

Early Detection of Osteoarthritis Stage by Applying the Classification Technique to Human Joint Imagery

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Technique to Human Joint Imagery

Thesis Title

Abstract

Osteoarthritis (OA) is a degenerative joint disease or degenerative arthritis which is the most common chronic condition of joint inflammation coursing various paints such as joint paint, stiffness, swelling, creaking or creating sound, decrease ability to move and bone spur. It is a major cause of disability in older people. The risk of OA increases from age 45 and older. Early diagnosis is typically made using X-ray imagery.

An automated mechanism for OA stage screening is proposed in this thesis work. The fundamental idea is to segment X-ray image so as to obtain the X-ray pixels describing the area of interest (ROI) and representing these segmentations using an appropriate representation mechanism to translate an X-ray image into a form that serves to captures key information while remaining compatible with the classification process. By pairing each representation with its OA stage, a classifier can be generated to predict the OA stage accounding to the nature of a selected representation. The generated classifier can be then used to provide a quick and easy machanism for labeling the unseen X-ray imagery.

Keywords:

Image classification, Image Analysis, Image Processing, Osteoarthritis, X-ray image Analysis, Data Mining, Osteoarthritis Analysis, Osteoarthritis Stage Analysis, Medical Image Processing,

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Chapter 1

Introduction and background

1.1 Introduction

"Knowledge Discovery in Data (KDD) is the non-trivial process of identifying valid, novel, potentially useful, and ultimately understandable patterns in data"[5]. With respect to the work presented in this thesis data mining is an essential element which concerned with the discovery of the desired hidden information within the data. Data mining can be performed on any kind of data repository that includes a relational database [6]. A data warehouse is an enormous subject-oriented database integrated from multiple sources in a given time period for decision-making process. A transactional database involves day-to-day operation data for example from banking or supermarket transactions. Advanced database systems are provided to meet the requirement of new database applications, for example object-oriented and object-relational databases, and specific application-oriented database such as spatial data, time-series data, text data, multimedia data, and World Wide Web. Consequently, data mining includes sub-fields such as image mining, graph mining, and text mining. The work described in this thesis is concerned with image mining.

Image mining is a mechanism for extraction of useful knowledge and correlations from within image sets. Large amounts of visual information, in the form of digital images, are generated on a daily basis with respect to many domains such as the remote sensing and medical domains. Extracting useful knowledge from within these images presents a significant challenge. Image mining also encompasses elements from fields such as image analysis and content-based image retrieval.

According to the work presented in this thesis is directed at image classification which is a non-trivial problem, because of the typically complex structure of image data, and is still a very active field of research. In image classification, a collection of prelabelled images are taken as input and used to generate (train) a classifier which can then be applied to unseen images. Image classification typically involves the preprocessing of collections of images into a format whereby established classification techniques could be applied. As with many data mining applications the main challenge in the preprocessing of image data is to produce a representation whereby no relevant information is lost while at the same time ensuring that the end result is accurate enough to allow for the application of effective data mining.

The rest of this introductory chapter is organised as follows. The motivation for the work is presented in Section 1.2. Some previous work is presented in Section 1.3. The research questions are identified in Section 1.4. The research objectives and associated research issues are presented in Section 1.5. The expected outcome is discussed in Section 1.6. The research

methodology used to address the research issues is presented in Section 1.7. The programme of work and a summary of this chapter are presented Section 1.8 and 1.9, respectively.

1.2 Motivation

Arthritis is a chronic condition that affects joint which is the major cause of disability. Symptoms generally coursing various paints such as joint pain, stiffness, redness, warmth, swelling and decreased range of motion of the affected joints [7]. There are more than a hundred different types of arthritis and related conditions [8]. Arthritis could be happened to people at all ages and sexes, but more frequently occurs in elder people. In general arthritis is categorised into four groups: (i) Inflammatory Arthritis, (ii) Infectious Arthritis, (iii) Metabolic Arthritis and (iv) Degenerative Arthritis.

Degenerative arthritis is the most common form of arthritis, usually affecting the hands, feet, spine, and large weight-bearing joints such as the hips and knees. Also known as osteoarthritis and degenerative joint disease.

Osteoarthritis (OA) is a joint disease that mostly happens when the top layer of cartilage breaks down and wears away. The cartilage is the slippery tissue that covers the ends of bones in a joint. Healthy cartilage allows bones to glide over one another so that it can be helped to absorb the energy from the shock of physical movement. In contrast, the poor condition of cartilage between bones, the joints and the bone rub together which make pain, loss of motion, swelling and joint stiffness. OA could be affected to any joints such as knees, hips, neck, fingers, thumbs and lower back. Osteoarthritis of the knee is most common form of OA. The knee OA patient make difficulty on doing many everyday activities such as walking or climbing stairs. The example of knee X-ray image is presented in Figure 1.



Figure 1 An example of knee X-ray image

There are some statistics about the OA reported in 2013 include: in [9] it was reported that 8.75 millions people in the UK and 30 millions people in the USA have sought treatment for OA. The worldwide figure was reported in [1], it was estimated that more than 100 million people are affected by OA. Osteoarthritis condition is typically diagnosed by a doctor or clinician. Diagnosis can be made with reasonable by clinical examination based on: (i) the symptoms, (ii) the personal and family medical record, (iii) the physical signs such as joint tenderness, creaking or grating (crepitus) sounds, bony swelling, excess fluid, reduced movement, joint instability and muscle thinning and (iv) X-ray or Magnetic resonance imaging (MRI) may help confirm the condition [10].

With respect to the work in this thesis an alternative approach to detect the OA stage is proposed using a classifier applied to human bone imagery. To act as a focus for the work, the X-ray image of knee is considered. The fundamental idea is to generate a classifier using labelled X-ray image which can then be applied to detect stage the OA condition. The proposed approach offers a number of advantages: (i) speed of processing, (ii) automate processing and (iii) use widely. The disadvantage is mentioned about the accuracy in term of diagnosis from the doctor or clinician due to the OA condition complexity.

1.3 Related Work Overview

Digital image has been increasing use for medical diagnosis so that digital image becomes more important in the health care. There are a number of image formation used the example including: (i) X-ray, (ii) Computerised Tomography (CT), (iii) Magnetic Resonance Imaging (MRI), (iv) Ultrasound, (v) Positron Emission Tomography (PET) and (vi) Single Photon Emission Computed Tomography (SPECT) [11]. Medical imagery has been used for medical condition diagnosis in a number of applications. Some examples in the brain conditions were analysed using MRI brain scan such as epilepsy [12] and detection and classification of the brain tumour into malignant or benign [13].

The prevalence of knee OA, computer-based applications for OA using medical image have not been widely available. In [14] reported on the knee OA was segmented using shape analysis while MRI image was again used in [15] for segmentation for knee bone OA using graph-cut strategy. The MRI images are used to detect the knee OA. Although MRI offers an advantage on the OA joint evaluation, however X-ray images were used to detect OA condition has been reported in [3,4,5,6,7,8,18,19]. So, the work in X-ray medical image in OA detection is presented in this work.

Image analysis is an important and fundamental concept in many domains such as computer vision and pattern recognition where the main goal is to understand the characteristics of an image and interpret its semantic meaning. Image classification is an emerging image analysis technique which can be used to classify image sets according to a collection of predefined data. The performance of classifiers depends on how the best to translate raw dataset into a form

that can capture the embedded information in the image. The idea is to transform the input image into a set of features in order to produce the feature vector space representation. The image features can be categorised into two different groups: (i) general feature and (ii) domain-specific feature [16]. General feature is an application independence feature including: (i) colour, (ii) texture and (iii) shape. Domain-specific feature is an application dependent such as elements of the human face used for face recognition.

Colour features are typically used in the context of colour image representation methods, for example using colour histogram. Colour histograms offer advantages of: (i) rotation and translation invariance, (ii) computational simplicity and (iii) low storage requirements. Texture features give information about the spatial arrangement of colours and/or intensities in a given image. There are many texture extraction methods that may be adopted such as the use of Grey Level Co-occurrence Metrics (GLCMs), texture spectrums and wavelet transforms. Shape features are commonly used to describe the "geometry" of the relative position of the elements containing within a given image. The shape of an image could be implemented a graph representation [17].

1.4 Research Question

As a sequence of the motivation given in the previous section, this thesis is focused on the classification technique of OA stage analysis in term of human joint knee imagery. So, the thesis question is identified:

How the stage of knee OA can be classified by using the image mining technique from human joint x-ray images?

This research question encompass a number of issues as follows:

- 1. How to obtain the Area Of Interest (ROI) from X-ray images?
- 2. What is the information that should be extracted from the identified sub-images and how can this information best be extracted?
- 3. Once the desired information has been extracted what is the best way of representing these images so as to support the effective generation and usage of classifiers?
- 4. What are the most appropriate classification techniques for stage of OA detection from given image in the context of different information?

1.5 Research Objectives

Given the above the broad goal of the proposed research is to determine how to obtain the stage of osteoarthritis by applying image classification techniques to human joint imagery. This objective can be subdivided into a number of contributing sub-objectives:

- 1. To investigate, demonstrate and evaluate the classification techniques for predicting the stage of osteoarthritis.
- 2. To explore how to obtain area of interest from X-ray images.
- 3. To investigate and demonstrate how best to represent image data with respect to the desired application.
- 4. To determine a general framework to support the proposed stage of osteoarthritis detection process.

1.6 Expected Outcome

With the respect to work in this thesis, the expected outcome is a general framework of the detection of osteoarthritis stage process. The benefit of the proposed mechanism can be listed as follows:

- 1. To help the MD (Medical Doctor) for screening the OA stage from X-ray image.
- 2. To postpone or prevent the OA patients.
- 3. To improve the aging society healthcare system.

Chapter 2

Background and Literature Review

2.1 Introduction

In this chapter the background and some previous works are described. The content of this chapter is organized as follows: Section 2.2 Medical Image, a brief history of medical image technology and the implementation of medical image in human debases diagnose, typically Knee-OA is presented. Section 2.2 Image Processing, the basic idea of image processing technology is described. There are three image processing techniques: (i) Image Segmentation, (ii) Image Enhancement and (iii) Feature Extraction/Image Representation. The brief detail of each technique is presented in Sub-section 2.2.1, Sub-section 2.2.2 and Sub-section 2.2.3, respectively. The summary information of image classification is suggested in Section 2.3. The detail of the section is described as follows: (i) the fundamental idea of data mining (ii) the detail of image classification and (iii) the classification learning methods are presented Sub-section 2.3.1 to 2.3.3, respectively. The process after get the classification result is how to evaluate and measure the result. Thus, the evaluation measurement methods is discussed including: (i) Confusion Matrix, there are five evaluation measures also included (Accuracy, Recall, Sensitivity, Specificity and F-Measure), (ii) Area Under Curve (AUC) and (iii) The comparison measure which is a statistical model used to analyse the differences across multiple test attempts are presented in Sub-section 2.4.1 to 2.4.4, respectively. Finally, the summary of this chapter will be described in section 2.5.

2.2 Medical Image

Medical image has spread widely and has been an interdisciplinary to many research fields for example computer science, computer vision, engineering, statistic, biology and medicine. The technology in computer-aided diagnose processing has become an important for clinical healthcare. The medical image is used to diagnose inside human structure such as bone, brain and renal system. There are various types of medical image with respect to image capturing mechanisms including: (i) X-ray (radiography), (ii) Computed Tomography (CT), (iii) Magnetic Resonance Imaging (MRI), (iv) Ultra-sound, (v) Positron Emission Testing (PET) and (vi) Single Photon Emission Computed Tomography (SPECT). With respect to the work presented in this research, to act as a focus for the work X-ray image is considered. However there are three medical image capture mechanisms are presented in this section including: (i) X-ray, (ii) CT and (iii) MRI.

2.2.1 X-Ray Image

In 1895, a German physicist Wilhem Rontgen discovered X-ray image. The first medical image of X-ray image was made from his wife's hand. Wilhelm Rontgen took this radiograph of his wife's left hand on December 22, 1895, shortly after his discovery of X-rays. The image of Wilhelm's wife's hand is shown in Figure 1.



Figure 1 The image of Wilhelm wife left hand [44].

X-ray is a form of electromagnetic radiation which can pass through the human body with producing the internal structure image, the resulting of X-ray is an image, and this image is called a radiographic, also known as "X-Ray" or "plain film".



Figure 2 The difference radiographic density in X-ray image [45].

There are five types of principal radiographic densities in an X-ray image. The Figure 2 illustrates the difference of X-ray radiographic density indicating by the number including (i) number 1 is air (the darkest area), (ii) number 2 presents fat (the dark grey), (iii) number 3 is the soft tissue (grey), (iv) number 4 illustrates bone (bright grey) and (v) number 5 is the metal (the white/ the lightest area) [45].

Medical imaging is the technique and process of creating visual representations of the internal structures hidden by the skin and bones inside of a body such as bone, brain, lung and liver which used for medical analysis, as well as to diagnose and treat disease.

X-ray images were used for diagnose Knee-OA which presented in many works [4,20,21,22,23,24,25,26]. There are two main objectives for Knee-OA research: (i) first the X-ray image was used to detect OA and non OA [20,22], while the second one is more advance than the first one because the X-ray imagery was used to detect the OA stages [21,23,24,25,26].

On the purpose of the detection of the OA stage, the Kellgren-Lawrence (KL) system [27,28] is considered. The KL standard is a five levels indicating using 0 to 4. '0' represents no OA symptom while '4' represents the most serve radiographic disease. The summary of the OA stage is shown in Table.1. There are five stage of OA has shown in the table which rank from zero to fourth stage. The stage 0 refer to the normal case, within this stage the sign of OA cannot detect because there is no sign of OA has happened. The second stage of OA is the doubtful stage sometime called stage 1 of knee OA; in this stage, the sign of OA has appeared slightly. In other words, in the stagse 1 of OA there is a sign to show very minor bone spur growth. The next stage is the stage 2 which consider as the minimal stage, the knee joints in this stage will reveal greater bone spur growth, but the cartilage is usually still at a healthy size, i.e. the space between the bones is normal, and the bones are not rubbing or scraping one another. At this stage, synovial fluid is also typically still present at sufficient levels for normal joint motion. However, this is the stage where people may first begin experiencing symptoms; pain after a long day of walking or running, greater stiffness in the joint when it's not used for several hours, or tenderness when kneeling or bending. Stage 3 of OA is considered as "moderate" stage. In this stage, the cartilage between bones shows obvious damage, and the space between the bones begins to narrow. People with stage 3 OA of the knee are likely to experience frequent pain when walking, running, bending, or kneeling. They also may experience joint stiffness after sitting for long periods of time or when waking up in the morning. Joint swelling may be presented after extended periods of motion. Stage 4 OA is considered as "severe." People in stage 4 OA of the knee experience great pain and discomfort when they walk or move the joint. Because of the joint space between bones is dramatically reduced; the cartilage is almost completely gone, leaving the joint stiff and possibly immobile. The synovial fluid is decreased dramatically, and it no longer helps reduce the friction among the moving parts of a joint.

OA Stage	Stage 0	Stage 1	Stage 2	Stage 3	Stage 4
	(normal)	(doubtful)	(minimal)	(moderate)	(severe)
Properties	No sign of OA	Show very minor bone spur growth	Reveal greater bone spur growth	Cartilage was damage and narrowing of joint space	Great pain and discomfort when walking

Table 1 The summary of the OA Stages.

Figure 3 shows the X-ray image of each OA levels: Figure 3(a) presents stage 0 OA or normal, Figure 3 (b) illustrates stage 1 OA (doubtful), Figure 3(c) shows stage 2 OA the minimal level and Figure 3 (d) presents stage 3 moderate level.

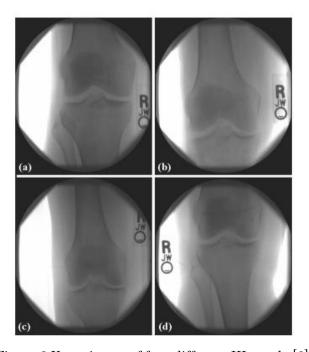


Figure 3 X-ray image of four different KL grade [6].

The work of OA stage detection using X-ray offers the benefit over the detection the OA and non-OA because knowing the level of OA can make medical doctor easy to prevent and slow down the OA speed to the OA patient. The most important of this part is the well-known of x-ray because X-ray is widely used in OA detection. The last benefit is the budget for x-ray imaging which is cheap than other medical imaging.

2.2.2 Computed Tomography

Computed Tomography (CT) is one of the well-known medical imagery for disease diagnosis. A CT scan uses a collection of X-ray images taken from different angles to produce cross-sectional image using the computer processing. Thus the structures inside of the body can be presented without cutting. CT provides the clear image so the medical doctor can be used to analyse or diagnose the disease.

2.2.3 Magnetic Resonance Imaging

Magnetic Resonance Imaging (MRI) uses the magnetic properties of spinning hydrogen atoms to make the internal structure image. It makes the powerful image about internal structure and very clear image. With respect to OA detection, the MRI images were also applied presented in [29,30,31,32]. The works show the benefit of OA analysis with short time consuming and produce a good result. The example of MRI image of human knee is shown in Figure 4.

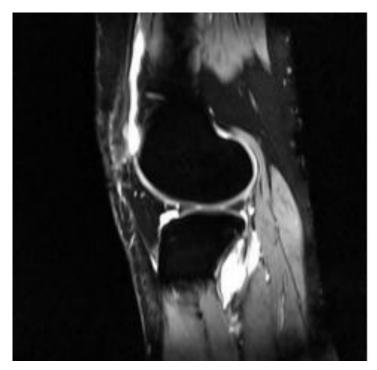


Figure 4 Knee MRI image.

To summarise, OA detection is the challenge task in medical and computer science. The image processing is applied to OA detection using medical image is the best solution as well as to analyse the OA stage. With the respect to this research to act as a focus in Thailand, so the medical X-ray image is considered with low cost of processing. With medical X-ray image apply with image processing to diagnose OA stage, the image processing technique is presented in the next section.

2.3 Image Processing

The fundamental concept image processing is discussed in this section. Image processing is a challenge area in computer science, which has the main purpose to analyze and manipulation of a digitised image. In addition, image processing has been applied to improve the image quality. There are main three key words in image processing that can be listed as follows:

(i) Image processing (input as image and output as image)

- (ii) Image Analysis (input as image and output as the measurement)
- (iii) Image Understanding (input as image and output as the high-level description)

An image is a set of array, a matrix, a square of pixel which arrange in columns and rows. In 1920s, the first image processing was discovered and it was called Bartlane cable picture transmission system. Nowadays, image processing has been applied in various applications such as education, engineering and medical. To get the deep meaning of image processing, the definition of an image should be known clearly first. An image is considered of two real variables, for example I(x,y), with I as the amplitude of the image at the real coordinate position (x,y). Subimage is the sub set of image, sometime it considers as the region of interest (ROIs). The smallest element of image is pixel. Image processing generally separates into two main categories: (i) digital image processing and medical image processing [49].

In general the medical image processing provides the image in white and black colour, or greyscale image in technical word. An 8-bits greyscale image has which picture element has an assigned intensity that ranges from 0 to 255. Figure 5 shows each pixel of greyscale image with the value from 0(black) to 255 (white) because 8 bit equal to 256 grayscales.

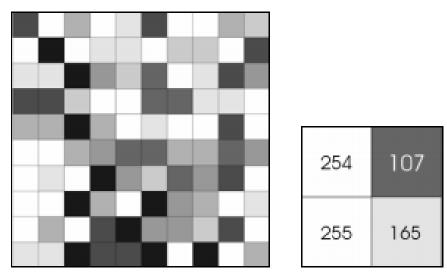


Figure 5 Each pixel has a value from 0 (black) to 255 (white) [46].

There are three image processing processes including: (i) image segmentation, (ii) image enhancement and (iii) image representation/Feature extraction. The detail of each process is discussed in the Sub-section 2.2.1 to 2.2.3 respectively.

2.3.1 Image Segmentation

Image segmentation is one of the most important processes of the image processing. Image segmentation is the division of an image into sub-image that is called region which correspond to different object. The main purpose of image segmentation is to divide the given image in to salient image regions, meaningful region, and homogeneous with specific cluster pixel. Moreover, object recognition, occlusion boundary estimation within motion or stereo systems, image compression, image editing, or image database look-up should be used in image segmentation field. A good segmentation of an image can be defined by: (i) pixel in the same region or categories has similar or equal pixel colour/intensity and (ii) the neighbouring pixel in different categories has different values or dissimilar values.

Although image segmentation is one of the main process in image processing, but it has many algorithms to applied with based on the different kind of work. Some of existing image segmentation methods that will be given the benefit to this research study should be included: (i) Segmentation based on threshold segmentation or clustering, (ii) segmentation based on region and (iii) segmentation based on edge. The brief detail of each technique is discussed in this subsection.

Threshold segmentation

Threshold segmentation is one the existing methods for image segmentation. The separation of light and dark region is the primary purpose of threshold segmentation. In general, the Thresholding is applied to create the binary image which make from grey-level ones by turning all pixels that below threshold become zero (0) and all pixel above threshold become one (1). For example: if I(x,y) is a threshold version of i(x,y) at some global threshold T, as equation 1 belows:

$$I(x,y) = \begin{cases} 1 \text{ if } i(x,y) \ge T \\ 0 \text{ otherwise} \end{cases}$$
 (1)

Threshold segmentation has a drawback which is only the pixel value is used to consider on other hand no relationship between the pixels is examined. Therefore, there is no any guarantee that the pixels identified by the thresholding process are contiguous.

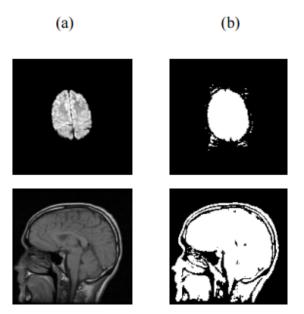
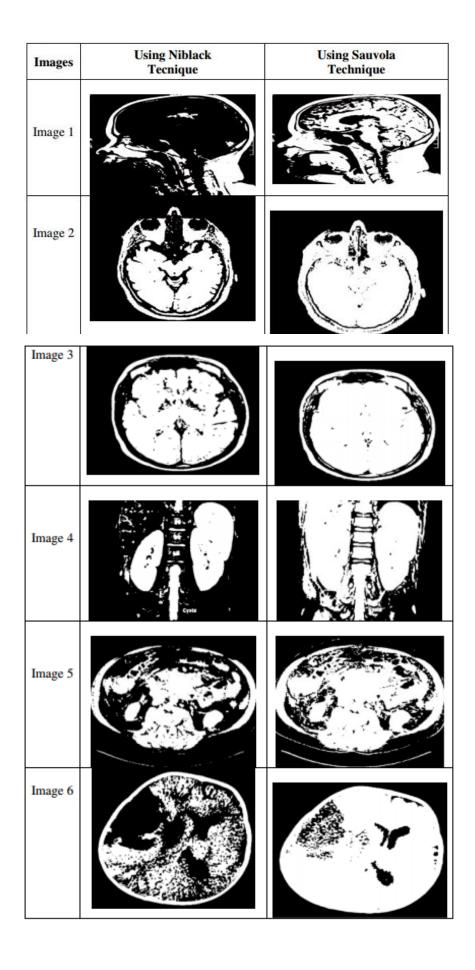


Figure 6 The human brain scan using Otsu method [31]

There are some applications which use threshold mechanisms were proposed for medical image segmentation. In [31] the implementation of Otsu method to improve medical image segmentation. The Otsu method is a method to reduce the greyscale to a binary image, the result shown that the method performs better than other thresholding methods with a good binary images. Otsu threshold method is a very good and efficient method to threshold the greyscale image. However, the drawback of this method was high complexity of the computation. Figure 6(a) shows the original of human brain scan and Figure 6(b) presents the corresponding segmented image using Otsu method.

Local adaptive thresholding also is applied in [32]. The work reported the implementation of local adaptive threshold technique to removes background using local mean and standard deviation. The Niblack and Sauvola local thresholing algorithm were used for comparison. The results shown that the Niblack local threshold algorithm reduced the background noise better than Sauvola local threshold algorithm. Figure 7 shows the difference of twelve sub-images of Niblack technique and Sauvola technique.



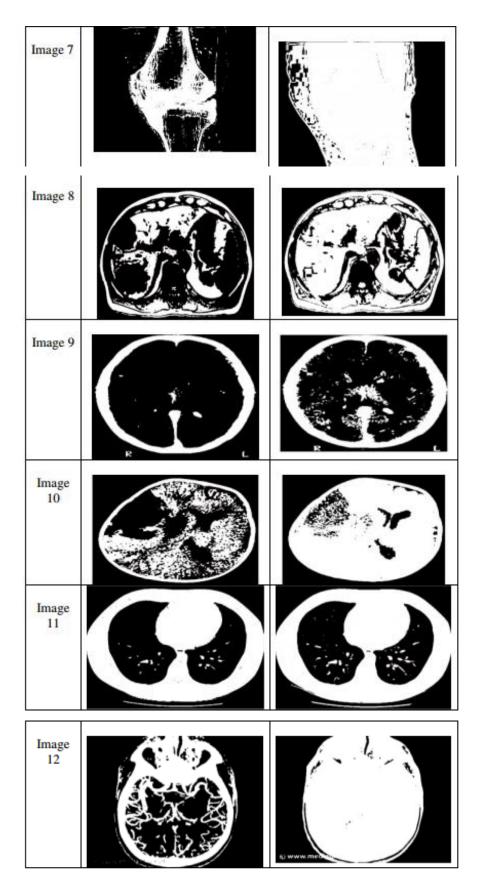


Figure 7 The comparison of 12 medical images by using Niblack and Sauvola Technique [32].

The local adaptive thresholding and Otsu thresholding method are the two techniques of many well-known techniques in medical image segmentation.

Region-Based Segmentation

Region-based segmentation is one of image segmentation techniques which has the main goal to find coherent regions in the image. The coherent region contains pixels which same or some similar property. The transitivity of the similarity relationship in the image is considered. Region-based segmentation offers the advantages on (i) fast processing and (ii) more efficient than edge-based and threshold method. However the region-based also has the drawback concerning it probably grow further away from global pixel because it drifts as one. There are many existing region-based segmentation algorithms for example watershed segmentation, flooding-based watershed segmentation, marker-controlled watershed segmentation and interpixel watershed segmentation. The implementation of watershed segmentation in the context of medical image with texture-based region merging was proposed in [33]. The work presented in [33] reports the texture-based region merging was applied to MRI images in order to improve the segmentation efficiency. The comparison of images using watershed segmentation with/without text-based region merging and with texture-based region-based region merging is illustrated in Figure 8.

Original region of interest	riginal region of interest Segmentation map using conventional algorithm without merging process	
V	72 partitions	12 partitions
	85 partitions	7 partitions
	81 partitions	9 partitions

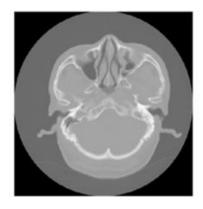
Figure 8 The comparison of image using watershed segmentation with/without texture-based [33]

Edge-Based Segmentation

In edge-based segmentation mechanism focuses on the edge in an image. The edge corresponds to singularities in the images. The edge in image generally represents as the shape of objects in the scene. The purpose of edge-base segmentation is to extract the edge or line in the image with good orientation. There are some existing edge operators for example Gradient operator, Prewitt operator, Sobel operator, Compass operators and Laplacian operator. Figure 9 shows the Prewitt and Sobel operator.

Figure 9 Prewitt and Sobel operator for edge detection

With the respect to the edge detection in the context of medical image segmentation, it can be found in [34]. The edge detection was applied to the brain scan images to detect the tumour regions based on the gradient magnitude information. The example image after applied the edge detection is shown in Figure 10.



(a) Original brain scan image



(b) The processed image using edge detection

Figure 10 The examle of brain scan image using edge detection[34]

From Figure 10(a) shows the original brain scan while Figure 10(b) illustrates the brain scan image after the edge detection was applied.

2.3.2 Image Enhancement

Image enhancement is one of the most important and complex techniques in image processing. The fundamental idea of image enhancement mechanism is to improve visual appearance of an image, or to present a better transform representation of the image. The main challenge in image enhancement is quantifying the criterion for enhancement.

There are some various types of images such as personal images, medical images, satellite images and aerial images. The photographer/user/viewer probably suffer from some image quality problems such as poor contrast and noise. Therefore image enhancement plays the main role to enhance the contrast and remove the noise to increase image quality. Typically, the image enhancement is applied is to improve the quality and perception of information contained in the image for human viewers. During image enhancement process an input is an image and the output also an image with better than the input image by changing the pixel intensity. Furthermore, image enhancement works as the important roles in many factors for example hyper spectral image processing, remote sensing, high definition television (HDTV), industrial X-ray image processing, microscopic imaging, other image/video processing applications and medical image processing [18].

In addition image enhancement technique can be used to increase dynamic range of the chosen features in of an image such as point operation, spatial operation, transform operation and pseudo colouring. There are some existing methods of image enhancement for example: Filtering with morphological operators, Histogram equalization, noise remove using a Wiener filter, Linear contrast adjustment, Median filtering, Unsharp mask filtering, contrast-limited adaptive histogram equalization (CLAHE) and Decorrelation stretch.

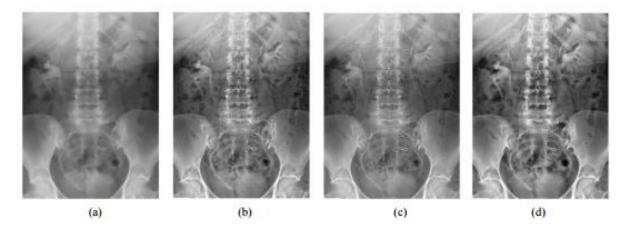


Figure 11 The example of X-ray image of renal system using morphological transform [35]

In the context of medical image, image enhancement also applied in many applications. Morphological transform was applied to medical images [35]. The work reported on using a disk shaped mask is used in Top-Hat and Bottom-Hat transform in order to remove noise and increase the contrast quantity in medical images. Figure 11 shows the morphological transform was used in X-ray of renal system. Figure 11(a) presents the original image. Figure 11(b) illustrates optimum enhanced image. While Figure 11(c) and Figure 11(d) shows the transformed image with different radius using CNR (Carrie to Noise Ratio) and PSNR (Peak Signal-to-Noise Ratio, respectively.

2.3.3 Feature extraction or Image representation

The fundamental information of feature extraction or image representation is presented in this sub-section. In image processing, feature extraction is the main rule for measuring the data and builds derived values (features) in order to find a good data representation. The objective of feature extraction in digital image is to reducing the amount of resources which required to describe a large set of data. Feature extraction is directed relation to dimensional reduction; dimensional reduction is the process of reducing the number of random variables under consideration [36]. With the respect to dimensional reduction there are some existing techniques including: Independent component analysis, Isomap, Kernel PCA, Latent semantic analysis, Partial least squares, Principal component analysis, Multifactor dimensionality reduction, Nonlinear dimensionality reduction, Multilinear Principal Component Analysis, Multilinear subspace learning, Semidefinite embedding, Autoencoder and Deep feature synthesis.

However in the context of image representation or feature extraction in digital images can be done using various using image properties. There are three main image properties including: (i) colour, (ii) texture and (iii) shape. The detail of each image feature extraction technique is described as follows:

Colour analysis

Colour is simplest features in digital image, it makes an image is more colourful and more interesting to human view. The colour analysis for image representation is a technique that can use widely with RGB image. RGB image is one type of digital image which contain RGB color system; RGB system refers to the system that representation in Red, Green and Blue through computer display view. The level of red, green and blue can range from 0 to 100 percent of full intensity. For each level of red, green and blue is represented by the range of decimal numbers from 0 to 255 (256 levels for each colour), or to be a binary is 000000000 or 111111111. RGB also know as true colour image. With respect to digital image which store as an m-by-n-by-3 data

array that defines red, green, and blue colour components for each individual pixel. The purpose of colour image analysis is to define or calculate the percent of each colour has contained in an image.

Texture analysis

The texture analysis concerns with description of characteristic image properties by textural features. Texture analysis method is broadly divided into three categories: (i) statistical methods, (ii) structural methods and (iii) model based methods [37]. A number of existing texture analysis models have been developed including autoregressive model, Gausian Markov random field, Gibbs random fields, Wold model, wavelet model, multichannel Gabor model and steerable pyramid.

Texture analysis is also implemented in the context of medical image. The example is found in [22] the long rage, standard deviation and greyscale entropy texture analysis techniques were applied to X-ray images to detect OA and non-OA, the Haralick feature was calculated in order to measure the texture image in term of contrast, correlation, sum of square, sum of average and homogeneity.

In [38] described texture analysis that texture is an important approach for describing the region. The texture can be defined as three different categories: (i) Statistical such as smooth and coarse, (ii) Structural and (iii) Spectral.

In [30], Frequency Filtering texture analysis was applied using Fourier transform to calculate filter data as the equation (2)

$$VD^* = F * VD (2)$$

where VD is the filter data from the convolution theorem [38] while F is the filter in the spatial domain.

The examples various filters are illustrated in Figure 12 including: (i) Low pass filter (Figure 12(a)), (ii) High pass filter (Figure 12(b)) and (iii) Band pass filter (Figure 12(c)). The results after applied given filters to human knee MRI image are presented in Figure 12(d), Figure 12(e) and Figure 12(f), respectively.

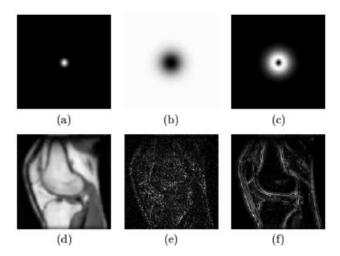


Figure 12 The example of various filters applied to MRI of human.

Shape analysis

Shape analysis is one of the branches in image science study which is directed with the edge of shape feature in image. Shape refers to all the geometrical information that remains when location, scale and rotational effects are filtered out from an object [39]. Shape is any connection of point in the image. The examples of some existing shape analysis mechanisms include: Point distribution, Active Shape, Active Appearance models, Fourier Snakes, Active Contours and Parametrically-deformable models. The shape analysis was applied in the context of medical image in a number of applications. With respect to the work in this research is directed to Knee OA, the existing examples are found in [20, 29, 19].

In [20] the shape analysis was used to calculate the thickness between the bone and joint space in knee, the seven-steps process was proposed: Step 1: map the intensity values in greyscale image to new values in such a way that few data is saturated at low and high intensities of image. Thus the contrast of the output image was improved. Step 2: threshold the image with its mean intensity values. Step 3: edge detection of the image. Step 4: cropping the image. Step 5: detecting the boundary of joint space. Step 6: making the binary for cropped image. And the final step (Step 7) the distance calculation. As the result the thickness was calculated and found that in the range of 1.69 mm to 2.55 mm was non-OA. In contrast if the thickness is less than 1.69 mm then the symptom of OA was possible. Figure 13 shows the cropping image and applies with binary operation to make edge detection in order to detect the boundary between the bone and joint space.



Figure 13 Cropping the image and apply in binary operation [20]

With the respect to the work presented in [19], the shape analysis was applied to calculate the area of cartilage. The initial step in the pixel-based segmentation is used to segment the cartilage. The texture filtering was then applied for calculating the area of cartilage by the number of pixel as presented in Figure 14.



Figure 14 MRI image of knee after the texture filtering [19]

In summary, the image segementation, image enhancement and image representation are the three main process in digital image processing. With the respect to the work presented in this thesis the image segementation, enhancement and representation are applied as the preprocessing process. The next section is 2.3 image classification is discussed.

2.4 Image Classification

The basic information of image classification is presented in this section. With the respect to this work in this thesis the image classification mechanism is commenced. This section is organised as follows: the fundamental idea of data mining is presented in Sub-section 2.3.1. Subsection 2.3.2 suggests the detail of image classification and finally Sub-section 2.3.3 presents the brief detail of classification learning methods.

2.4.1 Image Mining

Data mining is the process of discovering and searching insightful, interesting, useful, as well as descriptive, understandable and predictive models from large-scale data [40]. Data mining considered as the non-trivial of novel, implicit and actionable knowledge from large datasets. The alternative terms refer to knowledge mining from database, knowledge extraction, data analysis and data archaeology.

In addition, data mining is considered as the core element in the knowledge discovery in database (KDD) process .The phase knowledge discovery in database was suggested in the first

KDD workshop in 1989. The KDD processes including data selection, data cleaning, data transformation, pattern searching (data mining), finding presentation, finding interpretation and finding evaluation.

In [41] KDD consists of five processes: (i) Selection, (ii) Preprocessing, (iii) Transformation, (iv) Data mining and (v) Interpretation and evaluation as shown in Figure 15. From the Figure 15 can be observed that the KDD is a five-process mechanism. During selection process a large collection of data is selected. When selected data is delivered which only interested subset of data is choose the preprocessing process is then performed [41], an example of preprocessing process including the date attribute is computed into the day of week from a date field. After that the transformation process is then commenced in order to prepare appropriate data format to be ready for the data mining process. During the data mining process, the data mining algorithms or techniques are implemented. There existing data mining algorithms or techniques include: decision tree algorithms, logistic regression algorithms, and Naïve Bayes algorithm. When the pattern or hidden information is delivered the interpretation and evaluation process is applied to get the accurate knowledge.

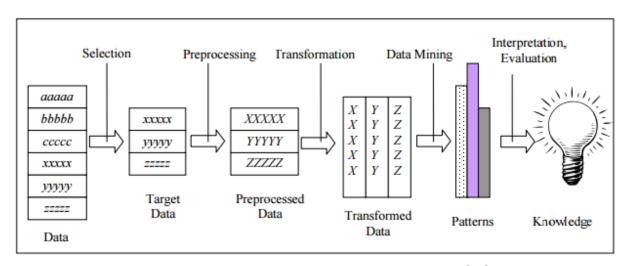


Figure 15 KDD process for producing the knowledge [31].

Data mining can be divided into two categories: (i) predictive data mining and (ii) descriptive data mining. Predictive data mining concerns with the prediction of behaviour based on historical data. In contrast descriptive data mining is directed at the discovery of patterns in existing data in order to use as the guideline for making the decision in the future.

Data mining can be applied to any kind of data repositories. In the traditional, data mining was applied to relational database (relational database is a database structure that can store data in tabular format which has two dimensional form row and column). In recently data mining can be applied to various of data formats such as spatial data, time-series data, free text, multimedia data, hypertext data and video and image data.

In [47] the task of data mining is generally divided into three main categories: (i) Classification, (ii) Association and (iii) Clustering.

Classification is the finding of a model that describes the data classes or concepts based on its attribute. The purpose of classification is to be able to use the derived model in order to predict the class of objects whose class label is unknown. Therefore, classification model can be used to classify the unknown class or future object and develop the classes of the objects in the database in order to get more understanding.

Association is the discovery which use to find the familiarity of identified patterns that are frequently together. Association rule reveals the associative relationship among the objects [47], for example: a Tesco in Phuket generates an association rule that shows that 50% of time milk is sold with bread and only 20% of times biscuits are sold with bread.

Clustering is the process to group of similar kind of objects. Clustering analysis refers to forming group of objects that are very similar to each other. On the other hands, they are highly different from the objects in other clusters.

In the context of image data, image mining is then suggested. Image mining is one of a form of data mining which dealing with the extraction of implicit knowledge, image data relationship or other patterns not explicitly stored in a collection of images [42]. Image mining consists of multiple components [43] such as: image analysis, image classification, image indexing, image retrieval and data management. The component of image mining is presented in Figure 16.

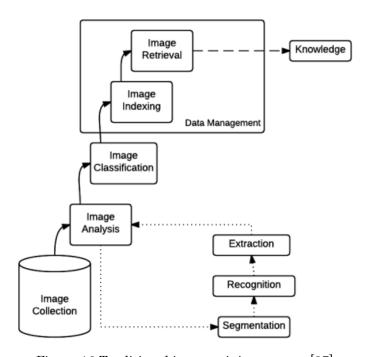


Figure 16 Traditional image mining process [27]

2.4.2 Image Classification

Image classification is a branch of image analysis, which refers to classify an image by separate or group the similarity properties from the image into the same category.

With respect to digital image, the image classification technique is applied to assign digital image to classes or categories using the pixel property. Image classification is also widely used in medical image, statelier image and remote sensing. The Pattern recognition is another cornerstone in computer science to help image classification classify an image, pattern recognition can be finding the similarities or patterns among small, decomposed problems that help to classify the complex image. In general, pattern recognition consists of three steps: (i) spectral pattern recognition, (ii) spatial pattern recognition and (iii) temporal pattern recognition.

Image classification process, the initial step is a collection of image (known-classes) is used as input data to generate the classification model which can be used to predict the class of unseen data in the future.

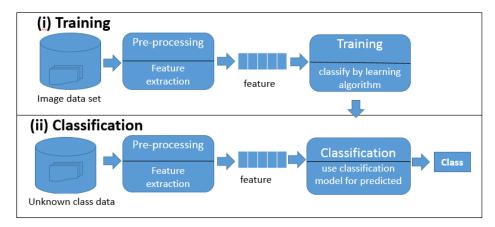


Figure 17 Image classification process

Image classification process is presented in Figure 17. From the Figure 17, image classification is a two sub-process: (i) training and (ii) classification.

During the training process, a collection of image data is applied using pre-processing process to enhance the quality of image and segment the area of interest. The feature extraction is then applied resulting in a feature vector format. The next step is a classifier generation, which classification learning methods are applied to construct the desired classifier. The training data set used in training process, the pre-label record that typically represented by using an n-dimensional feature vector representation which in turn a set of attribute values and a class label $\{i_1, i_2, ..., i_{n-1}, c_n\}$ where i_i is an attribute value and c_n is a class label that $c_n \in C$.

The second process is classification, in this process the generated classifier from training process is applied to the unknown/unseen image data in order to classify the data class.

As noted above, image classification technique has two main process where in the classification application process need to predict on the usage of discrete class label. In order to make prediction concern with the real value, regression is suggested.

Regression is a statistical technique which used to analyse the relationship between a response variable and one or more predictor variable. Liner regression and multiple linear regressions are the frequently is the existing technique for regression analysis. Linear regression is the one of well-known regression which use to analyse the correlation between predictor variable (known as the independent, explanatory, regress, input or exogenous variable) considered as x and response variable (known as dependent, target, output or endogenous variable) denoted by y.

Simple linear regression (SLR) is the most well-known of linear regression where in SLR have only one predictor x. The equation 3 below presented the simple regressing model.

$$y = a_0 + a_1 x$$
 (3)

where a_0 and a_1 are the constant for the regression. However, in real-life data the response variable (y) might have relevant incorrect related to each predictor variable (x) value. Residual value (e) is the differentiation of value between an actual response value and a predicted response value. The simple regressing model with residual value as shown in equation 4.

$$y = a_0 + a_1 x + e (4)$$

Multiple Linear Regression *(MLR)*, a number of predictor variable is used. The equation 5 presents MRL model for predicting the response value.

$$y = a_0 + \sum_{i=1}^{n} a_i x_i + e$$
 (5)

where y is the response variable, x_1 , x_2 ,, x_n are the predictor variables, e is a random error and a_0 , a_1 ,, a_n are the regression coefficients extracted from input data.

Linear regression has the limitation of linear regression such as: (i) a lack of auto correlation and (ii) limited to numerical response prediction

2.4.3 Classification Learning Methods

As mentioned before the work in this thesis is concerned to image classification. There are five classification learning methods are presented in this sub-section: (i) decision tree generator, (ii) Naïve Bayes, (iii) Logistic regression, (iv) Sequential Minimal optimization and (v) Neural Network.

Decision Tree

Decision Tree is considered as the most popular and widely use for classification learning methods. A decision tree is a mechanism to classify which is a recursive partition a collection of instances. Root tree is considered as the top node of the decision tree. In other words, decision tree is a classifier which instance by sorting them downs the tree from the root to some leaf node (also known as terminal or decision nodes). Root node is a direct tree with a node that has no coming edges. Every other node has exactly one coming edges. An internal or test node is the node that outgoing edge. In general, test node splits the instance space into two or more subspaces according to a certain discrete function of the input attributes values. Figure 18 illustrates the decision tree for the direct mailing response.

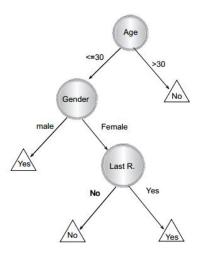


Figure 18 Response to Direct Mailing Decision Tree [48].

Naïve Bayes

Naïve Bayes has been widely implemented in clustering and classification. Naïve Bayes sometime called idiot Bayes, simple Bayes or Bayes classifier. The Bayes theorem is used in Naïve Bayes for prediction both in classification and clustering. The Bayes theorem is shown as the equation (6) below:

$$P = \frac{p(d|c_i) p(c_i)}{P(d)}$$
 (6)

where $p(c_i|d)$ is the probability of instance d being in class c_i , $p(d|c_i)$ is probability of generating instance d given class c_i , $p(c_i)$ is the probability of occurrence of class c_i and p(d) = probability of instance d occurring.

Logistic regression

Logistic regression is a statistical method which used to analyse a dataset. The dataset can be one or more independent variables which determine an outcome. The outcome is measured with a dichotomous variable (dependent variable normally is binary or dichotomous). The purpose of logistic regression is to discover the best fitting model for evaluating the relationship between a set of independent variables (predictor) and the dichotomous characteristic of interest (Outcome variable). Logistic regression provides the formula to predict a logit transformation of the probability as the Equation (7), where p is the probability of presence of the characteristic of interest.

$$logit(p) = b_0 + b_1 X_1 + b_2 X_2 + b_3 X_3 \dots + \dots b_k X_t$$
 (7)

Sequential Minimal optimization

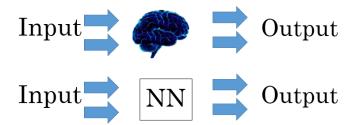
Sequential Minimal optimization(SMO) is a computer algorithm which is used to solve the quadratic programming(QP) problem that happen during the training of support vector machines(SVM). In order to solve the SVM QP problem, SMO decomposes SVM QP problem into QP sub-problems then solves the smallest possible optimization problem which involves two Lagrange multipliers, at each step. In other words, SMO algorithm can search with feasible region of two kind or type of problem then maximize the objective function. The equation (8) below is presented the SMO two α

$$\mathscr{L}_D \equiv \sum_{i=1}^{N} \alpha_i - \frac{1}{2} \sum_{i,j} \alpha_i \alpha_j y_i y_j \mathbf{x}_i \cdot \mathbf{x}_j,$$
(8)

With the respect to the research work, the knee feature (shape and texture) extraction will be able to use SMO for SVM classifier the OA stage.

Neural Network

Neural Network (NN) is a computer algorithm/system which sets the model design as the human brain and nervous system.



The use of neural network has 3 main different criteria such as (i) classification, (ii) noise reduction and (iii) Prediction. With the respect to the research study the implementation of NN in classification is proposed to x-ray image feature extraction.

2.5 Evaluation and Measurement

To evaluate the classification performance for the generated classifier, the evaluation measure is applied. With the respect to the work presented in this thesis there are a number of evaluation measurement is applied: (i) Confusion Matrix, i(i) Accuracy, (iii) Recall, (iv) Sensitivity, (v) Specificity, (vi) F-Measure, (vii) Area Under Curve (AUC) and (viii) Comparison using Friedman test. The detail of each measurement method will be discussed below:

2.5.1 Confusion Matrix

The confusion matrix is used to measure how well of applied data mining algorithm perform on a given data set. It can help to find out which data mining algorithm give the best performance or the worst performance. The confusion matrix is performed about the predict class (predict value) and actual class (actual value), Table 2 presents the confusion matrix.

		Predicted Class		
		True	False	
Actual	True	TP	TN	
Class	False	FP	FN	

Table 2 Confusion Matrix

From the Table 2, TP stands for True Positive rate used to measure the proportion of correctly identified positive test records. TN refers to the True Negative which used to measure the proportion of correctly identified negative test records. The FP is False Positive rate which used to measure the proportion of incorrectly identified positive test records and FN stands for False Negative which use to measure the proportion of incorrectly identified negative test records. There are a number of evaluation measures based on the confusion matrix including: (i) Accuracy, (ii) Rate error, (iii) Recall, (iv) Precision, (v) Sensitivity and (vi) Specificity. The equation for each measure is presented in the Equation 8 to Equation 13, respectively.

Accuracy =
$$(TP + FN) / T$$
, where $T = TP + TN + FP + FN$ (9)

Rate error =
$$(TN+FN) / T$$
 (10)

$$Recall = TP / (TP + TN)$$
 (11)

$$Precision = TP / (FP + TP)$$
 (12)

Sensitivity =
$$[TP / (TP + FN)] * 100$$
 (13)

Specificity =
$$[TN / (TN + FP)] *100$$
 (14)

2.5.2 Area Under Curve (AUC)

Area under curve (AUC) is considered as the best method for indicating the overall quality of the classifier [43]. The AUC can be calculated from ROC graph. ROC graph is a two-dimensional plot of the TP rate as the y-axis and the FP rate as the x axis [43]. The ROC graph is shown in Figure 18.

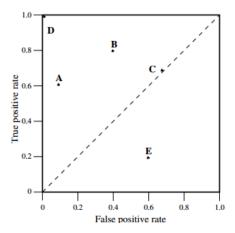


Figure 19 The ROC graph with 5 discrete classifiers [43]

AUC refers to the area under the ROC curve. Figure 19 presents the AUC in ROC graph which the random value from 0.0 to 1.0 in probabilistic classifier.

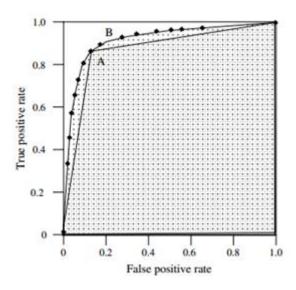


Figure 20 Area Under ROC Curve (AUC) [43]

2.5.3 Comparison Measure

Friedman test is the statistical test used for classifier comparison of classifiers or across multiple data sets. Friedman first introduced the Friedman test, which is a non-parametric alternative of the repeated-measure ANOVA. The Friedman test is the test to show the rank of algorithm which algorithm is the best performance. For the best-performing algorithm get the first rank, the second best-performing algorithm get the second rank and so on.

The Friedman test can be calculated using the Equation 14. Let r_i is the rank of the jth of k algorithms in the ith of N data sets. In this case, the implement of Friedman test for comparing the average ranks of algorithms, $R_j = (1/N) \sum_i r_i^j$, with null-hypothesis.

$$\chi_F^2 = \frac{12N}{k(k+1)} \left[\sum_j R_j^2 - \frac{k(k+1)^2}{4} \right]$$
 (15)

The X_F^2 with k-1 degrees of freedom, when N and k are big-enough (N>10 and k>5). Friedman's X_F^2 is undesirably conservative and derived a better statistic that was introduced by Iman and Davenport (1980) as shown in Equation 11.

$$F_F = \frac{(N-1)\chi_F^2}{N(k-1) - \chi_F^2} \tag{16}$$

which is distributed according to the F-distribution with k-1 and (k-1)(N-1) degrees of freedom.

2.6 The comparison of related work

This section the comparison of previous or related work, including OA detection, is discussed in this section The table 3 presents the comparison of related works in this research study:

Related work or	Image	Osteoarthritis	Osteoarthritis	Accuracy/result,
Previous work	Types	detection	Stage Detection	methodology
Sanjeevakumar	MRI		Just classified (KOA,	Calculate the thickness
Kubakaddi [29]			Normal and Doubtful	of cartilage (Shape
			KOA)	analysis).
Mahima	X-ray	✓		Calculate the thickness
Shanker				of cartilage with the
Pandey [20]				rate of result is higher
				than 65% (Shape
				analysis).
Lior Shamir	X-ray		Predict OA stage by	Use texture analysis to
[21]			the radiographic	predict the OA stage

			bone texture.	by comparing the value
				of each stage and
				another prediction use
				the values of predicted
				example to lower than
				the certain value.
Shivanand [22]	X-ray	OA/ Non OA		The result provides
		detection by		87.92 % by using
		using		different features
		different		calculations such as
		technique of		shape, statistical, first-
		classification		four, Haralick and
				texture.
Chao Jin [50]	Infrared	OA/Non OA		The result provides
		by extraction		85.49% of accuracy
		the patella-		rate, 85.72% of
		centering of		sensitivity and 85.51%
		knee feature.		of specificity buy using
				SVM classifier.

2.7 Summary

In conclusion, many types of medical images have been applied to diagnose OA disease with image processing techniques as mention above. As some examples [20-29] have been applied to various medical image to analyse OA/non-OA and OA stage by using image classification and image procession techniques. To summarise of this chapter is to tell a brief of each previous work had implemented in image processing techniques such as image enhancement, image segmentation, feature extraction, feature selection, evaluation and measurement. There were some brief studies for machine learning algorithms for example, neural network, decision tree and logistic regression could be used for classification process. The detailed of each process work in medical X-ray image analysis is discussed in Chapter 3.

Chapter 3

Framework and Methodology

3.1 Osteoarthritis Stage Mining

The proposed osteoarthritis stage mining framework is presented in this section. A schematic of the proposed framework is given in Figure 1. The framework supports a three-phases process: (i) Image preprocessing, (ii) Feature selection and (iii) Classifier Generation.

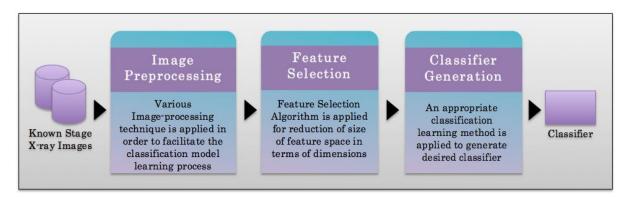


Figure 1 Proposed osteoarthritis stage from X-ray imagery framework

During the image preprocessing phase (left hand block in Figure 1) the X-ray image input data is prepared ready for the application of the classifier generation phase. There are various image processing techniques are applied: (i) Image enhancement; image quality improvement, (ii) Image segmentation; the whole X-ray image is segmented to obtained area of interest and (iii) Feature extraction; the appropriate format is generated in order to facilities the classification construction phase, image preprocessing phase is considered in more detail in next section (Section 3.3).

Feature selection is the second phase (centre block in Figure 1) in the proposed framework during which dimension space is reduced from the feature vector generated from labelled training produced during the preprocessing phase as described above. The final phase (right hand block in Figure 1) is a classifier generation where the desired classifier is generated from selected features (reduced dimension feature space).

3.2 Image Preprocessing

This section describes the detail of the image preprocessing. Figure 2 shows the schematic of the image preprocessing process. The preprocessing phase comprised three individual stages: (i) Image enhancement, (ii) Image segmentation and (iii) Feature extraction.

Image enhancement stage (left hand block in Figure 2) is applied to the input data. As mentioned in Sub-section 2.2.2 of Chapter 2 image enhancement of the mechanism for image

quality improvement, thus image enhancement is applied in order to the X-ray image is facilitated for the image segmentation stage.

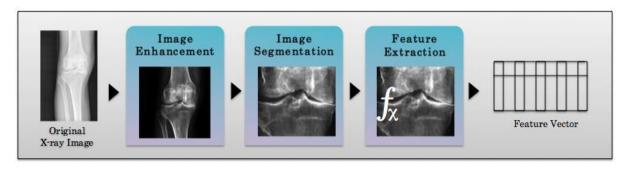


Figure 2 Schematic illustration the image preprocessing process

Once the image enhancement stage is completed the next stage is to segment the X-ray image. The second stage is image segmentation (centre block in Figure 2) is applied to X-ray image so that the area of interest (ROI) could be identified. The examples of ROI are presented in Figure 3. From the figure it can be seen that the enhanced X-ray image is shown in Figure 3(a). Once various segmentation techniques are applied so that the ROI is identified. The Figure 3(b) ROI is the area of joint space (the gap between femur and tibia) while Figure 3(c) ROI is the tibia (area under joint space).

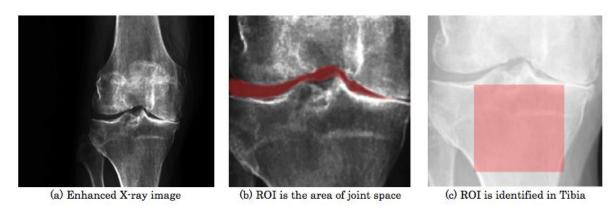


Figure 3 The example of identified ROI

The third stage is to represent the images in such a way that some image classification techniques could be applied. As already noted in Sub-section 2.2.3 of Chapter 2, there are a number of image representation or feature extraction techniques that have been proposed in the literature. Two feature extraction techniques used for the proposed osteoarthritis stage mining are considered: (i) a shape analysis and (ii) a texture analysis, which various statistics could be extracted. The main reason for choosing these only two features because an x-ray image is not a true color image, so the color feature cannot be applied in this research study.

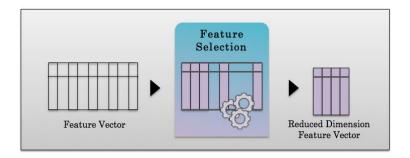


Figure 4 Schematic illustration the feature selection process

3.3 Feature Selection

Once a feature vector has been generated the next process in the framework (as shown in Figure 1) is feature selection to which is then applied so as to reduce the number of dimensions in the feature space, but in such a way that the reduced set still provided for a good discrimination between classes. The schematic illustration of the feature selection process is presented in Figure 4. Once the feature vector is identified using the feature extraction process described above, a feature selection algorithm is applied for selecting subset of relevant features is selected for usage in classifier generation process.

3.4 Classifier Generation

The classification generation process is discussed in this section. With respect to proposed osteoarthritis stage mining framework, during the generating the classifier process a selected feature vector represented X-ray image, with known the stage or grade of knee osteoarthritis, is used. The set of X-ray feature vectors are input into the classifier generator and a "machine learning" method applied to produce the desired classifier. Once the generation process is complete the produced classifier could be applied to predict the labels of unseen data. In other words to predict the class (stage or grade of OA) of a previously unseen set of X-ray images.

3.5 Evaluation Set Up

To evaluate the proposed process, labeled knee X-ray image as a training set is collected by Dr. Kittipon Naratrikul, MD, Hai Yai Hospital, Songkla, Thailand.

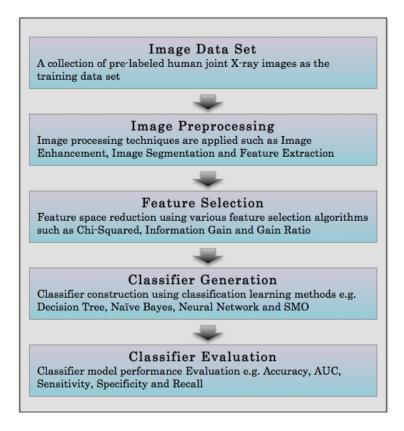


Figure 5 Schematic illustrate the evaluation set up

An overview of the experimental set up is given by the schematic shown in Figure 5. The X-ray image data set is processed as described in Section 3.3 image preprocessing above, Matrix Laboratory (MATLAB) is used to implement this process. Feature selection is then applied to reduce the number of feature to be considered to those that best served to distinguish between stages of osteoarthritis. Three feature selection strategies are considered: (i) Chi-Squared, (ii) Information Gain and (iii) Gain Ratio. Then the classification learning methods are applied. The Waikato Environment for Knowledge Analysis (WEKA) machine learning workbench is used for classifier generation process purposes. There is a number of classification learning mechanisms is commenced to generate the desired classifier the example includes: (i) Decision Tree, (ii) Naïve Bayes, (iii) Neural Network and (iv) Sequential Minimal Optimisation. The performance of each classifier is recorded in terms of: (i) Accuracy, (ii) AUC, (iii) Sensitivity, (iv) Specificity and (v) Recall.

3.6 Programme of work

A six phases programme of work is envisaged, with each phase comprising a series of "Work Packages", as presented in Table 1. Each phase has one or more deliverables associated with it. Note that phases three and four are directed specific types of technique, however, as the research progresses it may be that alternative techniques may be "switched in". It should also be noted that each phase has one or more deliverables associated with it which will be used to monitor the progress of the research.

WP	Start Date	End Date	Description	Deliverables					
Phase 1 Background work and Literature Review									
1.1	15 Aug 2016	15 Dec 2016	Preparing proposal	Research Proposal					
1.2	15 Aug 2016	15 Sep 2016	Review of Image Processing Techniques	Working Document 1					
1.3	15 Sep 2016	20 Nov 2016	Review of Data mining technique and Image classification	Working Document 2					
1.4	15 Aug 2016	20 Jan 2017	Production of literature review	Literature Review Document					
Phase 2	Phase 2 Review of data set and potential interpretation techniques.								
2.1	15 Jan 2017	20 Mar 2017	Preparing Data	Training set and test data collection					
2.2	30 Jan 2017	20 Feb 2017	Investigation of potential image enhancement and segmentation	Working Document 3					
Phase 3 Applying Classification Technique for a format 1 (Shape)									
3.1	20 Feb 2017	10 Mar 2017	Investigating techniques for converting images into format 1	Technical report for applying classification technique for the format 1					
3.2	10 Mar 2017	10 Apr 2017	Experiment using Classification techniques with the format 1						
3.3	10 Apr 2017	15 May 2017	Evaluation of proposed technique	Research paper (Proceeding 1)					
Phase 4	4 Applying Clas	sification Techr	nique for a format 2 (Texture)						
4.1	15 May 2017	15 Jun 2017	Investigating techniques for converting images into format 2	Technical report for applying classification technique for the format 2					
4.2	15 Jun 2017	20 Oct 2017	Experiment using Classification techniques with the format 2						
4.3	20 Oct 2017	15 Nov 2017	Evaluation of proposed technique	Research paper					
				(Proceeding 2)					
Phase 8	Phase 5 Refinement and Final Evaluation								
5.1	15 Nov 2017	30 Dec 2017	Final work and refinement of work	Final evaluation document and research paper (Journal)					
Phase 6	Phase 6 Thesis Writing								
6.1	10 Jan 2018	20 May 2018	Thesis writing using technical reports and research papers generated in earlier section as the foundation for the final thesis	Thesis					

Table 1 Programme of Work

3.7 Summary

To summarise, as noted above the methodology was introduced. The proposed osteoarthritis stage mining framework was suggested. The brief image preprocessing processing including: (i) image enhancement, (ii) image segmentation and (iii) feature extraction were presented. Feature selection and classifier generation process was then commenced. Finally, an overview of the experimental set up was identified ready for studying with respect to the work in this thesis.

Reference:

- [1] Arthritis Research UK, "Osteoarthritis", retrieved from "http://www.arthritisresearchuk.org", accessed December 18th 2016.
- [2] National Center for Biotechnology Information(ncbi), "Knee osteoarthritis prevalence, risk factors, pathogenesis and features: Part I", retrieved from "https://www.ncbi.nlm.nih.gov", accessed December 18th, 2016.
- [3] Kwankamon Dittakan and Frans Coenen, "Early Detection of Osteoarthritis Using Local Binary Patterns: A Study Directed at Human Joint Imagery", The 14th Pacific Rim International Conference on Artificial Intelligence (PRICAI 2016): Trends in Artificial Intelligence, Vol 9810 of the series Lecture Notes in Computer Science pp 93-105, August 2016.
- [4] P. Podsiadlo, L. Dahl, M. Englund, L. S. Lohmander and G. W. Stachowiak, "Difference in trabecula bone texture between knees with and without radiographic osteoarthritis detected by fractal method", Sciencedirect, Osteoarthritis and Cartilage, vol 16, 2007.
- [5] U. Fayyad, G. Piatetsky-Shapiro, and P. Smyth. "Advances in Knowledge Discovery and Data Mining: Toward a Unifying Framework", Knowledge Database Discovery-96 proceeding, American Association for Artificial Intelligence (www.aaai.com), pp. 82-88, 1996.
- [6] J.Han, M.Kamber and J Pei (2012), *Data Mining : Concepts and Techniques*. Waltham, MA:Morgan Kaufmam, USA, pp 8-10.
- [7] National Institute of Arthritis and Musculoskeletal and Skin Diseases (NIAMS), "Arthritis and Rheumatic Diseases", retrieved from "https://www.niams.nih.gov", accessed December 20th, 2016.
- [8] Arthritis Foundation, "What Is Arthritis?", Arthritis Foundation National Office, retrieved from "http://www.arthritis.org/" accessed December 20th, 2016.
- [9] Arthritic Research UK.: Osteoarthritis in general practice data and perspectives. Technical report, Arthritic Research UK, Copeman House, St Marys Gate, Chesterfield, S41 7TD (2013).

- [10] S. Zahurul, S. Zahidul, and R. Jidin. "An adept edge detection algorithm for human knee osteoarthritis images. In Signal Acquisition and Processing", International Conference on Signal Acquisition and Processing (ICSAP), pages 375–379, Feb 2010.
- [11] R. Alattas and B. Barkana, "A comparative study of brain volume changes in Alzheimer's disease using MRI scans", 2015 Long Island Systems, Applications and Technology, Farmingdale, NY, pp. 1-6, May 2015.
- [12] A. Udomchaiporn, F. Coenen, M. Garcia-Finana, and V. Vanessa Sluming, "3-D MRI brain scan classification using A point series based representation", Data Warehousing and Knowledge Discovery - 16th International Conference, DaWaK 2014, Munich, Germany, September 2-4, pages 300–307, 2014.
- [13] M. Nasir, A. Khanum, and A. Baig, "Classification of brain tumor types in MRI scans using normalized cross-correlation in polynomial domain", Frontiers of Information Technology (FIT), 2014 12th International Conference on, pages 280–285, Dec 2014.
- [14] S. Soh, T. T. Swee, S. S. Ying, C. Z. En, M. bin Mazenan, and L. K. Meng, "Magnetic resonance image segmentation for knee osteoarthritis using active shape models", The 2014 Biomedical Engineering International Conference (BMEiCON-2014), pages 1–5, Nov 2014.
- [15] S.Ababnehand and M.Gurcan, "An efficient graph cut segmentation for knee bone osteoarthritis medical images", 2010 IEEE International Conference on Electro/Information Technology, Normal, IL, pp. 1-4, May 2010.
- [16] J. Han and K. Ma, "Fuzzy colour histogram and its use in colour image retrieval", in IEEE Transactions on Image Processing, vol. 11, no. 8, Aug 2002, pp. 944-952.
- [17] R. C. Gonzalez and R. E. Woods, "Digital Image Processing. Pearson Prentice Hall", Third edition, 2007.
- [18] Xushu Zhang, Yuan Guo and Peng Du, "The Contour Detection and Extraction for Medical Images of Knee Joint", 2011 5th International Conference on Bioinformatics and Biomedical Engineering, Wuhan, pp. 1-4, 2011.
- [19] Bhagyashri L. Wagaj and M.M. Patil, "Osteoarthritis Disease Detection with the Help of Image Processing Technique", Emerging Application of Electronic System, Signal Processing and Computing Technologies Conference (NCESC 2015), pp 1-4, 2015.
- [20] Mahima Shanker Pandey, S S Soam and Surya Prakash Tripathi, "Detection of Knee Osteoarthritis Using X-Ray", ACEIT Conference, pp 1-4, 2016.
- [21] Lior Shamir, Salim Rahimi, Nikita Orlov, Luigi Ferrucci and Ilya G. Goldberg, "Progression Analysis and Stage Discovery in Continuous Physiological Process Using Image Computing", EURASIP Journal on Bioinformatics and Systems Biology, Vol 2010, Art ID 107036, pp. 1-7, 2010.
- [22] Shivanand S. Gornale, Pooja U. Patravali and Ramesh R. Manza, "Detection of Osteoarthritis using Knee X-Ray Image Analyses: A Machine Vision based Approach", International Journal of Computer Application, Vol 145-No.1, pp. 20-26, July 2015.

- [23] L. Shamir, S. M. Ling, W. Scott, M. Hochberg, L. Ferrucci and I. G. Goldberg, "Early detection of radiographic knee osteoarthritis using computer-aided analysis", Osteoarthritis and Cartilage (2009), Vol 17, pp. 1307-1312, Oct 2009.
- [24] Lior Shamir, "A computer analysis method for correcting knee X-rays with continuous indicator", International Journal of Computer Assisted Radiology and Surgery (2011), vol 6, pp. 699-704, March 2011.
- [25] Lior Shamir, Shari M. Ling, William W. Scott, Jr., Angelo Bos, Nikita Orlov, Tomas J. Macura, D. Mark Eckly, Luigi Ferrucci and Ilya G. Goldberg, "Knee X-Ray Image Analysis Method for Automated Detection of Osteoarthritis", in IEEE Transactions on Biomedical Engineering, vol. 56, no. 2, pp. 407-415, Feb. 2009.
- [26] Amir Herman, Ofir Chechik, Ganit Segal, Yona Kosashvili, Ran Lador, Moshe Salai, Amit Mor, Avi Elbaz and Amir Haim, "The correlation between radiographic knee OA and clinical symptoms-do we know everything?", Clin Rheumatol (2015), vol 34, pp.1955-1960, 2015.
- [27] J. H. Kellgren and J. S. Lawrence, "Radiologic assessment of osteoarthritis", Annals of the Rheumatic Diseases, vol. 16, pp. 494–501, Dec. 1957.
- [28] J. H. Kellgren, M. Jeffrey, and J. Ball, "Atlas of Standard Radiographs", Blackwell Scientific, Oxford, UK, 1963.
- [29] Sanjeevakumar Kubakaddi, Km Ravikumar and Harini DG, "Measurement of Cartilage Thickness for Early Detection of Knee Osteoarthritis (KOA)", 2013 IEEE Point-of-Care Healthcare Technologies (PHT), Bangalore, pp. 208-211, Jan. 2013.
- [30] P.N. Jebarani Sargunar and R. Sukanesh, "Texture Defect Detection Technique for MRI Knee Images", ACEEE International Journal on Signal and Image Processing Vol 1, No. 2, pp. 48-52, Feb. 2009.
- [31] Ch.Hima Bindu, "An Improved Medical Image Segmentation Algorithm Using Otsu Method", International Journal of Recent Trends in Engineering, Vol 2, No. 3, pp. 88-90 November 2009.
- [32] Senthilkumaran N and Vaithegi S, "Image Segmentation By Using Thresholding Techniques For Medical Images". Computer Science & Engineering: An International Journal (CSEIJ), Vol.6, No.1, pp. 1-13, February 2016
- [33] H.P. Ng, S. Huang, S.H. Ong, K.W.C Foong, P.S Goh and W.L. Nowinski, "Medical Image Segmentation Using Watershed Segmentation with Texture-Based Region Merging", 30th Annual International Conference of the IEEE Engineering in Medicine and Biology Society, Vancouver, BC, pp. 4039-4042, Oct. 2008.
- [34] Samir . Kumar Bandyopadhyay, "Edge Detection in Brain Images", (IJCSIT) International Journal of Computer Science and Information Technologies, Vol. 2 (2), pp. 884-887, 2011.
- [35] Raihan Firoz, Md. Shahjahan Ali, M. Nasir Uddin Khan, Md. Khalid Hossain, Md. Khairul Islam and Md. Shahinuzzaman, "Medical Image Enhancement Using Morphological Transformation.", Journal of Data Analysis and Information Processing, pp. 1-12, Feb. 2016.

- [36] Roweis, S. T. and Saul, L. K, "Nonlinear Dimensionality Reduction by Locally Linear Embedding". *Science*. Vol 290 (5500), pp. 2323–2326. Dec. 2000.
- [37] Jianguo Zhang and Tieniu Tan. "Brief review of invariant texture analysis methods". The journal of the pattern recognition society, Pattern Recognition 35, pp. 735–747, 2002.
- [38] Rafael C. Gonzalez, Richard E. Woods and Steven L. Eddins. "Digital Image Processing Using Matlab".(IND,2004).
- [39] I. L. Dryden and K. V. Mardia. "Statistical Shape Analysis. John Wiley & Sons", 1998.
- [40] Mohammed J. Zaki and Wanger Meira JR. "Data Mining and Analysis Fundamental Concepts and Algorithms". (USA,2014).
- [41] Gary M. Weiss and Brian D. Davison. "Handbook of Technology Management, H. Bidgoli (Ed.), John Wiley and Sons, 2010.". 2010.
- [42] M. C. Burl, C. Fowlkes, and J. Roden, "Mining for image content," Systemics, cybernetics, and informatics/information systems: analysis and synthesis, 1999.
- [43] Tom Fawcett. "An introduction to ROC analysis". Science direct on Pattern Recognition Letter, vol 27, pp.861–874, 2006.
- [44] Hannah Waters, "The First X-ray,1895", The Scientist, retrieve from "http://www.thescientist.com", accessed December 25th 2016.
- [45] David Lisle, "Imaging for Students Fourth Edition", (UK, 2012).
- [46] "Introduction to Image Processing", retrieved from https://www.spacetelescope.org/static/projects/fits_liberator/image_processing.pdf", accessed December 26th 2016.
- [47] Yongjian Fu. "Data Mining: Task, Techniques, and Application", Cleveland State University, retrieved from "http://academic.csuohio.edu/fuy/Pub/pot97.pdf", accessed December 26th 2016.
- [48] Lior Rokach and Oded Maimon. "Data Mining and Knowledge Discovery Handbook". (Israel, 2010).
- [49] M. Jayathi, "Comparative Study of Different Techniques Used for medical image Segmentation of Liver from Abdominal CT Scan", International Conference on Wireless Communications, Signal Processing and Networking (WiSPNET), Chennai, 2016, pp. 1462-1465, Sep. 2016.
- [50] Chao Jin, Yang Yang, Zu-Jun Xue, Ke-Min Liu and Jing Liu, "Automated Analysis Method for Screening Knee Osteoarthritis using Medical Infrared Thermography", Journal of Medical and Biological Engineering 33, vol 5, pp. 471-477, Feb.2012.