



A survey on Automatic Analysis of Statistical Properties of Rheumatoid Arthritis using Digital Image Processing

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ABSTRACT

Rheumatoid arthritis destroys joints of the body like erosion in bones which intern may cause deformity and characterized by inflammation of the tissue around the joints as well as in other organs of the body. At the beginning of this disease mainly the joints of hand and wrist are affected making hand radiograph analysis very important. Lately manual JSW measurement in hand X-ray digital radiograph of Arthritis patients were in use but it has disadvantages like inaccuracy, inter-reader variability. The reproducible quantification of the progression of joint space narrowing and the erosive bone destructions caused by RA is crucial during treatment and in imaging biomarkers in clinical trials. Current manual scoring methods exhibit interreader variability, even after intensive training, and thus, impede the efficient monitoring of the disease. Hand radiograph analysis is difficult for radiologist as there are 14 number of hand joints. To avoid observer dependency, computer-aided analysis is required. Wrist joint space narrowing is a main radiographic outcome of rheumatoid arthritis (RA). Yet, automatic radiographic wrist joint space width (JSW) quantification with statistical properties for RA patients has not been widely investigated. The automated analysis of statistical properties helps to reduce need of skilled personnel.

Keywords: digital image, joints, Rheumatoid Arthritis, statistical properties.

1. INTRODUCTION

Arthritis is a common bone disease that mainly affects the joints of the body; basically fingers, hands, knees. This may lead to disability, premature mortality and chronic ill-health. There are mainly two types of arthritis Rheumatoid arthritis (RA) and Osteoarthritis (OA). Rheumatoid arthritis is a complex disease in which the patient's immune system attacks its own tissue causing swelling and inflammation in the joints and damage of tissues and other organs. Rheumatoid arthritis (RA) causes pain, swelling, stiffness, and loss of function in the joints. Much of the rheumatoid arthritis research being conducted focuses on immune mechanisms involved. Rheumatoid arthritis destroys joints of the body like erosion in bones which intern may cause deformity and ankylosis in the later stage of the disease. At the beginning of this disease mainly the joints of hand and wrist are affected making hand radiograph analysis very important. Lately manual JSW measurement in hand X-ray digital radiograph of Arthritis patients were in use but it has disadvantages like inaccuracy, inter-reader variability. The causes of rheumatoid arthritis are still unknown, but we do know that autoimmune diseases have

a start and halt progression, with periods of active disease followed by periods of remission. Rheumatoid arthritis (RA) is characterized by inflammation within the joint space as well as erosion or destruction of the bone surface. Persistent inflammation over time can damage affected joints. The severity can vary from mild to severe. Treatments include disease-modifying medicines to suppress inflammation, which can prevent or delay the progression of the disease, and medication to ease pain. The earlier treatment is started, the less joint damage is likely to occur. Surgery is needed in some cases if a joint becomes badly damaged. The various imaging technologies were studied in order to evaluate the severity of disease. But, limitation exists in all the current technologies. The assessment of multiple joints with magnetic resonance imaging (MRI) is time-consuming and too expensive for routine use [16].

1.1 Stages of RA : Fig 1 shows the different stages of Rheumatoid arthritis

- Normal subject
- Early RA
- Advanced RA
- End-stage burnt-out RA

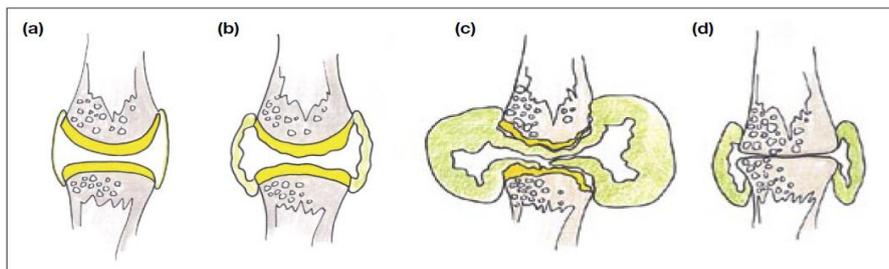


Fig 1. Stages of RA

1.2 The Use of Imaging Scans in the Detection and Monitoring of RA

For decades, X-ray images have been used to help detect RA and to monitor for the progression of bone damage. In early RA, however, X-rays may appear normal although the disease is active – making the films useful as a baseline but not much help in getting a timely diagnosis and treatment. Enter modern imaging techniques, including ultrasound and magnetic resonance imaging (MRI), which can reveal early, non-bony signs of RA that are invisible on X-ray. “Both MRI and ultrasound are more sensitive at detecting bone erosion than X-ray. That capability has become increasingly important with the development of ways to slow the RA disease process, before serious bone and joint damage occurs, using disease-modifying antirheumatic drugs (DMARDs), including biologics. Both ultrasound and MRI can detect inflammation of the lining of the joints, and tendon abnormalities. In addition, MRI detects areas of increased fluid (edema) in bone marrow that is a predictor for the development of bony erosions. While musculoskeletal MRI is quite expensive and requires an experienced radiologist to read, rheumatologists often have access to power Doppler ultrasound in their offices – used most frequently to guide joint aspirations and injections. Many rheumatologists are able to add ultrasound to their physical exam, but quantifiable measures of what distinguishes RA from normal on ultrasound (or how many joints need to be examined) have been lacking. In research presented at the American College of Rheumatology (ACR) annual meeting in November 2012, researchers from Brazil looked at how ultrasound

measurements taken in small, medium, and large joints could help physicians diagnose RA. For most joints, they were able to quantify a level of change that clearly distinguished RA patients from controls. However valuable ultrasound is not a solo diagnostic test for rheumatoid arthritis.

There are 35-70 million people are affected by RA in world wide. Since there is no proven cure for RA available yet, current treatments mainly focus on pain relief, inflammation reduction, and slowing down or stopping joint damage. In order to prevent irreversible joint damage, early detection of RA is essential. For an effective medical treatment it is important that the disease can be monitored closely. Joint damage assessment in hand radiographs is a frequently used method for monitoring the progression of RA.

The following questions about RA are still unanswered which motivate to do the research on RA.

- Where and why does RA start?
- Why does it attack the joints?
- Why doesn't it go away?

The motivation behind such system is that most of the previous related works of image analysis systems are not automated for the analysis of RA according to statistical properties like age factor and continuous progression of RA activity. There is not automated system for recognizing the severity of RA stage by seeing the image of affected area like hand, knee etc. Rheumatoid arthritis occurs as a result of uncontrolled joint inflammation, but it is not.

Fig 2 shows the details of joints of hand.

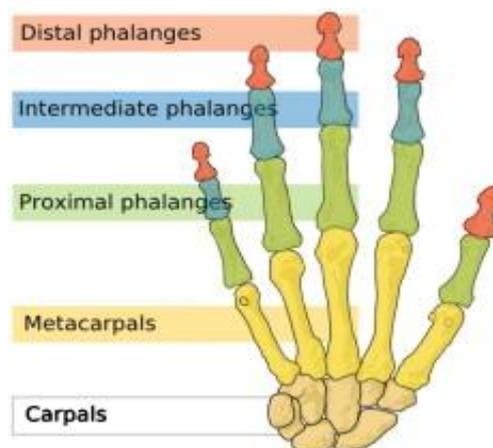


Fig 2 Joints of hands with names of joints plane

II. LITERATURE SURVEY

Rheumatoid Arthritis (RA) results in pain of primary peripheral joints like fingers, wrist and feet. This disease results in joint pain, stiffness, swelling of the joints which showed deformity and ankylosis in the late stages of the disease. So its detection is of at most important in early stages, Various researchers has proposed methods like MRI, X-Ray, Radiography, Thermography etc. for the detection of RA. For supporting research in RA Sumitra Nair[1] have proposed application based on machine learning for study of EMG images patterns for RA. Kernal LSK method is used for study of the limb displacement in space.

Georg Langs et al have proposed Automatic Quantification of Joint Space Narrowing and Erosions in RA [2]. This method does the quantification of radiographic changes in RA by measuring two indicators for disease



progression. Based on a hand radiograph bone positions and contour delineations are determined by algorithm. Cao Mills et al have proposed “Towards Quantitative Assessment of RA using Volumetric Ultrasound[3] This algorithm for segmenting 1) the 3 -D bone surface and 2) the 3-D joint capsule region. They extends 2-D bone extraction methods to 3-D and make algorithm more robust to the intensity loss due to surface normal’s facing away from incident acoustic beams. The extracted bone surfaces coupled with a joint-specific anatomical model are used to initialize a coarse localization of the joint capsule region. The joint capsule segmentation is refined iteratively utilizing a probabilistic speckle model.

Yinghe Huo et al. have proposed automatic joint detection in rheumatoid arthritis hand radiographs [5]. This method has focused on both joint location and joint margin detection. An automatic unsupervised joint location and joint margin detection method with a high detection rate was proposed. Secondly, the margin span is anatomically defined. The experiments are carried out on five 16-bit grayscale hand radiographs with resolution 2500×2000 pixels (0.1×0.1 mm). All detected joints are evaluated with the manual joint delineation. Syaiful Anam et al. have proposed automatic bone boundary detection in hand radiographs by using modified level set method and diffusion filter [6]. The method proposed by them has shown good extraction performance. However, the mask operation in this method could not work well for some images because the pixel intensities of bone and the pixel intensities of other areas are similar in some parts. Koay et al. [7] extracted ten features from thermal image for each quadrant of each breast. Quadrants were defined having nipple as a common point. Extracted features were skewness, standard deviation difference, mean temperature difference, entropy, maximum, median, minimum, kurtosis, area and heat content.. The SPSS statistical software was used to determine the correlation among the features after feature extraction.

Lipari and Head [8] used asymmetry between breasts and quadrants. The features extracted are: standard deviation median, mean, minimum and maximum value of temperature for each breast and Segmentation of Thermal Images for Evaluation of Rheumatoid Arthritis Disease quadrant. The paper didn’t reflect any results in sensibility or specificity, some comparisons of values between features extracted from each breast and quadrant. Kuruganti and Qi [9] segmented each breast in thermal images using generalized HT to extract parabolic curves defining the lower part of the breast. In return, the following features were extracted from the histogram of each breast region four moments mean, variance, skewness, kurtosis and entropy measures. They used six normal and eighteen cancer affected thermograms to validate the proposed classification method. It was observed by the authors that the high-order statistics skewness and kurtosis proved to be most effective features to measure asymmetry, while low-order statistics and entropy do not help in the detection of asymmetries. To validate their methodology, the authors had implemented the correlation measure to asymmetric analysis. From the set of features derived from the testing images, the existence of asymmetry is verified by computing the ratios of features of left to the right region of interest.

Schaefer et al. [10] proposed an application of content- based image retrieval (CBIR) to thermal medical images. CBIR allows the retrieval of same images based on features extracted directly from the image data. Padmavathi et al. [11] indicated that fuzzy c means algorithm with thresholding is an effective algorithm in segmenting underwater images. Zhou et al. [12] in their study indicated that fuzzy c means algorithm has been shown to be work well for clustering based segmentation. S. Belongie et al. [13] used color and texture based image segmentation using EM and its application to content based image retrieval.

Table 1: Summery of Literature survey

Paper	Feature extraction	Evaluation Method	Characteristics
Sumitra Nair[1]	Focused on EMG patterns of radiograph images	Kernal LSK algorithm nueral network linear discriminant algorithm	Subset of muscles are most differentiated
Georg Langs[2]	Automatic quantification of JSN and erosion in RA	JSN	Measures two indicators for disease progression
Cao Mills[3]	Volumetric ultrasound	Quantitative assessment of RA	Detected 3 D bone joints
Yinghe Huo et al. [5]	This method has focused on both joint location and joint margin detection	An automatic unsupervised joint location and joint margin detection method	Detected joints are evaluated with the manual joint delineation
Syaiful Anam et al. [6]	Aautomatic bone boundary detection in hand radiographs	By using modified level set method and diffusion filter	Automatic bone boundary detection by modified diffusion filter
Koay et al. [7]	Mean, standard deviation, median, maximum, minimum, skewness, kurtosis, entropy, area and heat content	Artificial neural network (ANN) with back propagation	At end, they use just two features: mean and standard deviation
Lipari and Head. [8]	Mean, median, standard deviation, maximum and minimum value of temperature for each mama and each quadrant	An algorithm of semi-automatic segmentation	Four reference points Each of these points connected to the nipple separates the image into four distinct quadrants.
Kuruganti and Qi. [9]	Means, variance, skewness, and kurtosis, the peak pixel intensity of the correlated image, entropy and joint entropy	Correlation measure to asymmetric analysis	Segmented each breast in thermal images using generalized HT to extract parabolic curves defining the lower part of the breast.
Schaefer et al. [10]	Basic statistical features, moments, histogram features, cross co-occurrence matrix, mutual information, and Fourier analysis	Fuzzy rule-based classification system	Statistical methods and fuzzy classification to diagnose breast cancer
Padmavathi et al. [11]	This method had focus on ROI	Fuzzy c means algorithm in segmentation	Fuzzy clustering algorithm videlicet fuzzy c-mean and adaptive fuzzy clustering algorithm
Zhou et al. [12]	Pixel intensities	Fuzzy C-means algorithm	Clustering based segmentation
S. Belongie et al. [13]	Color and texture based image segmentation	EM algorithm	Automatic segmentation based on image features

III. LIMITATIONS OF EXISTING LITERATURE

- Detection of RA in early stage repeatedly requires X-rays
- No automatic tool for detecting and suggesting the treatment for RA.
- The treatment available only slow down progression of RA
- No analysis of RA available by considering statistical characteristics of patient

IV. COMPARATIVE STUDY OF EXISTING LITERATURE

Table 2 shows the different algorithms accuracy result for hand x-ray input image .

Table 2: Different algorithms and their accuracy

Type of Input image	Algorithm	Accuracy
Hand x-ray	DTCWT based watershed algorithm	92.9
Hand X-ray	Joint space detection algorithm	96
Hand x-ray	shape LLM	95
Hand Knee x-ray	Backpropagation	82
Hand Knee x-ray	Perception	71

V.PROPOSED SYSTEM

The doctor examines each joint, looking for tenderness, swelling, warmth and painful or limited movement. The number and pattern of joints affected can also indicate RA. For example, RA tends to affect joints on both sides of the body. A specialist with specific training and skills diagnose and treat RA. In its early stages, RA may resemble other forms of inflammatory arthritis. No single test can confirm RA. To make a proper diagnosis, the rheumatologist will ask questions about the personal and family medical history perform a physical exam and order diagnostic tests. For performing the research, image processing and feature extraction techniques will be used. The block diagram of the proposed system is given as follows in Fig .2

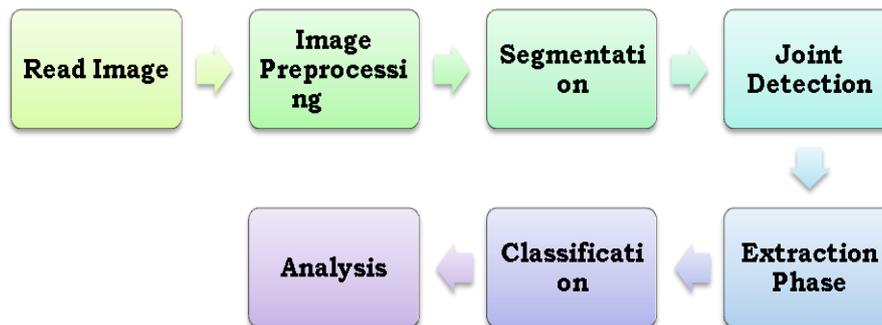


Fig 3 Proposed system block diagram

VI. CONCLUSION

Currently RA is treated based on x-ray images and preparing treatment plan. For the lab observation of positive subjects no automated or alternative system is available. Major of the work carried out till date concentrates on medical images like X-ray, CT, MRI, EMG. As no computerized system is available ,expert opinion is required every time. Automatic detection of JSN and erosion for predicting RA factor and ESR is the need of time which will definitely help the doctors to act as an automotive tool for predicting the erosion and RA in the early stage.

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