver of innovation and excellence within the dynamic and vital field of food production, safety, and distribution.

**Keywords**: Image processing; hyperspectral imaging; 3D imaging; pathogen detection, applications of image processing.

**I. INTRODUCTION**

 Image processing is a multidisciplinary field that involves the manipulation, analysis, and interpretation of digital images using various techniques and algorithms. It includes methods such as image enhancement, object recognition, pattern recognition, image restoration, segmentation, image compression, and image analysis. Image processing is used to extract useful information from digital images and improve the quality of the image such as sharpening, smoothing, and enhancing contrast (Kulkarni et al. 2015). It plays a crucial role in a wide range of industries, enabling the extraction of valuable information from visual data. This field is employed in a variety of applications, including medical diagnostics, surveillance, robotics, autonomous vehicles, food industry and 3D object recognition.

 This technology has revolutionized medical diagnosis and research by allowing healthcare professionals to decipher intricate details from X-rays, MRI scans, and other medical images, thereby improving disease detection, treatment planning, and patient monitoring. Additionally, image processing contributes to public safety and threat prevention by detecting anomalies, tracking movement, and recognizing faces (Gupta et al. 2020). This technique also helps in autonomous vehicles by processing visual information from cameras placed on the vehicle to detect obstacles and navigate the environment and also used in robotics for vision-guided tasks, such as object manipulation and navigation. Further, image processing is used in agricultural operations to optimize farming practices and monitor crop health, resulting in increased yields and a sustainable farming system. This process ensures product excellence in manufacturing by identifying defects and maintaining consistency (Rao, 2006)

 Image processing is integral to the food processing industry, ensuring product quality and optimizing operational efficiency through a variety of benefits. The significance of this tool lies in its ability to provide insight into each stage of production in real-time, without invasive measures. Image processing systems utilize advanced cameras and sensors to analyze food items in detail, identifying defects, inconsistencies, and contaminants which may compromise quality and safety. It minimizes the risk of substandard products reaching consumers, streamlines the quality control process, reducing manual inspections, and expedites production (Turgut et al. 2014).

Moreover, image processing has a significant impact on the optimization of food processing operations. The analysis of visual data allows the automation of tasks such as sorting, grading, and packaging. This not only expedites the production process but also reduces reliance on manual labor, leading to improved efficiency and reduced operational costs. As well, image processing algorithms can closely monitor and control various parameters during processes such as baking, frying, and cooking, ensuring consistent results and minimizing waste.

 This chapter explores comprehensively the applications and benefits of image processing within the food industry, addressing objectives such as identifying defects in products and ensuring product consistency, enhancing quality control through image processing, enhancing food safety by detecting contamination, and optimizing operational efficiency through image processing. Moreover, it provides an overview of advanced techniques and emerging trends, while acknowledging challenges and describing possible future directions for applying image processing to various aspects of food production, safety, and distribution.

**II. FUNDAMENTALS OF IMAGE PROCESSING**

 The fundamental concepts behind image processing provide the framework for manipulating and analysing digital images. At its core, an image is composed of individual picture elements known as pixels, each representing a specific point on the image grid. Pixels store color and intensity information, forming the basis for all image-related operations. Image sampling and quantization translate real-world scenes into discrete digital representations, determining resolution and detail. Convolution and filtering involve sliding predefined masks over images in order to perform operations such as blurring and edge detection. Histograms provide insights into intensity distribution, aiding in contrast enhancement and dynamic range adjustment.

 In digital image processing, the techniques of image filtering, noise reduction, and image transformation play pivotal roles in refining and extracting meaningful information from visual data. Image filtering involves applying mathematical operations to highlight or modify specific features, enabling tasks like edge detection and texture enhancement. Noise reduction techniques mitigate distortions caused by unwanted variations, preserving image clarity and essential details. Image transformation encompasses operations that modify an image's attributes, from geometric adjustments like rotation and scaling to frequency domain transformations for tasks like compression and analysis. These techniques collectively empower image processors to enhance quality, reduce distortions, and uncover hidden patterns, contributing to a spectrum of applications across scientific, artistic, and technological domains

**III. METHODS OF IMAGE PROCESSING**

 Image processing can be broadly categorized into two main methods: analog image processing and digital image processing. These methods differ in their approach, techniques, and underlying technologies.

**A. Analog Image Processing**

 Analog image processing involves manipulating images in their continuous, physical form by utilizing various optical and electronic devices for processing. Techniques like filtering, masking, and optical lens systems were commonly used to modify images. Analog image processing is mainly suited for real-time applications where immediate visual alterations are required, such as television broadcasting and video signal processing. However, it is limited by its susceptibility to noise, degradation, and the lack of precision inherent in analog systems.

**B**. **Digital Image Processing**

Digital image processing involves manipulating digital representations of images using computational algorithms. This method gained widespread use with the advent of computers and digital devices. Digital processing offers numerous advantages, including superior precision, flexibility, and the ability to apply complex transformations and analyses to images. Digital images are composed of discrete pixels, and operations are performed using numerical values, enabling more accurate control over processing outcomes (Mahmud et al. 2014).

**IV. TECHNIQUES OF DIGITAL IMAGE PROCESSING**

The various techniques of digital image processing to analyse digital images are as explained in detail as follows

* + Image acquisition
	+ Image Pre-processing
	+ Image Enhancement
	+ Image Restoration
	+ Morphological Processing
	+ Image Segmentation
	+ Object Recognition
	+ Image Data Compression

**A. Image Acquisition**

 Image acquisition is the process of capturing real-world visual information using imaging devices and sensors, which convert light or electromagnetic radiation into digital signals. These signals are then transformed into discrete pixel values through sampling and quantization, producing a digital image that represents the scene (Moustakides et al. 2000). Image acquisition encompasses technologies like digital cameras, scanners, and medical imaging equipment, forming the initial step in the digital image processing workflow and providing the foundation for subsequent analyses and applications across diverse fields

**B. Image pre-processing**

 Image pre-processing is a crucial technique that involves enhancing and preparing raw digital images for further analysis by refining their quality and characteristics. It encompasses operations like noise reduction, sharpness enhancement, and color correction, which collectively enhance image fidelity and mitigate irregularities (Gonzalez, 2009). By applying image pre-processing, unwanted artifacts and variations are minimized, allowing subsequent image processing algorithms to operate more effectively and accurately extract meaningful information from the images (Jayaraman et al. 2009).

**C. Image Enhancement**

 Image enhancement is a method used to improve the visual quality of digital images by applying various algorithms to adjust attributes such as brightness, contrast, and sharpness. This technique aims to highlight important features, enhance details, and enhance overall image appearance, making it more suitable for analysis, interpretation, and presentation (Jayaraman et al. 2009). By modifying pixel intensities and distributions, image enhancement techniques effectively optimize image clarity and visual appeal, contributing to better understanding and utilization of the image content. This technique is categorised into two types spatial domain and frequency domain methods. spatial domain methods manipulate images in their raw pixel form, while frequency domain methods analyze images by breaking them down into their frequency components (Rao, 2006).

**D. Image Restoration**

 Image restoration is a technique within digital image processing that aims to recover and improve the quality of degraded or distorted images caused by factors like blurring, noise, or atmospheric interference (Maru, 2014). This process involves applying algorithms to estimate and reverse the degradation process, enhancing image details and fidelity. By salvaging valuable information from corrupted images, image restoration contributes to better image interpretation, scientific understanding, and practical applications in various domains (Abbas, 2020).

**E. Morphological Processing**

 Morphological processing, a technique in digital image processing, involves shaping and refining images using mathematical operations inspired by set theory and topology. By applying operations like dilation (expanding object boundaries) and erosion (shrinking object boundaries), morphological processing modifies image structures and features. These operations are particularly useful for tasks such as removing noise, highlighting specific patterns, and extracting distinct objects (Vijayran, 2013).

**F. Image Segmentation**

 Image segmentation is a fundamental technique in digital image processing that involves dividing an image into distinct and meaningful regions based on certain criteria by identifying boundaries and isolating objects or areas of interest within an image (Kaur, 2014). This technique, achieved through algorithms like thresholding, edge detection, feature, model and region-based methods, enables enhanced understanding and targeted analysis of intricate visual data (Aly et al. 2011).

**G. Image Recognition**

 Image recognition involves training algorithms to identify and classify objects or patterns within images. Utilizing intricate neural network architectures like convolutional neural networks (CNNs), image recognition systems are trained to comprehend the intricate nuances and distinctive characteristics of visual data (Ramadevi et al. 2010).

**H. Image Compression**

 Image compression minimizes file sizes while preserving visual content; it employs algorithms to remove redundancy and optimize data representation. This process is crucial for efficient storage, fast transmission, and reduced bandwidth usage, enhancing multimedia applications, web content, and more. Image compression strikes a balance between compactness and quality, facilitating seamless sharing of visual information across various digital platforms (Dhawan, 2011).

**V. APPLICATIONS OF IMAGE PROCESSING IN FOOD INDUSTRY**

 Image processing has a diverse range of applications within the food industry, revolutionizing various aspects of production, quality control, and safety.

**A. Detection of Defects, Blemishes, and Contaminants in Food Products**

 In the realm of food safety and quality control, the detection of defects, blemishes, and contaminants holds a significant role, safeguarding consumer health and upholding industry standards. The application of advanced imaging has revolutionized the detection of defects and blemishes in food items. Automated systems equipped with high-resolution cameras and intricate algorithms meticulously examine a diverse range of products, from delicate fruits to packaged goods (Gupta et al. 2020). Through precise visual analysis, these systems can identify imperfections such as cuts, bruises, and variations in color.

 The rapid and accurate detection of contaminants within food products is important to ensuring consumer safety and product integrity. Advanced technologies such as X-ray scanning, infrared spectroscopy, and machine vision, contaminants like foreign particles, pathogens, or hazardous substances are meticulously identified and isolated from goods intended for consumption or use (Chandrasekaran et al. 2019). This process plays a pivotal role in safeguarding consumer health, preventing product recalls, and upholding industry standards.

**B. Grading and Sorting of Fruits, Vegetables, and Other Food Items**

 Image processing technology facilitates accurate sorting and grading of food items based on specific criteria such as size, color, and quality. This process involves the use of high-resolution cameras, sensors, and advanced algorithms to assess the external characteristics of products. By harnessing the power of these technologies, the industry ensures that only items meeting specific quality standards proceed to distribution, contributing to both efficiency and consumer satisfaction (Kumar et al. 2020).

 In fruits and vegetables, Image-based grading and sorting systems for fruits and vegetables offer rapid and accurate assessments that are critical for market competitiveness. These systems categorize produce into different grades, allowing for precise distribution based on visual attributes (Ireri et al. 2019). For instance, fruits can be sorted based on color consistency, size uniformity, and the absence of blemishes. Similarly, vegetables can be evaluated for shape, texture, and any visible flaws. This automated approach not only reduces the reliance on manual labor but also minimizes subjectivity in the grading process, resulting in a more objective and consistent outcome.

 The application of this technology extends beyond the fresh produce, branching into an array of food items ranging from grains to packaged goods. Grains, such as rice or wheat, can be meticulously sorted based on attributes like size and the presence of any visible defects, ensuring that the end product maintains uniformity and quality. Nuts, appreciated for their appearance and texture, can be subjected to sorting based on factors like size, color, and the integrity of their shells. Even packaged items undergo the scrutiny of image-based sorting. Labels, packaging integrity, and overall visual presentation are meticulously assessed to guarantee that products reach consumers in optimal condition, reinforcing trust and satisfaction (Xu et al. 2019).

**C. Texture Analysis**

 Texture analysis, a critical aspect of product evaluation, is greatly facilitated by image processing. The technology captures detailed images of products, allowing algorithms to analyze surface patterns, irregularities, and tactile properties. For instance, in the realm of baked goods, image processing can assess the crumb structure and crust texture of bread, cakes, and pastries (Saha & Manickavasagan, 2021). Similarly, in the assessment of fruits and vegetables, the technology can discern the ripeness, firmness, and tenderness of produce, aiding in determining optimal harvest times and ensuring consistent quality.

**D. Color Analysis**

 Color analysis through image processing is a sophisticated technique that involves the quantification and interpretation of color information within digital images. the context of the food industry, color analysis entails capturing high-resolution images of food items using specialized cameras or scanners. These images are then processed using advanced algorithms that extract quantitative data about the color properties of the products. These properties include parameters like hue (the dominant wavelength of color), saturation (intensity or purity of color), and brightness (lightness or darkness of color). By converting visual information into numerical values, color analysis enables precise and objective comparisons of color across different batches, products, or processing conditions (Castro et al. 2019).

**E. Automated Cooking and Quality Monitoring**

Image processing is a pivotal technology in the domain of automated cooking and quality monitoring within the food industry, utilizing advanced imaging techniques and algorithms to ensure precise cooking control and real-time quality assessment. Through high-resolution cameras and sophisticated analysis, image processing monitors cooking processes, adjusting parameters like temperature and time to achieve consistent culinary outcomes (Lee et al.2019). Additionally, it visually inspects cooked products, assessing attributes such as color, texture, and presentation, enabling prompt identification and rectification of deviations from quality standards. This integration of image processing streamlines automated cooking, enhancing product consistency, quality, and consumer satisfaction.

**F. Monitoring of Packaging and Seal Integrity**

 The application of image processing technology is also useful in the domain of packaging integrity and seal quality. This advanced approach utilizes high-resolution cameras and sophisticated algorithms to ensure that packaging materials maintain their integrity, protecting the contents from external contaminants and preserving product freshness.

 The primary functions of image processing in this context is the detection of any defects or irregularities in the packaging material itself. Cameras capture detailed images of packaging surfaces, allowing algorithms to analyze them for imperfections such as tears, punctures, or uneven seals. Any discrepancies are rapidly identified, alerting operators to take corrective actions before products are distributed. Moreover, the technology is adept at evaluating the consistency of seal quality, ensuring that closures are secure and airtight. This is particularly crucial for perishable goods or products sensitive to environmental conditions, as a compromised seal could lead to spoilage or contamination (Xu et al. 2019).

 Additionally, this technology aligns with industry regulations and safety standards, reinforcing the overall integrity of the food supply chain. In essence, image processing plays a vital role in fortifying the protective barrier between food products and their external environment, ensuring that they remain safe, fresh, and of the highest quality until they reach the consumer's table.

**G. Traceability and Authentication**

 Image processing technology has emerged as a pivotal tool in ensuring traceability and authentication across diverse industries, particularly in sectors where product origin, quality, and authenticity are of paramount importance. In the context of the food industry, image processing plays a significant role in guaranteeing the transparency, safety, and credibility of products, fostering consumer trust and regulatory compliance (Massaro et al. 2019). Through the capture and analysis of visual data, image processing facilitates meticulous traceability by creating a digital roadmap of a product's journey from its inception to market, ensuring rapid identification of affected items and optimizing supply chain efficiency. Furthermore, image processing enhances product authentication by utilizing distinct visual attributes as digital fingerprints for verifying authenticity, protecting against counterfeit products and fortifying the bond between producers and consumers (Liu et al. 2022).

**VI. ADVANCED IMAGE TECHNIQUES IN FOOD INDUSTRY**

Advanced image processing techniques encompass a wide array of sophisticated methodologies that have significantly elevated the capabilities of analyzing and manipulating images. In recent years, Machine Learning and Deep Learning techniques have revolutionized image processing by enabling computers to learn and recognize patterns from large datasets. Convolutional Neural Networks (CNNs), a type of deep learning architecture, have demonstrated exceptional capabilities in tasks like image classification, object detection, and image generation.

**A. Machine Learning and Deep Learning**

 Machine learning and deep learning have revolutionized image processing techniques, particularly in industries like the food sector. These advanced methodologies harness the power of data-driven models to autonomy ously learn and extract meaningful patterns from images, bringing a new level of precision and efficiency to various applications. In the realm of food processing, machine learning algorithms, and deep learning architectures, such as convolutional neural networks (CNNs), have significantly enhanced quality control and inspection processes. They enable the automatic detection of defects, blemishes, and contaminants in food products by analyzing visual data, ensuring that only items meeting stringent standards reach consumers (Naranjo-Torres et al.2020). Moreover, these techniques excel at sorting and grading tasks, accurately categorizing products based on attributes like color, size, and shape. This not only ensures consistent product quality but also optimizes supply chain efficiency by facilitating precise categorization and distribution.

Furthermore, machine learning and deep learning play a pivotal role in enhancing food safety. By processing images, these techniques can identify potential hazards such as foreign objects or signs of contamination, bolstering the overall integrity of food products. Predictive analysis is another notable application, as machine learning models can forecast food quality attributes like freshness and shelf life based on visual cues (Simonyan & Zisserman, 2014). This aids in efficient inventory management and waste reduction. Additionally, these techniques support new product development by analyzing consumer preferences and visual attributes, guiding the creation of innovative and appealing food offerings (Habib et al. 2021). The machine learning and deep learning image processing techniques have ushered in a new era of accuracy and efficiency in the food industry, spanning quality control, safety, predictive analysis, and product innovation.

**B. Hyperspectral Imaging**

 Hyperspectral imaging has emerged as a game-changing technology within the food industry, offering a sophisticated and detailed perspective on food products that traditional imaging methods cannot match. By capturing a multitude of narrow and contiguous spectral bands for each pixel in an image, hyperspectral imaging enables precise analysis of composition, quality, and safety attributes in various food items (Park et al. 2011). This capability has led to a range of applications, from detecting contaminants and foreign materials to assessing freshness and authenticity of products.

 In quality control and sorting, hyperspectral imaging excels by detecting subtle variations in color, texture, and ripeness that human eyes or conventional cameras might overlook. This capability enables automated grading and sorting of fruits, vegetables, and other produce, ensuring that only products meeting specific quality criteria are delivered to consumers. Furthermore, hyperspectral imaging enhances food safety protocols by swiftly identifying contaminants, foreign objects, or pathogens through their distinct spectral signatures. This swift and accurate detection contributes to minimizing health risks and upholding rigorous safety standards in the food industry.

 The versatility of hyperspectral imaging extends to authenticity verification and traceability. By establishing unique spectral profiles for different products, this technology aids in confirming the origin and authenticity of high-val ue food items like spices, oils, and beverages. These spectral fingerprints create a transparent trail that can be tracked throughout the supply chain, bolstering traceability efforts (Ma et al. 2019). Additionally, hyperspectral imaging contributes to research and innovation by providing valuable insights into food processing techniques, shelf-life assessment, and packaging integrity. In essence, hyperspectral imaging stands as a transformative tool that empowers the food industry to achieve higher levels of quality, safety, and transparency, ultimately enhancing consumer confidence and driving industry advancement.

C. **3D Imaging and Tomography**

 The integration of 3D imaging and tomography has introduced a transformative dimension to analysis, visualization, and quality assessment within the food industry. In quality control and inspection, 3D imaging offers a meticulous assessment of food products, enabling the measurement of volume, shape, and surface characteristics. This is particularly crucial in industries like confectionery and bakery, where uniformity is paramount. Complementing this, tomography employs complex mathematical algorithms to reconstruct cross-sectional images, revealing internal structures and potential defects that might not be visible externally. This depth of insight enhances the identification of irregularities and inconsistencies (Wang et al. 2018). 3D imaging and tomography excel in detecting foreign objects, contaminants, or irregularities within food items.

 The role of 3D imaging and tomography extends beyond mere examination, influencing packaging and design strategies. By capturing a comprehensive view of packaged products, manufacturers can evaluate the effectiveness of packaging materials, and assess potential areas of vulnerability (Vidhya et al. 2017). This assists in optimizing packaging designs for better product preservation and protection, ultimately minimizing waste and ensuring product integrity.

**VII. CHALLENGES OF IMAGE PROCESSING IN FOOD INDUSTRY**

 Implementing image processing in the food industry brings about transformative possibilities, but it also entails a set of common challenges that require careful consideration for successful integration. The inherent variability of food products, spanning shapes, colors, textures, and sizes, poses a significant challenge. Designing algorithms that can accurately analyze this diverse range of attributes requires extensive data collection, comprehensive model training, and the exploration of advanced machine learning techniques. Adaptable and robust image processing methods are necessary to ensure consistent and accurate results across the entire spectrum of food items.

 The dynamic and often unpredictable conditions of food processing environments further complicate implementation. Fluctuating lighting, shadows, reflections, and varying backgrounds can impact image quality and distort captured data. Employing techniques like normalization, background subtraction, and color correction becomes crucial to mitigate these distortions and ensure reliable analysis. Moreover, real-time adjustments and calibration mechanisms are essential to account for changing conditions on production lines, enabling accurate and consistent image processing outcomes.

 Integrating image processing seamlessly into existing production workflows is a complex endeavor. Adapting capture and analysis systems to work harmoniously within established processes requires meticulous planning and coordination. Avoiding disruptions to production speed, maintaining product quality, and ensuring that the technology doesn't introduce bottlenecks demand rigorous testing and optimization. Training personnel to operate and interpret the image processing systems efficiently adds another layer of complexity, bridging the gap between sophisticated technology and practical implementation within the production environment.

**VIII. FUTURE ASPECTS**

 The field of image processing is poised to undergo significant transformations and advancements in the coming years, driven by emerging technologies, evolving applications, and innovative approaches. One prominent aspect is the seamless integration of image processing with artificial intelligence (AI) and machine learning (ML) techniques. The fusion of AI and image processing will not only elevate the accuracy and efficiency of visual data interpretation but also unlock new realms of innovation by enabling machines to perceive and comprehend the visual world in ways previously unattainable.

 Another significant future aspect revolves around the convergence of image processing with immersive technologies like augmented reality (AR) and virtual reality (VR). The synthesis of these technologies will enable the creation of immersive visual experiences that blend digital and physical worlds seamlessly. Moreover, this fusion will empower industries to leverage rich visual data for data visualization, spatial mapping, and real-time decision-making.

**IX. CONCLUSION**

 In conclusion, the integration of image processing in the food industry has ushered in a new era of technological advancement and innovation, fundamentally transforming the way food is produced, inspected, and consumed. Through its diverse applications, image processing has proven to be an invaluable tool for ensuring product quality, safety, and efficiency. It has enabled accurate detection of defects, contaminants, and blemishes in food products, facilitating stringent quality control measures. Moreover, image processing has revolutionized processes such as grading, sorting, and packaging inspection, streamlining operations and minimizing human error. The utilization of advanced techniques like hyperspectral imaging and deep learning has extended the capabilities of image processing, providing insights into texture, color, and composition that were previously inaccessible. However, challenges such as variability in product appearance and the need for seamless integration into existing workflows remain, highlighting the ongoing need for interdisciplinary collaboration and technological refinement. As the food industry continues to evolve, image processing stands as a cornerstone of its progress, playing a pivotal role in enhancing consumer trust, optimizing resource utilization, and driving the development of innovative and sustainable food products for a rapidly changing world.

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