

# The Origins of Digital Image Processing & Application areas in Digital Image Processing Medical Images

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**Abstract**—One of the first application of digital image was in the newspaper industry, when picture were first sent by submarine cable between London and New York. Introduction of the Bartlane cable picture transmission system in the early 1920s reduced the time required to transport a picture across the Atlantic from more than a week to less than three hours. Specialized printing equipment coded pictures for cable transmission and then reconstructed them at the receiving end. Some of the initial problems in improving the visual quality of these early digital pictures were related to the selection of printing producers and the distribution of intensity levels. In fact, digital images require so much storage and computation power that progress in the field of digital image processing has been dependent on the development of digital computers and of supporting technologies that include data storage, display and transmission. The digital image is composed of a finite number of elements, each of which has a location and values. These elements are referred to as picture element, image element & pixels. Pixels are used to denote the element of a digital image. The process of acquiring an image of the area containing the text, preprocessing that image, extracting the individual characters, describing the character in the form suitable for computer processing & recognizing those individual characters are Digital Image Processing.

**Keyword:** MRI – Magnetic Resonance Imaging, IPT – Electron Microscopy, CT – Computerized Tomography, CAT – Computerized Axial Tomography, PC – Personal Computer

## INTRODUCTION:

### 1. What is Digital Image Processing?

Digital image processing allows one to enhance image features of interest while attenuating details irrelevant to a given application, and then extract useful information about the scene from the enhanced image. An image may be defined as a two-dimensional function,  $f(x, y)$ , where  $x$  and  $y$  are spatial (plane) co-ordinates, and amplitude of  $f$  at any pair of

co-ordinates  $(x, y)$  is called the *intensity or gray level* of the image at that point. When  $x$ ,  $y$  and the amplitude values of  $f$  are all finite, discrete quantities, we call the image a *digital image*. The field of *digital image processing* refers to processing digital images by means of a digital computer. A digital image is composed of a finite number of elements, each of which has a particular location and values. These elements are referred to as picture elements, image elements and pixels. Pixels are the term most widely used to denote the element of digital image.

### 2. The Origins of Digital Image Processing:

One of the first application of digital image was in the newspaper industry, when picture were first sent by submarine cable between London and New York. Introduction of the Bart lane cable picture transmission system in the early 1920s reduced the time required to transport a picture across the Atlantic from more than a week to less than three hours. Specialized printing equipment coded pictures for cable transmission and then reconstructed them at the receiving end. Some of the initial problems in improving the visual quality of these early digital pictures were related to the selection of printing producers and the distribution of intensity levels. In fact, digital images require so much storage and computation power that progress in the field of digital image processing has been dependent on the development of digital computers and of supporting technologies that include data storage, display and transmission. The digital image is composed of a finite number of elements, each of which has a location and values. These elements are referred to as picture element, image element & pixels. Pixels are used to denote the element of a digital image. The process of acquiring an image of the area containing the text, preprocessing that image, extracting the individual characters, describing the character in the form suitable for computer processing & recognizing those individual characters are Digital Image Processing. Digital image processing techniques began in the late 1960s and early 1970s to be used in medical imaging, remote Earth resource observations and astronomy. The invention in the early 1970s of computerized axial tomography (CAT) also called computerized tomography (CT) is one of the most

important events of image processing in medical diagnosis. Computerized axial tomography is a process in which a ring of detectors encircles a patient and an X-Ray source, concentric with detector ring, rotates about the patient. The X-Ray passes through the object and is collected at the opposite end by the corresponding detectors in the ring. As the source rotates, this procedure is repeated. Tomography consist algorithms that use the sensed data to construct an image that represent the 'slice' through the object. Computer procedure are used to enhance the contrast or code the intensity levels into color for easier interpretation of X-Rays and other images used in industry, medicine and the biological sciences. Image enhancement and restoration procedure are used to process degraded images of unrecoverable objects or experimental result too expansive to duplicate. Image processing methods have successfully restored blurred pictures that were the only available records of rare artifacts lost or damaged after being photographed.

### 3.Application areas in Digital Image Processing

Today, there is almost no area of technical endeavor which is not impacted by digital image processing. Many application oriented image analyzers are available and are working satisfactorily in real environment. One of the simplest way to develop a basic understanding of image processing application to categorize images according to there resources e.g. Visual, X-ray and so on. The principal energy source for images in use today is the electromagnetic energy spectrum. Other important sources of energy include ultrasonic and electronic in the form of electron beams used in electron microscopy. In this section we discuss how images are generated in these various categories and the area in which they applied. Images based on radiation from the EM spectrum are most familiar, especially images in the X-ray and visual bands of the spectrum.

### 4. X-ray Imaging

X-ray imaging is perhaps the most familiar type of imaging. X-ray is among the oldest sources of EM radiation used for imaging. The use of X-rays is medical diagnostics, but they also are used extensively in industry and other areas like astronomy. X-ray for medical imaging are generated using X-ray tube, which is a vacuum tube with a cathode and anode. The cathode is heated, causing free electrons to be released. These electrons flow at high speed to the positively charged anode. When the electrons strike a nucleus, energy is released in the form of X-ray radiation. Figure (a) shows a familiar chest -ray generated simply by placing the patient between an X-ray source and a film sensitive to X-ray energy.



In digital radiography, digital images are obtained by one of two

methods: (1) by digitizing X-ray film; or (2) by having the x-ray that passes through the patients fall directly onto devices that Fig (a) X-ray

Convert x-ray to light. The light signal in turn is captured by a light sensitive digitizing system.

Angiography is another major application in an area called contrast enhancement radiography. This is used to obtain image of blood vessels. A catheter is an inserted into a vein in the groin. The catheter is threaded into blood vessel and guided to area to be studied. When the catheter reaches the site under investigation, an X-ray contrast medium is injected through the catheter. This enhances contrast of the blood vessels and enables the radiologist to see any blockages.

### 5. Steps in Digital Image Processing

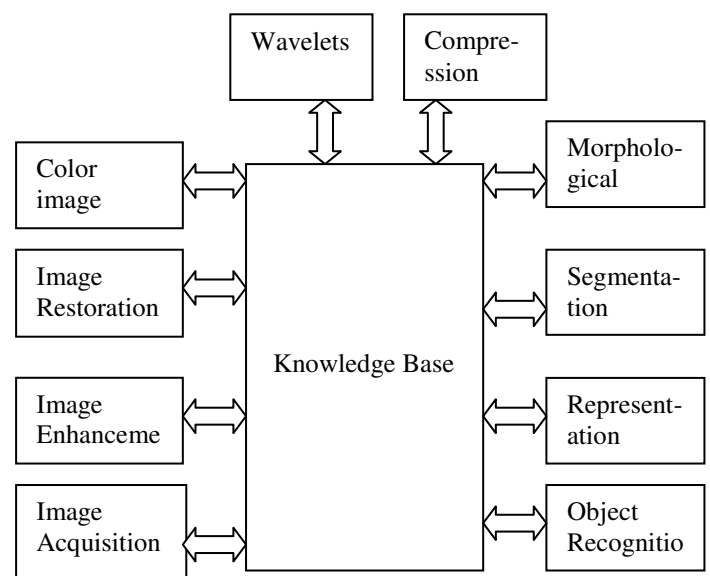


Fig. (b) Steps in digital image processing

The organization is summarized in fig (b). The diagram does not imply that every process is applied to an image. Rather the intention is to convey an idea of all methodologies that can be applied to images for different purpose and possibly with different objective.

*Image acquisition* is the first process introduces a number of basic digital image concept that are used. The image acquisition stage involves preprocessing such as scaling.

*Image enhancement* is the simplest and most appealing areas of digital image processing. Enhancement simply to highlight certain features of interest in an image. It is a very subjective area of image processing. Image restoration is an area that also deals with improving the appearance of an image. Image restoration is objective, the sense that restoration techniques tend to be based on mathematical or probabilistic model of

image degradation. *Color image processing* is an area that has been gaining in importance because of the significant increase in the use of digital images over the Internet. Color is used for extracting the features of interest in an image.

*Wavelets* are the foundation for representing images in various degree of resolution, in which images are subdivided successfully into smaller regions. *Compression* deals with techniques for reducing the storage required to save image or bandwidth required to transmit. *Morphological processing* deals with tools for extracting image component that is useful in the representation and description of shape. *Segmentation* producers partition an image into its constituent part or objects. Segmentation is one of the most difficult tasks in digital image processing. Segmentation procedure brings the process a long way toward successful solution of imaging problems that require object to be identified individually. *Representation* always follows the output of segmentation stage, which is raw pixel data, constituting either boundary of region or all the points in the region itself.

Choosing a representation is only part of the solution for transforming raw data into a suitable subsequent computer processing. Recognition is the process that assigns a label to an object based on its descriptors.

## 6. Component of an Image Processing System.

### Network

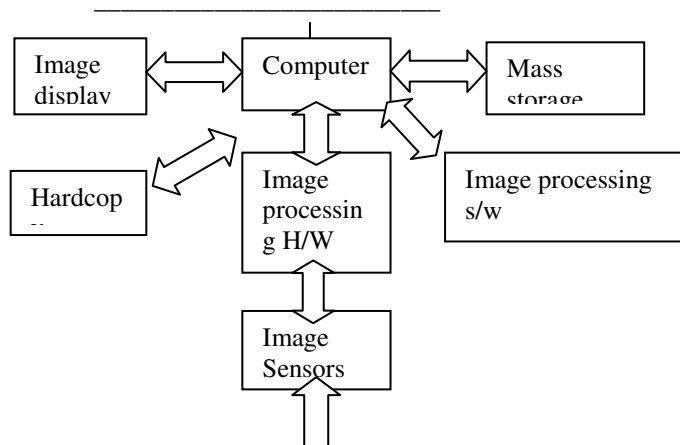
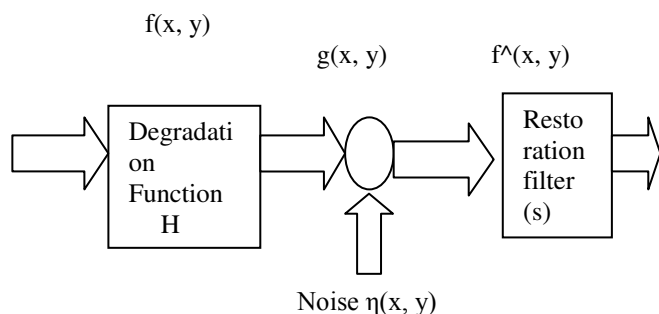


Fig. ©Components of a general purpose image processing system

In the 1980 and early in 1990s, the market shifted to image processing hardware in the form of single boards designed to be compatible with industry standard buses and to fit into engineering workstations cabinets and personal computers. In addition to lowering costs, this market shift also served as a catalyst for a significant number of new companies whose specialty is the development of software written specifically for image processing. Although large scale image processing

system still being sold for massive imaging application, such as processing of satellite images, the trend continues toward miniaturizing and blending of general purpose small computers with specialized image processing system. Fig © shows a typical general purpose system used for digital image processing. With reference to *sensing*, two elements are required to acquire digital images. The first physical device that is sensitive to the energy radiated by the object we wish to image. The second is the *digitizer* is a device for converting the output of physical sensing device into digital form. *Specialized image processing* hardware usually consist of the digitizer just mentioned, plus hardware that performs other primitive operations, such as an arithmetic logical unit, which performs arithmetic and logical operations in parallel on entire images. The *computer* in an image processing system is a general-purpose computer and can range from a PC to a supercomputer. In these systems, almost any well equipped PC type machine is suitable for offline image processing tasks. *Software* for image processing consists of specialized modules that perform specific tasks. A well designed package also includes the capability for the user to write code that, as a minimum, utilizes the specialized modules. *Mass storage capability* is a must in image processing applications. An image of size 1024 \* 1024 pixels, in which the intensity of each pixel is an 8 bit quantity, required one megabytes of storage space if the image is not compressed. Digital storage for image processing application falls into three principal categories: (1) short term storage for use during processing, (2) on-line storage for relatively fast recall and, (3) archival storage, characterized by infrequent access. Storage is measured in bytes, Kbytes, Mbytes, Gbytes, and Tbytes. *Image display* in use today is mainly color TV monitors. Monitor is driven by the output of images and graphics display cards that are an integral part of computer system. Image display application that cannot be met by display cards available commercially as part of computer system *Hardcopy* devices for recording images include laser printer, film camera, heat-sensitive device, inkjet units and digital units. *Networking* is almost a default function in many computer systems today. Because of large amount of data inherent in image processing application, the key consideration in image transmission is bandwidth. In dedicated networks, this typically is not problem, but communication with remote sites via internet is not always as efficient. [Gonzalez, 2005]

## 7. A Model of the Image Restoration Process



**Fig. model of image restoration**

A Fig shows, the degradation or restoration process is modeled. A degradation function that, together with an additive noise term, operates on an input image  $f(x, y)$  to produce a degraded image  $g(x, y)$ . Given  $g(x, y)$ , some knowledge about the degradation function  $H$ , and some knowledge about the additive noise term  $\eta(x, y)$ , the objective of restoration is to obtain an estimate  $f^{\wedge}(x, y)$  of the original image. We are estimating as close as possible to the original input image and, in general, the more we know about  $H$  and  $\eta$ , the closer  $f^{\wedge}(x, y)$  will be to  $f(x, y)$ . If  $H$  is a linear, positive-invariant process, then the degraded image is given in the spatial domain by

$$g(x, y) = h(x, y) * f(x, y) + \eta(x, y)$$

Where  $h(x, y)$  is the spatial representation of the degradation function and symbol '\*' indicates spatial convolution. That convolution in the spatial domain is equal to Multiplication in the frequency domain, so we may write the model in an equivalent frequency domain representation

$$G(u, v) = H(u, v) F(u, v) + N(u, v)$$

Where the term in capital letters is the Fourier transforms of the corresponding terms.

We assume that  $H$  is the identity operator and we deal with degradation due to noise.

### Why Medical Imaging?

The internal structure of the human body is not generally visible to the human eyes. However, by various imaging techniques images can be created through which the medical professionals can look into the body to diagnose abnormal conditions and guide the therapeutic procedures. Different medical imaging methods reveal different characteristics of the human body. With each method, image quality and structure visibility can be considerable, depending on characteristics of the imaging equipment, skill of the operator, and compromises with factors such as patient radiation exposure and imaging time.

**Noise in Medical Image:** It is generally desirable that image brightness is to be uniform except where it changes to form an image. There is a variation in the brightness of a displayed image even when no image detail is present. This variation is usually random and has no particular pattern reducing the image quality specifically when the images are small and have relatively low contrast. This random variation in image brightness is nothing but a noise. All medical images contain some visual noise. The presence of noise gives an image a grainy, textured, or snowy appearance. No imaging method is free of noise, but noise is much more prevalent in certain types of imaging procedures than in others. Most of the imaging methods can create image features that do not represent a body structure or object. These are image artifacts.

In many situations an artifact does not significantly affect object visibility and diagnostic accuracy. But artifacts can obscure a part of an image or may be interpreted as an anatomical feature. A medical image should give an accurate impression of body objects in terms of their size, shape, and relative positions; however, it may introduce distortion of these factors

## MEHODOLOGY:

### 1. Noise Models :

Real images are often degraded by some random errors – This degradation is usually called noise. Noise can occur during image capture transmission or processing and may dependent on or independent of image content. Basically, there are two types of noise models: Noise in the spatial domain (described by the noise probability density function) and noise in the frequency domain described by various Fourier properties of the noise. Now here with we are discussing about the noise is independent of image co-ordinates.

### 2. Gaussian Noise:

Gaussian noise is popular noise approximation. A random variable with Gaussian (normal) distribution has its probability density is given by the Gaussian curve.

The ID case the density function is

$$P(x) = 1/\sqrt{2\pi}\sigma e^{-(x-\mu)^2/2\sigma^2}$$

Where  $\mu$  is the mean and  $\sigma$  is the standard deviation of random variable. Gaussian noise that occurs in many practical cases.

### 3. Additive Noise:

When an image is transmitted through some communication channel, a noise which is usually independent of the signal occurs. Similar noise arises in video camera. This signal



independent degradation is called additive noise and can be described by the following models.

$$F(x, y) = g(x, y) + v(x, y)$$

Where  $v(x, y)$  is noise, which independent of input image  $g(x, y)$  and  $f(x, y)$  is corrupted image. The noise magnitude depends on the signal magnitude itself.

#### Multiplicative Noise:

If the noise magnitude is much higher in comparison with the signal we can write

$$F = g + gv = g(1+v) = gv$$

The above equation describes the multiplicative noises is the television raster degradation which depends on TV lines in ten area of line this noise is minimal. [Gonzalez, 2005]

#### Quantization Noise:

It occurs when insufficient Quantization levels are used for ex 50 levels for monochrome image in this case false contour appear. [Gonzalez, 2005]

#### Impulsive Noise:

Impulsive noise means that an image is corrupted with individual noisy pixels whose brightness significantly differs from the neighborhood.

#### Salt and Pepper Noise:

It is another type of noise is used to describe saturated impulsive noise an image corrupted with white and/ or a black pixel is an example. Salt and Pepper noise can corrupt binary image.

All the above noises can be added using the IPT function `imnoise()`. The basic syntax is

$$G = \text{imnoise}(f, \text{type}, \text{parameters})$$

Where  $f$  is the input image, type is the type of noise.

Function `imnoise()` converts the input image to class double in the range [0, 1] before adding noise to it. This must be taken account when specifying noise parameter.

- $G = \text{imnoise}(f, \text{Gaussian}, m, \text{var})$  adds Gaussian noise of mean  $m$  and variance  $\text{var}$  to image  $f$ . The default is zero mean noise with 0.01 variance.
- $G = \text{imnoise}(f, \text{localvar}, v)$  adds zero mean, Gaussian noise of local variance,  $V$ , to image  $f$ , where  $v$  is an array of the same size as  $f$  containing the desired variance values at each point.
- $G = \text{imnoise}(f, \text{'localvar'}, \text{image intensity}, \text{var})$  adds zero mean, Gaussian noise to image  $f$ , where the local variance of the noise,  $\text{var}$ , is a function of the image intensity values in ' $f$ '. The image intensity and  $\text{var}$  arguments are vectors must contain normalized intensity values in the rang [0,1]
- $G = \text{imnoise}(f, \text{'salt \& pepper'}, d)$  corrupts image  $f$  with salt & pepper noise, where  $d$  is the noise density (i.e. the present of the image area containing noise values). Thus, approximately  $d \times \text{numel}(f)$  pixels are affected. The default is 0.05 noise density.

- $G = \text{imnoise}(f, \text{'speckle'}, \text{var})$  adds multiplicative noise to image  $f$  using the equation  $g = f + n \cdot f$ , where  $n$  is uniformly distributed random noise with mean 0 and variance  $\text{var}$ . The default vale of  $\text{var}$  0.04.  $G = \text{imnoise}(f, \text{'Poisson'})$  generates Poisson noise from data instead of adding artificial noise to the data. In order to comply with Poisson statistics, the intensities

#### Periodic Noise:

Periodic noise is an image arises typically from electrical and / or electromechanical inferences during the image acquisition. This is the only type of spatially dependent noise. Periodic noise typically handled in an image by filtering in the frequency domain.

The model of periodic noise is a 2-D sinusoid with equation

$$R(x, y) = A \sin [2\pi [u_0(x+Bx)/ 2\pi [v_0(y+by)/N]$$

Where  $A$  is the amplitude,  $u_0$  and  $v_0$  determines the sinusoidal frequencies with respect to the  $x$  and  $y$  axes respectively and  $Bx$  and  $By$  are phase's displacements with respect to the origin.

#### DISCUSSION:

Different medical images like MRI, Cancer, x-ray and brain images have been studied. After finding the Gaussian noise in MRI image the various filtering techniques like Median filter, Adaptive filter and Average filter have been applied.

It is found that the Adaptive filter works better for the gaussian noise .Similarly after finding the Gaussian noise in cancer image the various filtering techniques have been applied and it is found that the adaptive filter works better for the noisy image. After finding the Gaussian noise in X-ray image various filtering techniques have been applied and it is found that the adaptive filter works better for the X-ray noisy image. After finding the Gaussian noise in brain image various filtering techniques have been applied and it is found that the adaptive filter works better for the noisy image. Similarly after finding the salt and pepper noise in MRI image various filtering techniques have been applied and it is found that the adaptive filter works better for the noisy image. After finding the salt and pepper noise in Cancer image various filtering techniques have been applied and it is found that the median filter works better for the noisy image. After finding the salt and pepper noise in X-ray image various filtering techniques have been applied and it is found that the median filter works better for the noisy image. After finding the salt and pepper noise in Brain image various filtering techniques have been applied and it is found that the adaptive filter works better for the noisy image.

So we can say that there are no filtering techniques which will work better for all types of noises like gaussian noise, salt and pepper noise, speckle noise and Poisson noise.

#### CONCLUSION:

In this work we have taken different medical images like MRI, Cancer, X-ray and Brain for detecting noises. We have

detected various noises like Gaussian, Salt & Pepper, Speckle, and Poisson and also removed these noises from the above medical images by applying the various filtering techniques like Median Filtering, Adaptive Filtering and Average Filtering. Through this work we have observed that the choice of filters for de-noising the medical images depends on the type of noise and type of filtering technique, which are used. It is remarkable that this saves the processing time. This experimental analysis will improve the accuracy of MRI, Cancer, X-ray and Brain images for easy diagnosis. The results, which we have achieved, are more useful and they prove to be helpful for general medical practitioners to analyze the symptoms of the patients with ease.

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