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Review Article

Comparative assessment of texture features for the identification of cancer in ultrasound images: A review

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ABSTRACT

In this paper, we review the use of texture features for cancer detection in Ultrasound (US) images of breast, prostate, thyroid, ovaries and liver for Computer Aided Diagnosis (CAD) systems. This paper shows that texture features are a valuable tool to extract diagnostically relevant information from US images. This information helps practitioners to discriminate normal from abnormal tissues. A drawback of some classes of texture features comes from their sensitivity to both changes in image resolution and grayscale levels. These limitations pose a considerable challenge to CAD systems, because the information content of a specific texture feature depends on the US imaging system and its setup. Our review shows that single classes of texture features are insufficient, if considered alone, to create robust CAD systems, which can help to solve practical problems, such as cancer screening. Therefore, we recommend that the CAD system design involves testing a wide range of texture features along with features obtained with other image processing methods. Having such a competitive testing phase helps the designer to select the best feature combination for a particular problem. This approach will lead to practical US based cancer detection systems which deliver real benefits to patients by improving the diagnosis accuracy while reducing health care cost.

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1. Introduction

23 In 2015, heart diseases were the leading cause of death in the United States and cancer was the second leading cause of 24 25 death. It is predicted that the order will reverse in the future [1]. Therefore, cancer is a big and growing public health problem 26 27 [2]. Table 1 lists public health data from the American cancer 28 society [3]. It shows both the estimated new cases and the 29 estimated deaths for ovarian, liver, thyroid, breast and prostate cancers. These cancers contribute 34.64% of all the 30 31 estimated new cancer cases and they are responsible for more 32 than 18.48% of cancer related deaths. In terms of public health, the problem can be partitioned into cancer prevention, 33 diagnosis and treatment. Cancer prevention is possible, 34 because healthy lifestyle choices lower the risk for developing 35 cancer. The link between lifestyle choices and cancer was 36 discovered by studies which showed that cancer rates of 37 migrants move towards the rate measured in the indigenous 38 39 population [4,5]. Smoking, consumption of calorie dense food 40 and reproductive behaviors are also known to increase the risk 41 of getting cancer [4].

42 Ultrasound (US) is a non-invasive, cost effective and safe¹ medical imaging modality which can be used to detect cancer 43 44 [6,7]. Achieving a good diagnosis performance with this imaging technology requires an integrate interplay of fine 45 motor skills (to operate the ultrasound transducer) and 46 47 cognitive abilities for image interpretation [8]. Hence, practitioners require extensive initial training and continuous 48 49 practice. A core problem of this human centric approach for disease diagnosis is the non-stationary diagnosis quality and 50 inter- as well as intra-operator variability [9]. Non-stationary 51 52 refers to the fact that the performance of human practitioners varies over time. These variations can be positive, such as 53 gaining more experience over time as well as negative 54 55 triggered by fatigue and other external factors. Overall, the 56 beneficial properties of US technology outweigh these pro-57 blems. Therefore, a vibrant research community explores a 58 wide range of application areas for this imaging methodology. 59 Initially, US was used only for application areas where tissue and bone formations led to sharp edges in the US images 60 [10,11]. Unfortunately, a wide range of diseases cannot be 61 diagnosed based on the edges within an US image alone 62 [12,13]. Many of the new application areas target diseases 63 whose symptoms and signs are changes in soft tissues [14]. A 64 65 prominent example of that problem class is cancer diagnosis, because cancer cells are very similar to normal cells. 66 67 Differentiating malignant from normal cells can be improved by interpreting image texture, since it contains information 68 about the scanned tissues [15]. For a human practitioner, the 69 70 changes in image texture, which indicate the presence of 71 cancer, appear to be minute. Hence, human texture analysis is 72 tiresome and error prone. As a consequence, a human centric 73 approach leads to a low diagnostic accuracy.

Computer Aided Diagnosis (CAD) can help to overcome the
problems of human texture analysis and thereby increase the
diagnosis accuracy [16,17]. The challenge for such computer
based texture analysis is twofold. First, we need to establish

Table 1 – Estimation of the number of new cases and death for selected cancers in the United states, 2015 [3].

Cancer type	Estimated new cases	Estimated deaths
Ovarian Liver and intrahepatic bile dust	21,290 35,660	14,180 24,550
Thyroid Breast Prostate	62,450 234,190 220,800	1950 40,730 27 540
All the cancer types from above together	574,390	108,950
All cancer types	1,658,370	589,430

mathematical definitions for relevant image textures. These mathematical definitions lead to texture analysis algorithms which can be used in practical CAD systems [18]. The second problem is to detect the texture changes, which indicate malignant tissues. It turns out that these problems cannot be solved a priori; the texture interpretation can only be done a posteriori. In other words, it is impossible to know what type of texture analysis algorithm will be sensitive for the subtle differences between normal and cancer cells in US images. Therefore, empirical methods identify which texture methods work well for a specific problem [19]. As a consequence, it is necessary to test a wide range of algorithms and select the ones which show the best performance on known data. Another complicating factor, for computer based texture analysis, is that most of the known texture algorithms depend on the image resolution. Hence, specific texture results are not transferable between different US capturing machines.

Texture information can be extracted using various methods. In order to select the best algorithm, it is necessary to have a good understanding of the available methods. The current review provides an overview of the available texture algorithms and their applications. We review texture-based US image analysis in the areas of breast, prostate, liver, ovarian and thyroid cancer detection. This review shows that texture features are vital for achieving the diagnostic accuracy needed for practical CAD systems. Furthermore, we give an overview of texture algorithms. During the review, we found that only a few CAD systems are solely based on texture methods. Research work, that considers only texture algorithms, aims to improve the understanding of the relationship between human tissue formations and US images. Robust and therefore practical CAD systems must be based on a range of different feature extraction methods, preferably coming from different imaging methods. We recognize that texture-based image analysis is a useful and cost effective enhancement of the wellknown US technology. For application areas, such as breast, liver, ovarian, prostate and thyroid cancer, US based texture features are vital for CAD.

To support our position, on texture analysis for medical US images, we have organized the article as follows. The next section provides pathological background on breast, prostate, liver, ovarian and thyroid cancer and their typical characteristics in ultrasound images. The material section contains a comprehensive review of texture algorithms and we introduce

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¹ It uses no ionizing radiation.

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