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## ● Original Contribution

# AUTOMATED SEGMENTATION OF THE MEDIAN NERVE IN THE CARPAL TUNNEL USING U-NET

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**Abstract**—Nerve area and motion in carpal tunnel syndrome (CTS) are currently under investigation in terms of prognostic potential. Therefore, there is increasing interest in non-invasive measurement of the nerve using ultrasound. Manual segmentation is time consuming and subject to inter-rater variation, providing an opportunity for automation. Dynamic ultrasound images ( $n = 5560$ ) of carpal tunnels from 99 clinically diagnosed CTS patients were used to train a U-Net-shaped neural network. The best results from the U-Net were achieved with a location primer as initial region of interest for the segmentations during finger flexion (Dice coefficient = 0.88). This is comparable to the manual Dice measure of 0.92 and higher than the resulting automated Dice measure of wrist flexion (0.81). Although there is a dependency on image quality, a trained U-Net can reliably be used in the assessment of ultrasound-acquired median nerve size and mobility, considerably decreasing manual effort. (E-mail: [pamadio@mayo.edu](mailto:pamadio@mayo.edu)) © 2021 World Federation for Ultrasound in Medicine & Biology. All rights reserved.

**Key Words:** Ultrasound, U-Net, Carpal tunnel, Median nerve, Segmentation.

## INTRODUCTION

Carpal tunnel syndrome (CTS) is characterized by tingling and numbness in the distribution area of the median nerve caused by compression at the carpal tunnel. No widely accepted gold standard for CTS diagnosis exists. Commonly, CTS diagnoses are made based on the combination of clinical symptoms, physical examination and neuroconductive studies (NCSs) with or without electromyography (American Academy of Orthopedic Surgeons [AAoS] 2016). On the basis of severity, local guidelines and patient preference, treatment can include splinting, corticosteroid injections and/or surgical division of the transverse carpal ligament.

Ultrasonography has emerged as a relevant imaging modality for patients with CTS. Currently, it is used mostly as an addendum to the classic NCS during the diagnostic workup, with sensitivity and specificity reaching similar levels (Fowler et al. 2011) and high (intra-class correlation coefficient [ICC] >0.9) intra-rater reliability (Ghasemi-Esfe et al. 2011; El Miedany et al. 2015; Fowler et al. 2015; Gonzalez-Suarez et al. 2018). Studies indicate, however, that inter-rater reliability is more prone to variation, with reported ICC values between 0.57 and

0.94 (Van Doesburg et al. 2010; Gonzalez-Suarez et al. 2018; Schrier et al. 2019). Although NCSs serve as a direct measure for alterations in the function of the nerve, it is a test with limited ability to predict clinical severity or intervention outcome (Finsen and Russwurm 2001; Naranjo et al. 2009; AAoS 2016). Ultrasound allows direct visualization of the nerve, its position in the carpal tunnel and its relation to the surrounding structures, making it a helpful guidance tool, for example, during injection (Evers et al. 2017). Research has also suggested that median nerve cross-sectional area (CSA) at the level of the carpal tunnel inlet is a reliable diagnostic feature (Buchberger et al. 1991; Wong et al. 2004; Yesildag et al. 2004; Ziswiler et al. 2005). In addition to size, there is a decrease in nerve mobility in CTS patients (Nakamichi and Tachibana 1995; Erel et al. 2003; Van Doesburg et al. 2012; Wang et al. 2014; Nanno et al. 2015; Kuo et al. 2016; Ellis et al. 2017; Schrier et al. 2019), which can be quantified using dynamic ultrasound.

Segmentation of the nerve not only helps to acquire nerve size, but also allows for tracking during finger and wrist motion. Patients with CTS often have edematous, fibrotic connective tissue surrounding the median nerve, causing inhibition of nerve motion. Nanno et al. (2017) reported significant changes in median nerve motion after CTR. Nerve dynamics

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training of U-Net, images were sampled from 3336 (60% of the total) frames. One thousand one hundred twelve (20%) images were used for validation, and 1112 (20%) for tests.

*Neural Network design and software*

The ultrasound videos were converted from DICOM to PNG format using MATLAB 9.5 (The Math-Works, Natick, MA, USA). Subsequent parts of the system were programmed in Python 3 (Open source, Python Software Foundation, Beaverton, OR, USA) using a.o. Numpy and the TensorFlow 1.5 (Open source, Google, Mountain View, CA, USA) libraries.

An Artificial Neural Network, a U-Net model, was adapted from [Ronneberger et al. \(2015\)](#). Two implementations of the U-Net model were trained: The first model is based on single-frame segmentation. In this model, every single frame of a video is assessed individually without prospective information or feedback from earlier segmented frames. A second version of the model was made using focus windows. In this model, spatial information from the previous segmented frame is used to redirect the focus of the search area for the next frame, hypothetically increasing the odds of a correct segmentation. The first step in this model was to crop the frame to the relevant section—focus window—of the image.

Tests were performed with the segmentation of the first frame done manually and with the first frame segmented by the single-frame network (referred to as the third model, “fully automated”).

Figure 2 illustrates the overall layout of the U-Net model starting with a single image entering at the top left. A single frame (.png file) was downsized from 1024 × 768 to 512 × 384 pixels (in the single-frame

model) or, as in Figure 2, cropped to 320 × 240 pixels around the nerve when using focus windows. On the downslope—left side—of the U-shaped model, the information of the image is transformed with an added dimension (z-axis) containing abstract information and the condensed shape in the x- and y-directions, becoming smaller in the x- and y-directions and deeper in the z-direction. On the upslope—right side—of the U-shape the information is scaled back up to the original size of the image with a single depth, where the segmentation emerges with help of the location information from the same-sized shape from the left side as depicted by the horizontal arrows across the ‘U’. Finally, the image with the segmentation of 320 × 240 is positioned back in the 1024 × 768 image where it was cropped from. The arrows depicting convolutional filters and transpose convolutional filters are convolution operations with parameters that are trained using the neural network methodology, making U-Net a convolutional neural network.

The U-Net model used for the initial frame of the video uses images and masks of 512 × 384, the U-Net model used for the consecutive frames uses images and masks of 320 × 240 (no compression) unless the size of a nerve is larger than the 320 × 240 \* 0.25 mm/pixel = 8 mm × 6 mm.

The model with the focus window uses the mask of frame  $n - 1$  to determine the focus area for frame  $n$ . For the training set, however, the 11 annotated frames from the total of 334 frames in the video were not consecutive frames but 33 frames apart. To train the model to be adaptable to a moving nerve, as is seen during finger and wrist flexion, the focus window was subjected to a random geometric translation around the known mask  $n$  (Fig. 3).

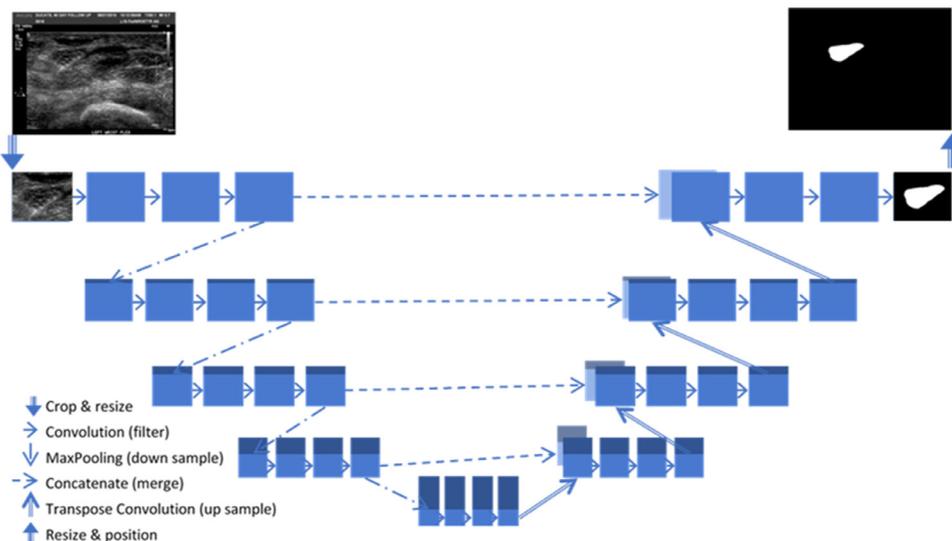


Fig. 2. Diagram of U-Net model.

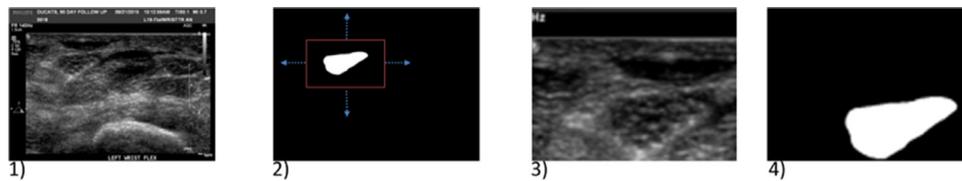


Fig. 3. From left to right: (1) Original frame  $n$ . (2) Truth mask  $n$  with a focus window to be translated randomly. (3) Translated focus window cut from frame  $n$ . (4) Cut from mask  $n$ .

### Primary outcome

During the training, as well as the final comparison with the manual segmentation, the mean Dice coefficient (95% confidence interval) was used as primary outcome. This measure represents the relative overlap of two segmentations, with “0” indicating no overlap at all and “1” representing perfect overlap. This measure therefore represents both morphological and spatial accuracy.

## RESULTS

The resulting Dice coefficients for both models are listed in Table 1. The first model was marginally outperformed by the focused frame model, and only if the first frame was defined manually by a user (0.82 vs. 0.85). Additionally, better results were acquired using images gained during finger flexion irrespective of U-Net model, but most prominently seen in the focused frames—manual approach (0.88 vs. 0.81). During the design it was noted that in particular cases, especially where the nerve exhibited larger excursions and the quality of the images was diminished, the resulting masks were less accurate, leading to a cascade of failed segmentations. After automated segmentation, the system produced an assortment of data for each frame of the video, providing potential research metrics (Fig. 4).

By use of the center point of the manual/automated overlay of the nerve, nerve size, perimeter and absolute coordinates were abstracted. The coordinates were used to track the nerve and calculate the overall motion (see Supplementary Video A [online only] for an example).

## DISCUSSION

This study found that a U-Net model can be used to accurately assess 2-D median nerve characteristics and mobility in ultrasound-based images using a medium-

sized data set. Performance can be enhanced by using a single manual segmentation to add prospective spatial information to the model.

In clinical practice, the system can aid in getting to a consistent method of diagnosing CTS using ultrasound. Another purpose of the model is to aid in predicting outcomes of CTS remedies. The computer easily outputs various data concerning the nerve: next to the standard measures area and perimeter, the movement of the center of the nerve compared with the previous frame or the change in shape compared with previous frame. With the model, this can be done for all the images in a video clip, creating a source of information on the dynamics of the nerve that, together with other structured data such as patient specifics and possibly existing NCS data, can be used to dial in on a predictive model informing both physician and patient on chances of success, facilitating their joint decision-making process.

With decreased nerve mobility as a potential marker for CTS diagnosis (Ellis *et al.* 2017) as well as prognosis in those undergoing injection (Schrier *et al.* 2019), the results of our study help in the analysis and assessment of the true added value of nerve mobility in clinical evaluation. Studies indicate that intra-rater reliability generally is high (ICC >0.9) when measuring the nerve cross-sectional area, but inter-rater reliability is lower, with reported ICC values between 0.57 and 0.94 (Van Doesburg *et al.* 2010; Gonzalez-Suarez *et al.* 2018; Schrier *et al.* 2019). Because the intra-rater performance of a U-Net model is 1.00, its added value is limited only by its accuracy. Through comparison of the accuracy with the aforementioned inter-rater accuracies starting at 0.57, the system performed better with scores up to 0.88. Still the main advantages are the increased number of measurements, consistency of measurement, as the algorithm will always provide the same segmentation, and

Table 1. Dice coefficients (95% confidence interval) for separate U-Net strategies

Data set	(1) Single frames	Focus windows	
		(2) Manual first frame	(3) First frame defined by “single frames” U-Net
Finger flexion (n = 594)	0.86 (0.85–0.88)	0.88 (0.87–0.89)	0.83 (0.81–0.86)
Wrist flexion (n = 518)	0.77 (0.75–0.80)	0.81 (0.79–0.83)	0.80 (0.78–0.82)
Both (n = 1112)	0.82 (0.81–0.84)	0.85 (0.84–0.86)	0.82 (0.81–0.83)

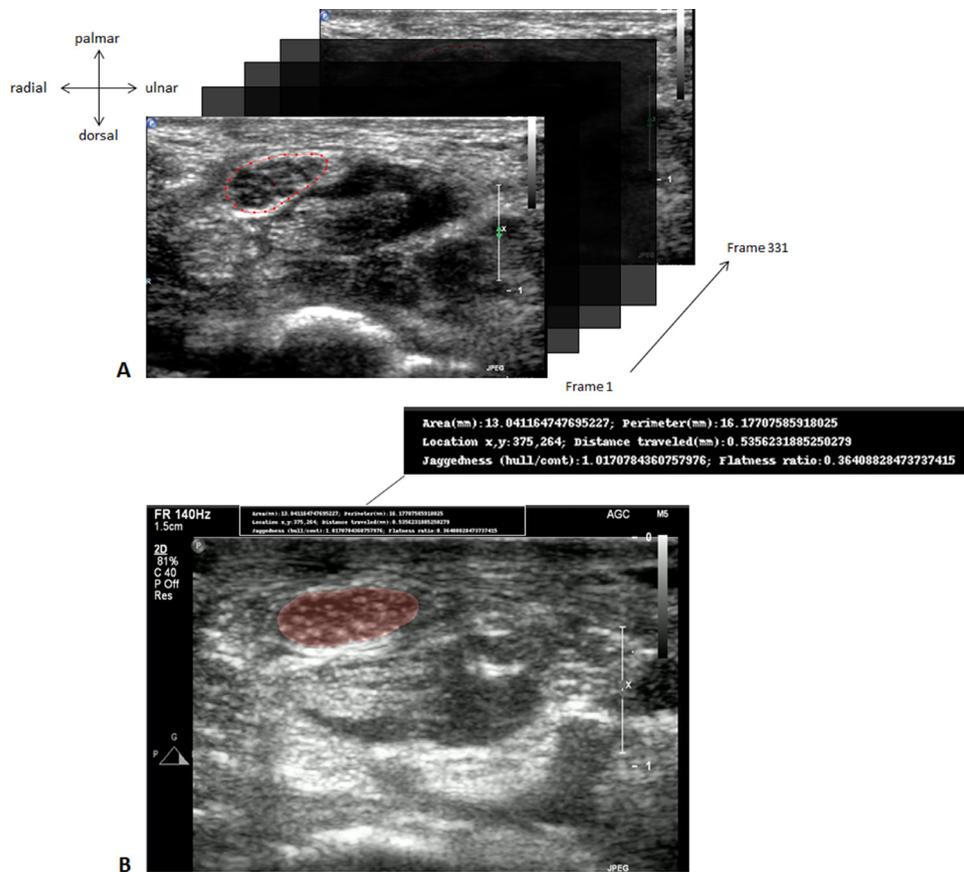


Fig. 4. (a) Example of an ultrasound image manually segmented (*red dotted outline*). A cine clip of 331 frames was recorded and manually segmented every 32nd frame to acquire 11 data points to quantify the nerve's motion. (b) Example of an ultrasound frame with an automated overlay in *orange*. At the top is a real-time display of several nerve parameters including area, perimeter and location coordinates.

speed. With an optimized hardware setup the analysis should be finished within minutes and be well applicable in the context of patient examination.

A challenge in dynamic measurements with US is the change in image features during motion created by the normal anisotropic characteristics of the nerve. Moreover, the shape of the nerve resembles the flexor tendons present in the carpal tunnel. Where a human interpreter would use anatomical knowledge and positive feedback tests (patient moving a finger), our model does not have this information. Consequently, the system can err from the truth. For practical usability it is important to know when this happens so the outlier can be eliminated from further analysis or perhaps a new clip of higher quality can be recorded. A way to cope with this is to add a visual check of the resulting video clip of 7 s. Improvements to the algorithm can be made, for instance, by employing Active Contour Model as post-processing step. If the visibility of the nerve can be maintained with wrist flexions, the model will perform better for those as well. Preprocessing of images for

sharpening, denoising, and so forth, is done by convolutional filters. As U-Net is a convolutional network, ideally its contents already cover those functionalities. However, established filters from classic—non-artificial intelligence—computer image analysis as a preprocessing step before U-Net could be tested to improve the results.

Limitations of this study are the single ultrasound system that was used; other probes will yield different images and will need additional training for the network to perform. Chances are there that the model misses the targeted segmentation. A visual check by the physician in the clinical context and an outlier detector in research context are recommended.

For future research we can look at the model itself with added preprocessing and postprocessing to increase accuracy. To extend the use of the model as a practical tool for a physician, it can be further developed to create a 3-D image of the median nerve. Finally, to use the information from the model to predict the probability of treatments being successful, further research is needed to

link the attained data with, for instance, NCSs of historic outcomes of those treatments.

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## SUPPLEMENTARY MATERIALS

Supplementary material associated with this article can be found in the online version at doi:10.1016/j.ultrasmedbio.2021.03.018.

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