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Automatic Definition of Regions of Interest on Renal Scintigraphic Images

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Abstract

Renography is a radionuclide technique introduced in medicine since the early sixties in the daily practice. It allows measuring the function of each kidney separately in case of unilateral or bilateral uro-nephrological disease. A new system for analysis of Renography is presented in this paper. Our system aims to show a fully automated method for drawing the renal regions of interest based on a multi-agent system that incorporates spatio-temporal interest points detection on scintigraphic images by using the HOG3D descriptor to initialize agents. We used both types of agents, namely supervisor agents and explorer agents, they communicate among themselves and they inspire in their behavior from the Fast Marching method. Our system was tested on many real patients' cases and we have obtained encouraging results. Clinical validation showed a constant agreement between the approach presented and manual segmentation by professionals in the nuclear medicine.

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Keywords : Renography ; Segmentation; Automated Method; Regions of interest; Multi-agent system; HOG3D; Fast Marching Method.

1. Introduction

The use of renal dynamic scintigraphy (renography) has become essential to assess the renal function. Nowadays, nuclear medicine professionals want to have a complete software image processing for analyzing the renography by taking into consideration the standards of the international guidelines. Renal ROIs should

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be traced on an image sum based on renal function. The ROI must be larger than the renal contour of one pixel to obtain all the informations existing in the renal region. In our study, we propose to use the points of interest spatial and temporal detection of the initialization of the contour and we also propose to use a multi-agent system to accelerate this approach.

Multi-agent approaches have been used in several areas and have proved their efficiencies to solve distributed problems. The segmentation requires the elaboration of a certain number of particular functions and modular parts which have been designed for solving such problems.

Many approaches based on spatio-temporal descriptors and interest point detectors for segmentation have gained much attention [1,2].

In the present article, an automated system for the definition of renal regions of interest is presented based on Spatio-temporal Descriptor HOG3D by taking into account the advantage of the Fast Marching Method [7], which can promptly calculate the shorter path distances over two dimensional Euclidean spaces.

We will see the related works in section 2, our system will be presented in details in section 3, segmentation results will be discussed in section 4. Section 5 will be dedicated to conclusions and future research.

2. Related works

The definition of Renal ROI represents an important topic in the prevalent literature. Several algorithms have been suggested to define the Renal ROI which can be classified into two distinct categories:

- **Semi-automatic approach:** The methods using this approach permit drawing the renal ROI starting from an initialisation point. In [4] Y.Aribi et al explained the different modules used in their application with an explanation of the algorithm REGION_GROW that they have used for semi-automatic tracing of ROI. Another method has also been developed to generate renal ROI and evaluate the renal function in a small population [6].
Y.Aribi et al in [10] developed a semi-automatic system for segmentation of regions of interest kidney based on the Fast Marching method.
- **Automatic approach:** In this approach, these methods used are based on an automatic system without any intervention of the user. In [3] Daniel Ståhl et al present a fully automatic system for segmentation of kidneys and also distinguish non-functional regions of the kidney using the concept of compartments. In [5] the author presented a system based on an automated method for detecting the renal ROI and tracing the adequate renograms. In the state of the art, many approaches using intelligent techniques were proposed to define the concept of renal ROI. Our objective through this work is to describe an automated system for the definition of renal ROI using HOG3D descriptor and Fast Marching Method (FMM).

3. Techniques of our system

Our lab team working on video processing has developed a video objects segmentation and Tracking Based on Spatio-temporal Descriptor [12].

3.1. Multi-Agent System

For Weiss [8], an agent is a computational entity, it can be considered as perceiving and acting independently of its environment. In the multi-agent system there is a series of IT processes taking place at the same time, as a result multiple agents live simultaneously and share common property resources and

communicate together. In our work, we define a multi-agent system to segment the dynamic scintigraphic images sequences.

3.2. HOG3D Descriptor

The HOG3D is a descriptor based on the spatio-temporal aspects; it is based on 3D gradient orientations histograms. The integral image representation is essential for the calculation of gradients. For a standard quantification of the orientation of the spatiotemporal gradients, we can use the regular polyhedra. The shape and movement descriptors are combined together simultaneously. In our work, we apply the HOG3D descriptor on the video. For this purpose, the video is divided into $n_x \times n_y \times n_t$ cells [9].

3.3. Fast Marching Method

For J. A. Sethian [11], Fast Marching method (FMM) allows to solve the boundary value problems of the eikonal equation:

$$|\nabla u|f(x,y) = 1 \text{ Subject to } u|_S = g(x,y). \quad (1)$$

This method is only valid when the front is moving in the same direction. Thus, it can't be applied to the classical case where the contour evolves according to its curvature.

This method computes in each point (x, y) the transit time $T(x, y)$ of the front (each point is examined only once). One can show that T evolves according to $|\nabla T| \cdot F = 1$ (2), this means that the transit time is inversely related to speed of the front. As information is propagated in one direction, it can be the smallest T , and thus find the propagation of the front. The idea is to move the head in a narrow band, which is advanced by freezing the pixels existing in inserting new pixels in the band.

4. Proposed system

Our Multi-agent system, using an adaptive type of agents, is initialized with the HOG3D descriptor and it is inspired from their evolving from the fast marching method.

4.1. DESCRIPTION

The approach of our system is as follows, a pixel can be owned by one or many agents. We consider that an agent is just a pixel. Each agent is operating in the image by altering the original pixel. We can treat each pixel of the image on several occasions, but we do not need to cope with all of the pixels in the image. Our evolving model was based on the cooperation involving all the agents' explorer images that has a limited ability and a partly known environmentally and ensures better communication with their supervisor agents. Any explorer agent is required to consider its nearest neighbors. It changes to achieve the frontiers of the object to be extracted. In fact, any agent controls the destination of its development from the reports reaching from neighboring agents and the next evolution of the algorithm proposed thereafter.

4.2. The Proposed algorithm

Once we've got points based on interests' descriptor HOG3D, each agent will position itself on one point of interest, at that time each agent will receive the eight pixels around it. In Figure 1 we can observe the positioning of the neighboring pixels

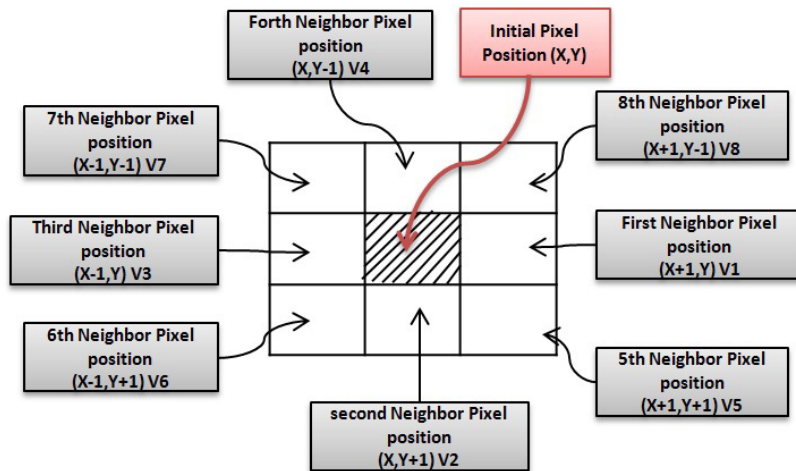


Fig. 1. Positions of pixels

After that, we need to calculate the pixel value which is the nearest neighbor's agents from the agent in progress. The evolution direction always follows the neighboring agents having the nearest pixel value.

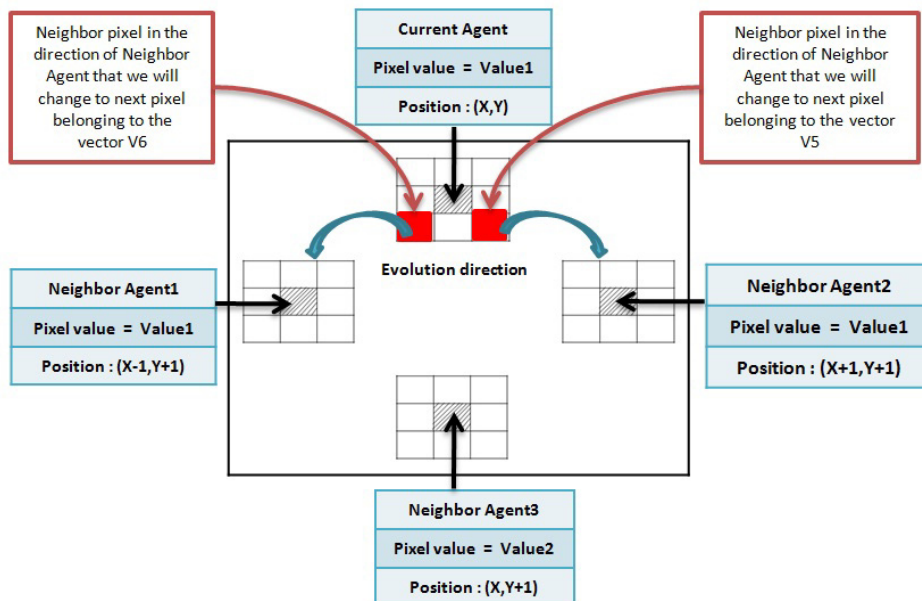


Fig. 2. Evolution direction of the pixels

If the pixel that we need to evolve is not owned by a different agent, the position of the original pixel must be modified to the pixel having the nearest value to the initial pixel. Then, the position of the original pixel will be modified with the nearest pixel value to the ancient pixel. In the opposite case, the supervisor agent reaction at this stage will handle the conflict and decide if such pixel is belonging to an agent. The seed

contour will be subdivided. The end of the algorithm is conditioned by the stabilization of all the agents.

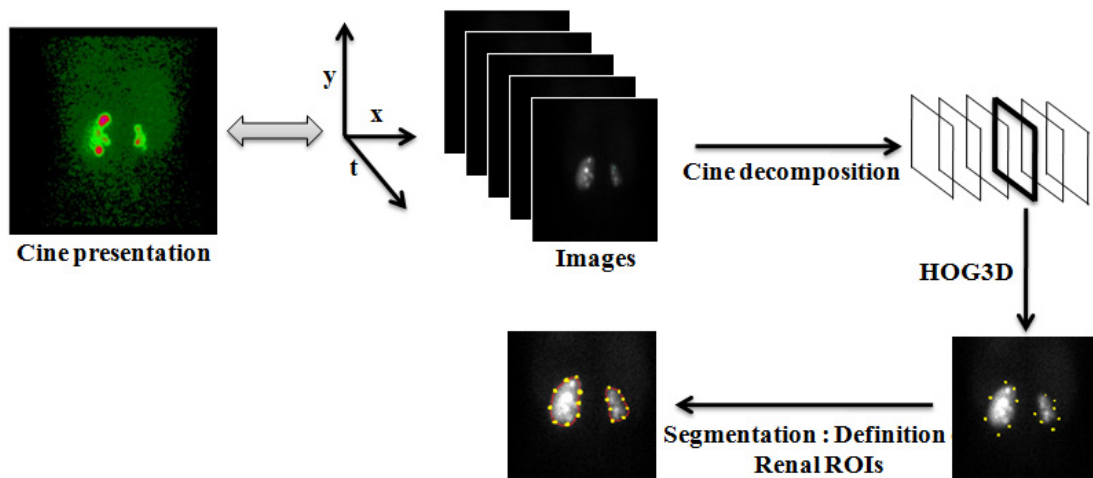


Fig. 3. Segmentation method

5. Discussion

For the evaluation of our system, we have studied real cases of patient images. Figure 4 shows the segmentation results using three different approaches. According to the experimental results, it is very clear that our system is better than other approaches in the accuracy of segmentation and also with regard to the time available for the execution.

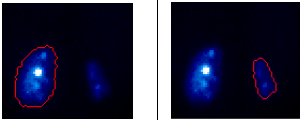

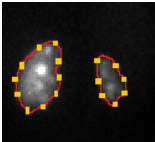
METHODS	Segmentation result	Execution time
Semi-automatic method based on Region_Grow [4]		17 s
Semi-automatic method based on FMM [10]		11 s
Automatic method based on the combinaison of : Multi-agent system, FMM and HOG3D descriptor		6 s

Fig. 4. Our approach vs two other segmentation method

6. Conclusion

Throughout this work, it became possible to make an automated system for detecting the renal ROI to facilitate the analysis of renography. Our system defined visually acceptable ROIs with no operator intervention. Automated ROI definition should enhance the efficiency of a good evaluation of renal function using a method based on the smart combination of three techniques: Multi-agent system, HOG3D descriptor and Fast Marching Method. To sum-up, the result of our system could be considered precise because we have used spatiotemporal descriptor. Finally, our system is a step forward in the field of medical image processing, waiting to receive expert advice.

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