

An image analysis method for quantification of hepatic perfusion based on contrast-enhanced ultrasound imaging

Yi-xin Li, Fang Yang & Ning Gu*

State Key Laboratory of Bioelectronics, Jiangsu Laboratory for Biomaterials and Devices, School of Biological Science and Medical Engineering, Southeast University, Nanjing 210096, PR China. *Correspondence should be addressed to N. Gu (guning@seu.edu.cn).

ABSTRACT

Information about hepatic perfusion is used in clinical liver disease diagnosis. An image analysis system can help physicians make efficient and accurate diagnosis. The objective of this study is to propose an image analysis method for the quantification of the hepatic perfusion based on contrast-enhanced ultrasound imaging (CEUI). The proposed method contains frame selection, image registration, digital subtraction and grey-scale calculation. Then, by processing an image sequence, a time-intensity curve (TIC) for hepatic perfusion is derived. The kernel of this image analysis technology is digital subtraction and its accuracy is improved by frame selection and image registration. The advantage of this method is that it can obtain the perfusion information of the whole liver which is rarely obtained by traditional image analysis technology; therefore, it is a supplement of the traditional image analysis method. This method is applied on the quantification of a rabbit's hepatic perfusion and the result shows the efficiency of it.

Keywords: Hepatic perfusion quantification, Image analysis, Image registration, Digital subtraction

1. INTRODUCTION

Blood perfusion describes the amount of blood delivered to the capillary beds of a block of tissue in a certain period of time and its quantity determines whether the energy status of the tissue is likely to become compromised [1]. Thus, when blood perfusion of the tissue is abnormal, many diseases will be caused and on the contrary, when diseases occur, abnormal blood perfusion will be present. Various liver diseases lead to significant alterations of hepatic perfusion and quantification of hepatic perfusion has the potential to improve the assessment and management of liver diseases [2]. Consequently, much effort has

been done to quantify hepatic perfusion with the help of different imaging modalities, such as MRI, CT and so on [1-3].

As a versatile, non-invasive, low risk, low cost and portable real-time imaging technique, ultrasound is the most frequently used imaging modality [4]. CEUI is a new ultrasonic imaging modality developed with the introduction of ultrasonic contrast agent (UCA) and the evolution of corresponding imaging technologies [5]. Since CEUI can image the blood flow, it has been extensively used in clinical diagnosis [4,6,7]. One of the most important possible applications in the clinic is that it is more and more used to assess and quantify the blood flow and blood perfusion of various tissues and organs [8-12].

Efficient image analysis technology is needed during the quantification of hepatic perfusion using CEUI. The conventional analysis method is to measure mean signal intensity (mean grey-scale) of the pixels in a manual placed, equally sized and shaped region of interest (ROI) in each of the image frame, and then the TIC of the ROI is obtained [1,3,11]. Some important perfusion parameters, for example the mean transit time (MTT) and blood volume fraction (BVF), can be obtained easily using this analysis method. Despite its efficiency, there are still issues to be addressed. Since the manual placed ROI is usually local, the perfusion information of the whole liver such as the three vascular phases of the liver perfusion, as an important perfusion parameter, can hardly be derived directly from the TIC of the local ROI.

As a supplement of the traditional image analysis method, to reflect the three vascular phases of the liver perfusion directly, the proposed image analysis method in this paper focuses on the derivation of the TIC of all the automatically segmented contrast-enhanced areas rather than that of the local ROI. The segmentation is implemented by subtracting the baseline image (background) from the contrast-enhanced image. After the subtraction, all the contrast-enhanced areas (all the pixels with a grey-scale above zero in the subtraction image) are obtained. To verify the method, a rabbit's hepatic perfusion video is acquired using CEUI and is analyzed using the proposed method. The result of the application shows the efficiency

of the proposed method.

The layout of the paper is as follows: Section 2 presents the theoretical framework of the image analysis method. The experiment is described in section 3 along with results and discussion in section 4. The paper concludes in section 5.

2. METHOD

The main processes of this method can be described as follows: 1) a frame before the injection of the UCA is selected as the baseline image, and then one appropriate frame for each second after the injection is selected as the contrast-enhanced image, and all the contrast-enhanced images construct the image sequence; 2) for each of the contrast-enhanced image, the baseline image is registered and subtracted to acquire all the contrast-enhanced areas in the image; 3) the mean signal density of all the contrast-enhanced areas is calculated for every contrast-enhanced image, and then the TIC can be derived at last.

2.1. Selection of appropriate frames

Tissue motion induced by respiration makes the frames at different time point have lots of motion artefacts which will heavily decrease the efficiency of digital subtraction. Therefore, after the first frame is selected, all the other selected frames should have the least motion with the first frame. The purpose of this step is to reduce the global motion between the frames. In this method, the energy of the histogram of differences (EHD), proved to be computationally cheap and yield accurate results [13], is applied as the similarity measurement criterion and the frame with highest EHD has the greatest similarity with the first frame. The theory of EHD can be briefly described as follows:

$$M_{EHD}(d) = \sum_{g=-G}^G H^2(g) \quad (1)$$

Here $M_{EHD}(d)$ is the energy of the histogram of differences, $g \in [-G, G] \subseteq [-255, 255] \subset Z$. $H(g)$ is defined as:

$$H(g) = \frac{1}{KL} \sum_w \delta(I_d(x, y), g) \quad (2)$$

Here K, L are the width and height of the block to be processed (the image frame in this study); $I_d(x, y) = I(x, y) - I_0(x+\Delta x, y+\Delta y)$ is the subtracted image, where $I(x, y)$ is the contrast-enhanced image and $I_0(x+\Delta x, y+\Delta y)$ is the aligned baseline image. $\delta(x, y)$ is the Kronecker delta function.

$$\delta(x, y) \begin{cases} 1, x = y \\ 0, x \neq y \end{cases} \quad (3)$$

2.2. Image registration and digital subtraction

The global artefacts are reduced by selecting the appropriate frame; however, the local motion artefacts still exist between the contrast-enhanced image and baseline image.

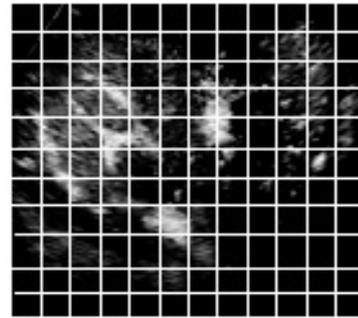


Figure 1. The schematic diagram for the selections of control points.

To remove the local motion artefacts, the baseline image is registered by finding the local motion vectors of the pixels in the baseline image and then aligning them. The most ideal method for registration is to align every pixel; however the computational cost makes this hard to be implemented. Given to the continuous motion of neighbouring area, an alternative approach is to align the selected control points first and then all the other pixels are aligned by interpolation. To acquire the control points, both the baseline image and the contrast-enhanced image are divided into an array of rectangular tiles as shown in Figure 1 and each vertex of the grid is selected as control point.

Based on the selection of control points, the whole process of the registration and digital subtraction can be described as follows:

Process 1:

Control points of both the images are selected as the centres and $R \times R$ pixels around each centre are set to be the ROIs. The ROI of baseline image is denoted as $A_b(i, j)$, and the corresponding area of the contrast-enhanced image is denoted as $A_c(i, j)$ ($i, j = 1 - R$).

Process 2:

$A_b(i, j)$ is shifted according to the rules of three-step search algorithm [14] and the position is located where $A_b(i+\Delta x, j+\Delta y)$ and $A_c(i, j)$ had the largest $M_{EHD}(d)$. The centre of $A_b(i+\Delta x, j+\Delta y)$ is recorded and pixel unit $(-\Delta x, -\Delta y)$ is defined as the local motion vector of the control point, where Δx and Δy are the number of pixels that have been shifted in x (horizontal) and y (vertical) directions. All the control points are processed by repeating this process.

Process 3:

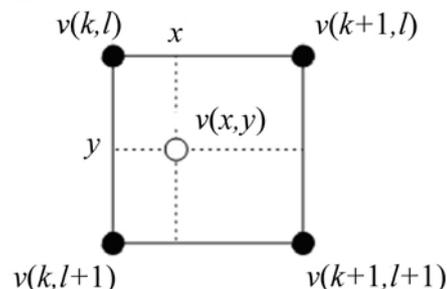


Figure 2. The bilinear interpolation.

The interpolation of local motion vectors is done in this step using bilinear interpolation, as shown in **Figure 2**. The interpolated value $v(x, y)$ at point (x, y) is calculated from the values of the four points surrounding the point (x, y) as follows:

$$\begin{aligned}
 v(x, y) = & \{(k+1) - x\} \{(l+1) - y\} v(k, l) \\
 & + \{(k+1) - x\} (y - l) v(k, l+1) \\
 & + (x - k) \{(l+1) - y\} v(k+1, l) \\
 & + (x - k) (y - l) v(k+1, l+1)
 \end{aligned} \tag{4}$$

Where $k = \lfloor x \rfloor$ and $l = \lfloor y \rfloor$ ($k = \lfloor x \rfloor$ denotes the largest integer that do not exceed x). After the interpolation, all the local motion vectors are obtained. Then, the registration of the baseline image is achieved by shifting the pixels according to the local motion vectors.

Process 4:

In this step, by subtracting the corrected baseline image from the contrast-enhanced image, the subtraction image is constructed and all the contrast-enhanced areas are preserved.

2.3. Derivation of TIC

Since the subtraction image contained only the contrast-enhanced areas, the mean signal intensity of the subtraction image is directly proportional to the amount of blood flow at this point of time. Thus, the TIC is acquired by performing following steps: (1) calculate the mean signal intensity of every subtraction image by calculating the average grey-scale of all the pixels in the contrast-enhanced areas; (2) let the horizontal axis denote the time while vertical axis the mean signal intensity.

3. EXPERIMENT

A rabbit's hepatic perfusion video was acquired using CEUI. The rabbit was anaesthetized using isoflurane gas (2-3%), and placed in dorsal recumbency on a heated imaging table (37°C) under anesthesia. Ventral abdominal

hair (around the liver) was removed using a depilatory cream. The imaging setup used in the experiment was LOGIQ® 9 (produced by General Electric (GE) Co.) and the scanning was carried out under a low power, low mechanical index (MI) contrast mode with the frequency of the transducer 3.5MHz and the MI value 0.2 (which was below the disruption threshold for the micro bubbles). All the conditions were fixed during the imaging time including the position of the ultrasound probe so that the motion was mainly led by the respiration of the rabbit. The UCA used in the experiment was provided by Jiangsu Laboratory for Biomaterials and Devices (China) and the synthesis and characterization of the UCA were described in another literature [15]. The UCA was injected by auricular vein bolus injection and the video began immediately after the injection. 120 seconds later, the video was stopped and each frame of the video was snapped and stored. In this experiment, the region contained the whole radial regions with the size 481×351 pixels were selected and the size of the block around the control point was 31×31 pixels (which was determined by balancing the computational cost and the accuracy of the registration).

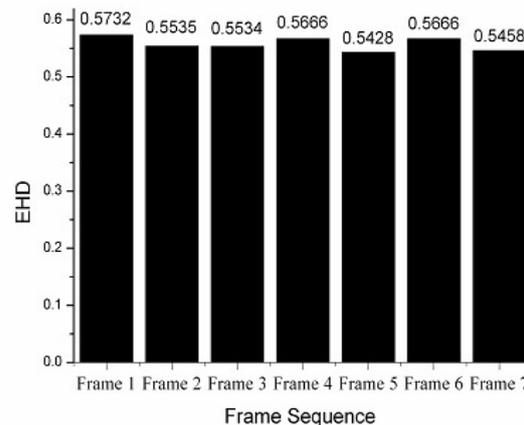


Figure 3. The EHD of each frame within one second

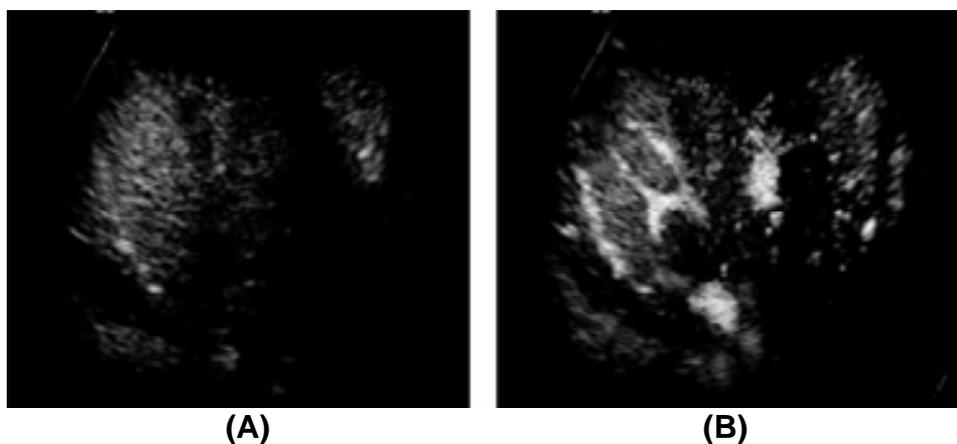


Figure 4. The result of frame selection: (A) the baseline image; (B) the contrast-enhanced image.

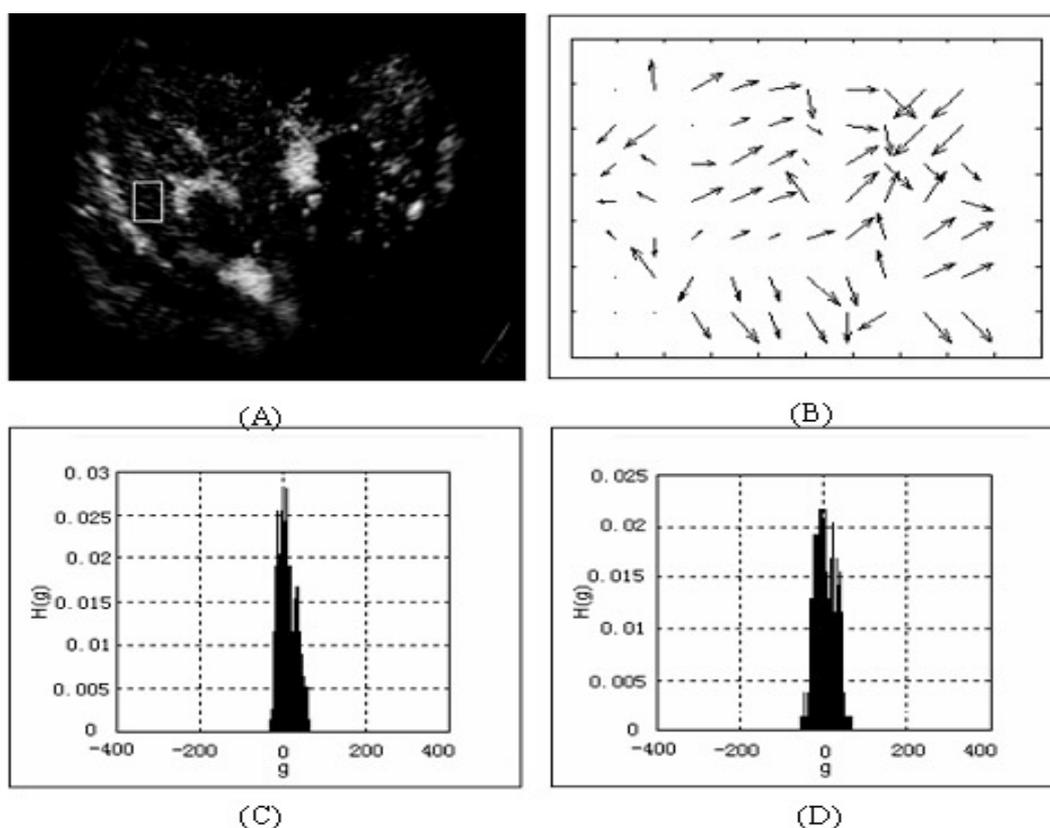


Figure 5. The results of registration and digital subtraction: (A) the subtraction image after registration and the selected ROI (in the rectangular); (B) motion vector chart for the control points; (C) the histogram of the ROI with registration; (D) the histogram of the ROI without registration.

4. RESULTS AND DISCUSSION

The video acquired contained seven frames for each second; therefore, in this study, the computer program calculated the EHD of seven frames for each second and then the frame with the highest EHD was selected to construct the sequence. **Figure 3** shows the EHD of the frames within the 17 seconds after injection and it can be found that the first frame has the largest EHD, thus, the first frame is selected as the contrast-enhanced image for the 17 seconds and **Figure 5** displays the contrast-enhanced image that is selected.

Figure 5 is the result of registration and digital subtraction for the images shown in **Figure 4**. From **Figure 5 (A)**, it can be found that the background is removed to a large extent. The motion vector shown in **Figure 5 (B)** illustrates that there is motion artefacts between the baseline image and the contrast-enhanced image. **Figure 5 (C)** and **(D)** show that after registration, the peak value around zero is higher and the band concentrates better at zero which means that the corresponding positions in the baseline image and the contrast-enhanced image are aligned efficiently, so the registration of the baseline image improves the background subtraction. However, **Figure 5 (A)** also shows that the background is not re-

moved completely and the reasons may include two main aspects: First, there are still some issues with the registration algorithm, for example, the registration algorithm only accounts for linear motion but not for rotation; Secondly, the imaging area and imaging direction changed slightly between the baseline image and the contrast-enhanced images which is inevitable because the probe must touch the body of the rabbit whose respiration means the probe will not be stationary.

It is reported that due to the specific blood supply to the liver, three vascular phases can be defined. The arterial phase (hepatic artery supply) starts 10–20s after intravenous injection and lasts for 10–15 s. It is immediately followed by the portal venous phase which extends from 30–35 s post-injection to 120 s. Thereafter, the late phase begins which ends with the disappearance of the bubbles (about 5 min post-injection) [10]. In **Figure 6**, the TIC begins to increase at about the 11 seconds and then quickly reaches the peak value which last until about 36 seconds; this indicates the arterial phase very well. While keeping high signal intensity, the TIC begins to decrease slightly at 36 seconds to 120 seconds, and this shows the process of portal venous phase very well.

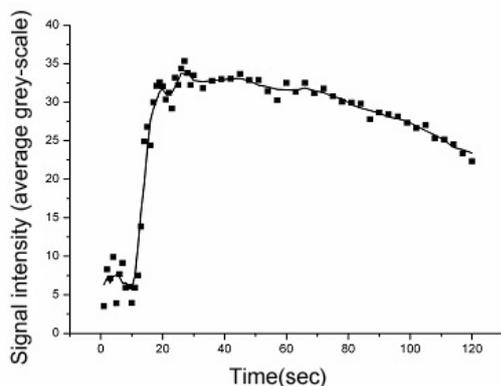


Figure 6. The TIC of the rabbit's hepatic perfusion obtained using the proposed method

5. CONCLUSION

In this study, as a supplement of the traditional image analysis method, an image analysis method used to quantify the total hepatic perfusion using CEUI is proposed.

The main contributions included in this method are frame selection, image registration, digital subtraction and grey-scale calculation. In the steps of frame selection and registration, EHD is selected as the criterion of similarity measure. The usefulness of the method is proved by applying it on the quantification of a rabbit's hepatic perfusion. Since ultrasound is the most widely used imaging modality, and as the development of the technology of CEUI, this method has great potential clinical value. In addition, it can also be used to quantify the perfusion of other organs such as the kidney. Although this method shows great potential advantages, there are still some issues to be addressed. First, the method is not accurate enough to make the background removed completely which affects the efficiency of quantification; secondly, the program for this method available now is off-line, which hampers the real-time clinical application. Further works are needed to address these issues so that the method can be better used for clinical diagnosis.

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