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Nonrigid registration of three-dimensional ultrasound and magnetic resonance images of the carotid arteries

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Atherosclerosis at the carotid bifurcation can result in cerebral emboli, which in turn can block the blood supply to the brain causing ischemic strokes. Noninvasive imaging tools that better characterize arterial wall, and atherosclerotic plaque structure and composition may help to determine the factors which lead to the development of unstable lesions, and identify patients at risk of plaque disruption and stroke. Carotid magnetic resonance (MR) imaging allows for the characterization of carotid vessel wall and plaque composition, the characterization of normal and pathological arterial wall, the quantification of plaque size, and the detection of plaque integrity. On the other hand, various ultrasound (US) measurements have also been used to quantify atherosclerosis, carotid stenosis, intima-media thickness, total plaque volume, total plaque area, and vessel wall volume. Combining the complementary information provided by 3D MR and US carotid images may lead to a better understanding of the underlying compositional and textural factors that define plaque and wall vulnerability, which may lead to better and more effective stroke prevention strategies and patient management. Combining these images requires nonrigid registration to correct the nonlinear misalignments caused by relative twisting and bending in the neck due to different head positions during the two image acquisition sessions. The high degree of freedom and large number of parameters associated with existing nonrigid image registration methods causes several problems including unnatural plaque morphology alteration, high computational complexity, and low reliability. Thus, a “twisting and bending” model was used with only six parameters to model the normal movement of the neck for nonrigid registration. The registration technique was evaluated using 3D US and MR carotid images at two field strengths, 1.5 and 3.0 T, of the same subject acquired on the same day. The mean registration error between the segmented carotid artery wall boundaries in the target US image and the registered MR images was calculated using a distance-based error metric after applying a “twisting and bending” model based nonrigid registration algorithm. An average registration error of 1.4 ± 0.3 mm was obtained for 1.5 T MR and 1.5 ± 0.4 mm for 3.0 T MR,
I. INTRODUCTION

Stroke is a leading cause of death and long-term disability in North America. Approximately 85% of strokes are ischemic, and these are believed to result from blockage of a cerebral artery or arteriole. Atherosclerosis at the carotid bifurcation is a major cause of occlusion or cerebral emboli. A number of studies have shown that the majority of patients with highly stenotic, atherosclerotic carotid plaque remain asymptomatic, whereas patients with atherosclerotic plaques that are vulnerable to disruption, fracture, or fissuring can be at higher risk for embolization, occlusion, and consequent ischemic strokes. The development of noninvasive imaging tools that better characterize arterial wall, and atherosclerotic plaque structure and composition may help to determine the factors which lead to the development of unstable lesions, and identify patients at risk of plaque disruption and stroke. Furthermore, a better understanding of dynamic characteristics of tissue strain and wall shear stress in the presence of different plaque components may also lead to better predictive models for atherosclerotic lesion disruption.

As an emerging research technique, multicontrast weighted carotid magnetic resonance (MR) imaging allows for the characterization of carotid vessel wall and plaque composition, as well as the characterization of normal and pathological arterial wall, the quantification of plaque size, and the detection of plaque integrity. MR angiography (MRA) and high-resolution black-blood imaging of the vessel wall can be combined to evaluate the severity of stenotic lesions with their spatial distributions, and composition of the plaque for risk assessment and selection of treatment options. On the other hand, various ultrasound (US) measurements have also been used to quantify atherosclerosis both in research and clinical settings. Doppler US is also used in the clinical setting to assess the severity of carotid stenosis, which is a well-established stroke risk factor, by measuring peak systolic and peak diastolic velocity or their ratio. The one-dimensional US measurement of intima-media thickness (IMT) is also widely used in research for assessment of cardiovascular risk factors. Recent developments in 3D US imaging have allowed for the measurement of total plaque volume, total plaque area, and 3D US vessel wall volume. In addition, the ability of US to identify different carotid plaque components has also been investigated using carotid endarterectomy specimens for validation.

Combining the complementary information provided by 3D MR and US carotid images may lead to a better understanding of the underlying compositional and textural factors that define plaque and wall vulnerability, which may lead to better and more effective stroke prevention strategies and patient management. Furthermore, it may be advantageous to combine noninvasive MR plaque composition information with 3D US images, since it would provide a tool for in vivo validation of US plaque characterization techniques. However, the patient's head position is different in each image acquisition session causing relative bending and twisting of the carotid artery in 3D images. Thus, the image data acquired using the two imaging modalities would have relative nonlinear deformations between them, and these misalignments must be corrected by nonrigid image registration to allow proper registration of images acquired using these two modalities.

A few reports of carotid image registrations have appeared in the literature. Slomka et al. reported a 3D registration algorithm for carotid images, in which they used mutual information (MI) for rigid registration of 3D MRA images with power Doppler US (and indirectly with 3D US). Fei et al. reported an automatic rigid registration algorithm for multiple contrast weighted MR images of carotid vessels. Chan et al. reported the first nonrigid registration algorithm for 3D carotid image registration, in which they applied an unconstrained thin-plate-spline based deformable registration to carotid MR and 3D US images of ex vivo carotid artery samples and observed mean registration errors of about 1 mm. The authors previously presented a single modality (3D US) nonrigid registration algorithm for the registration of 3D US carotid images obtained at two different imaging sessions using a “twisting and bending” model to correct misalignments between images due to the movements of the neck. The authors successfully used the twisting and bending model to keep the deformations in the images within the natural neck movements during the registration to avoid any possible unnatural warping of the image that can mask the actual changes in the plaque structure.

Accordingly, this paper presents a method for multimodality nonrigid registration of 3D US carotid images with 3D MR images of the same subject, which extends the twisting and bending model-based 3D US to US nonrigid registration of carotid images that the authors previously developed. Here MI is used as the image similarity measure since it has been used successfully in US–MR image registration. The authors applied the proposed algorithm to 3D carotid US and MR images obtained from patients with carotid total plaque area >0.5 cm² and evaluated the registration accuracy using a distance based error metric between segmented vessel surfaces.
II. MATERIALS AND METHODS

II.A. Nonrigid registration

In registration, any point, \((x,y,z)\) in a 3D MR image of the carotid artery (source) relates to a point in the reference 3D US image of the same region (target), which is acquired separately, using the optimal transformation,

\[
T:(x,y,z) \mapsto (x',y',z').
\]

(1)

In general, different head positions in different image acquisition sessions cause nonlinear deformations between the images, so that rigid transformation alone is not sufficient for registration. Therefore, a combined transformation, \(T\), which consists of a rigid transformation for coarse alignment and a nonrigid transformation for fine alignment, is required,

\[
T(x,y,z) = T_{\text{nonrigid}}[T_{\text{rigid}}(x,y,z)].
\]

(2)

II.A.1. Rigid transformation

Rigid transformation, in 3D, has six degrees of freedom, describing three rotations, \(\theta_x, \theta_y, \theta_z\), and the three translations, \(t_x, t_y, t_z\), and is defined in the following.

\[
T_{\text{rigid}}(x,y,z) = \mathbf{R}_{\theta_x, \theta_y, \theta_z}(x, y, z) + (t_x, t_y, t_z),
\]

(3)

where \(\mathbf{R}_{\theta_x, \theta_y, \theta_z}\) is the standard rotation matrix for angles \(\theta_x, \theta_y, \theta_z\).

II.A.2. Nonrigid transformation: Twisting and bending model

In previous work, the authors modeled the movement of the neck using a twisting and bending model with only six parameters for a nonrigid transformation to be used for carotid 3D US image registration,\(^{36}\) as summarized in the following. The twisting and bending model limits deformations to the natural neck movements only, avoiding unnatural changes to plaque morphology during the registration. The authors’ nonrigid transformation consists of two parts, twisting and bending,

\[
T_{\text{nonrigid}}(x,y,z) = T_{\text{bend}}[T_{\text{twist}}(x,y,z)].
\]

(4)

The twisting model assumes that the 3D carotid image is a part of a homogeneous cylindrical shaft, which is clamped at the bottom plane of the \(z\) axis (at \(z=0\)), as shown in Fig. 1(a). The center of the model, the “twist–bend center” defined by three parameters \((X_c, Y_c, Z_c)\), does not move during the deformation. The twisting is defined by one parameter, \(Q\), where \(|Q|\) represents the magnitude of the vector torque, \(Q\), and \(Q > 0\) describes clockwise and \(Q < 0\) describes counterclockwise twisting. Any plane perpendicular to the twisting axis is not distorted under the homogeneous cylindrical shaft assumptions.\(^{36}\) Each plane normal to the \(z\) axis rotates around the axis with a different angle, \(\phi(z)\), as shown in Fig. 1(a) and the transformation for twisting, \(T_{\text{twist}}(x,y,z)\), is then given by

\[
T_{\text{twist}}(x,y,z) = \begin{pmatrix} t_{11} & t_{12} & 0 \\ t_{21} & t_{22} & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x-X_c \\ y-Y_c \\ z-Z_c \end{pmatrix} + \begin{pmatrix} X_c \\ Y_c \\ Z_c \end{pmatrix},
\]

(5)

where \(t_{11} = \cos[\phi(z)],\ t_{12} = -\sin[\phi(z)],\ t_{21} = \sin[\phi(z)],\ t_{22} = \cos[\phi(z)]\), \(\phi(z) = (\pi/4L)Qz\).

The bending was modeled assuming a pure bending condition under equal and opposite bending moments, \(M\), acting at the two ends of the \(z\) axis.\(^{36}\) The 3D carotid image is assumed to be a part of a rectangular block for the bending model, with length, \(2L\) in the \(z\) direction and centered at the twist–bend center \((X_c, Y_c, Z_c)\) as shown in Fig. 1(b), where \(L\) is the length of the 3D carotid image along the \(z\) axis. The bending model has two parameters defining the two components, \(M_x\) and \(M_y\), of the bending moment, \(M\), in the two directions \(x\) and \(y\), respectively. The component of the bending moment along the \(z\) axis is always zero since the bending moment, \(M\), is perpendicular to the \(z\) axis. Thus, the direction of the bending is defined by the angle to the \(x\) axis, \(\theta_B\), as defined by
\[ \theta_B = \tan^{-1} \left( \frac{M_z}{M_x} \right) \]  

First, a new coordinate system, \((x', y', z')\) is defined at the twist–bend center, which is rotated around the \(z'\) axis by the angle \(\theta_B\) to make the bending always defined in the direction of the \(x'\) axis. Subsequently, the transformation for bending in \((x', y', z')\) coordinate system is given by Eq. (7) as shown in Fig. 1(b),

\[ T'_{\text{bend}}(x', y', z') = \begin{pmatrix} r_M - (r_M - x') \cos \left( \frac{z'}{r_M} \right) \\ y' \\ (r_M - x') \sin \left( \frac{z'}{r_M} \right) \end{pmatrix}, \]  

where curvature \(1/r_M = (\pi/4L)|M|\) and \(|M| = \sqrt{M_x^2 + M_y^2 + M_z^2}\).

Finally, the transformation, \(T'_{\text{bend}}(x', y', z')\) is rotated by \(-\theta_B\) and translated by \((-X_c, -Y_c, -Z_c)\) to define \(T_{\text{bend}}(x, y, z)\) in the original coordinate system.

### II.A.3. Registration process

The registration process finds the best transformation, \(T\), that maximizes the normalized mutual information (NMI) between the target image \(I_t\) and the transformed source image \(I'_s\). The corresponding values for all the registration parameters, three rotations \((\theta_x, \theta_y, \theta_z)\), three translations \((t_x, t_y, t_z)\), twisting torque \(\langle Q \rangle\), bending moments \((M_x, M_y, M_z)\), and twist–bend center \((X_c, Y_c, Z_c)\) are required to be found using an optimization process. The NMI \(H(I_t, I'_s)\) is defined using image entropies as given in Eq. (8), which is less sensitive to changes in image overlap.

\[ \text{NMI}(I_t, I'_s) = \frac{H(I_t) + H(I'_s)}{H(I_t, I'_s)}, \]  

where \(H(I_t)\) and \(H(I'_s)\) denote marginal entropies of corresponding images, and \(H(I_t, I'_s)\) defines their joint entropy. The calculations were based on the method of Mattes et al.,44,45 where the probability density distribution is estimated using Parzen histograms. This method was largely insensitive to the size of the histogram.46 We set the histogram to 64 bins for the experiments in this paper. The Powell optimizer46 was used for numerical optimization to find the best set of registration parameters that maximize the NMI. Linear interpolation was employed in registration when transforming the source image, as it provided a sufficiently smooth optimization within a reasonable computational time.47 The registration framework in the Insight Toolkit48 was used for implementing the authors’ algorithm.

### II.A.4. Preprocessing

A median filter with a kernel size of \(5 \times 5 \times 5\) voxels on the 3D US carotid images was used to reduce the effect of speckle as a preprocessing step. Speckle is an integral part of US images and conveys useful information to the radiologist, but causes problems in registration.49 In addition, the interference of other anatomical structures in carotid 3D US images may cause errors in registration. Therefore, the region outside of the carotid artery and its vicinity was masked out, keeping only a region about half a diameter of the common carotid from the outer carotid boundary. To define a conical region of interest (ROI), two circles were defined at the top and the bottom planes perpendicular to the \(z\) axis, enclosing the carotid artery (or two arterial sections on the top plane) and the surrounding area up to a length approximately equal to a half a diameter of the common carotid. To define the circles, the user chose two points on the perimeter of each circle at two ends of an imaginary line passing through the center. Then, the cone-shaped ROI was defined using those two circles as cross sections, and the volume outside was masked out as shown in Fig. 2. MR images contain carotid arteries of both sides of the neck together with other anatomical structures. Therefore, the authors used the ROI for registration to select only the corresponding carotid artery (left and right separately) and its surrounding area using the same procedure as for the 3D US images, as shown in Fig. 2(c).

### II.A.5. Registration initialization

To initialize the registration, the user first selected the carotid bifurcation in both target and source images. Then the source image was approximately aligned with the target image by aligning the user-selected bifurcations (BFs) [as shown in Fig. 3(a)] by introducing an initial translation. It has been shown that the carotid bifurcation can be located in 3D US images with a standard deviation of 0.56 mm.27 In addition, the initial rotation angles \(\alpha, \beta, \gamma\) were calculated using the BF points and another three user-selected points: center of the common carotid, \(c\), at the bottom and centers through internal, \(i\), and external, \(e\), carotid arteries at the transverse plane of the bifurcation as shown in Fig. 3(a). Since this initial coarse alignment was allowed to change during the registration, it was assumed that the variations in the choice of the initialization were relatively insignificant to the final registration results.

### II.B. Evaluation of accuracy

#### II.B.1. Image acquisition

Two 3D US images of the left and right carotids, and two MR images of the neck at field strengths, 1.5 and 3.0 T, were acquired from each subject of a patient population with carotid total plaque area >0.5 cm², who were recruited from the Premature Atherosclerosis Clinic and The Stroke Prevention Clinic at University Hospital, London Health Sciences Center, London, Canada. Images were selected from six subjects for a total of 12 separate 3D US images of left and right carotids and corresponding 1.5 and 3.0 T MR images to evaluate the proposed nonrigid registration algorithm. Two MR field strengths were used to study the effect of signal-to-noise (SNR) and contrast-to-noise (CNR) ratios at different field strengths on US–MR nonrigid registration accuracy.

All subjects were scanned at 1.5 and 3.0 T with GE EXCITE whole-body MR systems, with identical pulse sequences and identical custom-built six-element carotid-
bifurcation-optimized receive-only phased-array coils. Both receive-only coils were geometrically and electronically identical, but with one tuned for use at 1.5 T and the other for use at 3.0 T. For both scanners, a specialized foam head holder was also used to reduce positioning variability and to provide comfortable but secure immobilization of each subject’s head. The time frame between MRI scanning sessions at each field strength was approximately 2 h. Multiple MR sequences were acquired and the total acquisition time for all sequences was approximately 60 min. At both 3.0 and 1.5 T, a set of 100, 2-mm-thick axial 2D time-of-flight slices were acquired in 4 min to identify the superior–inferior level of the flow dividers on each side. A localized shim was performed prior to each of the remaining sequences to ensure good performance of fat saturation pulses. The remaining series consisted of a multislice 2D black-blood acquisition, with T1-weighted (T1w) contrast, using double inversion recovery (DIR with 180° pulse). Parameters for the T1w scans analyzed in this study and acquired at 1.5 and 3.0 T are provided in Table I. All images following the localizer were acquired with 0.5 mm in-plane resolution, 2 mm slice thickness, and fat saturation. Subject motion was monitored by examining odd-to-even slice misregistration, and image data sets with substantial motion were not used in the study. All postlocalizer series were planned with slices centered in the superior–inferior direction at a point 5 mm superior to the flow divider of the index bifurcation. The T1w sequences were acquired with 16 contiguous 2-mm-thick slices covering 32 mm in the head-foot axis. In this paper, T1w scans were chosen for analysis because of previously published analysis that showed that blood flow suppression was best achieved using the single slice DIR technique widely used for MR carotid plaque imaging.50,51

Carotid US scans were performed at the Imaging Research Laboratories, Robarts Research Institute, using an HDI 5000 US machine (Philips/ATL). An L12-5 probe with a central frequency of 8.5 MHz was used in the composite imaging (SonoCT) mode. The transducer was attached to a motorized linear mechanical mover, oriented and translated along the neck for about 4 cm (in ≈ 8 s) approximately centered on the bifurcation while capturing 2D US slices at 30 slices/s to produce 3D US images with voxel sizes ≈0.1 × 0.1 × 0.15 mm using a 3D imaging system developed in the authors laboratory.11,28

Fig. 2. Images showing preprocessing of 3D MR and US images for registration. The 3D MR and US images have been sliced to show longitudinal views of the branches of the carotid artery: (a) MR, (b) US, and (c) and (d) corresponding views after masking. The user-selected bifurcation is also shown.
responds to the adventitia-tissue boundary. Therefore, theible in MR carotid images and the segmented boundary cor-

However, the media-adventitia boundary is not clearly vis-

multimodality 3D US and MR carotid image registration.

The authors previously used the carotid artery media-
adventitia boundary to calculate the registration error for in-

tramodality 3D carotid US image registration. Thus, a simi-

lar approach was used to calculate the registration error in

multimodality 3D US and MR carotid image registration.

Although the algorithm used 3D US and MR im-

ages of the whole vessel for registration, the carotid artery

surface was examined in three sections (internal, external,

and common carotid) to estimate the registration error in

each branch. In each carotid surface section, \( k \) \((k = 1: \text{common}, 2: \text{internal}, 3: \text{external})\) the registration error was calculated as the distance, \( d_{i,j,k} \) between corresponding points, \( (x_{i,j,k}, y_{i,j,k}, z_{i,j,k}) \) on the target and \( (x'_{i,j,k}, y'_{i,j,k}, z'_{i,j,k}) \) on the registered surfaces in each plane, \( j \) \((j = 1, 2, \ldots, N_k)\) normal to the \( z \) axis at every 1 mm interval as shown in Fig. 4 and defined in the following:

\[
d_{i,j,k} = \sqrt{(x'_{i,j,k} - x_{i,j,k})^2 + (y'_{i,j,k} - y_{i,j,k})^2 + (z'_{i,j,k} - z_{i,j,k})^2},
\]

The corresponding points were defined by projecting radial lines, \( i \) \((i = 1, 2, \ldots, 90)\), from the center of each contour to the corresponding surface at the same angle at \( 4^\circ \) intervals, as described in the evaluation of nonrigid registration of 3D

![Fig. 3. Registration initialization: (a) User-defined points, bifurcation points (BF), center of the common carotid, \( c \), at the bottom and centers of internal, \( i \), and external, \( e \), carotid arteries at the transverse plane through the bifurcation, and (b) initial rotation angles \( (\alpha_x, \alpha_y, \alpha_z) \) on the three orthogonal planes.](image)

![Fig. 4. Registration error calculation using the distance between corresponding points, on the target \((x'_{i,j,k}, y'_{i,j,k}, z'_{i,j,k})\) and registered \((x'_{i,j,k}, y'_{i,j,k}, z'_{i,j,k})\) surfaces at the direction, \( i \), on each plane, \( j \), normal to \( z \) axes at every 1 mm in carotid section, \( k \).](image)

| Table 1. Magnetic resonance imaging parameters for T1-weighted double inversion recovery scans at 1.5 and 3.0 T. |
|----------------|----------------|
| Acquisition parameter | 1.5 T | 3.0 T |
| Echo time, TE (ms) | 12 | 11.4 |
| Recovery time, TR | 1RR | 1RR |
| Receiver bandwidth, RBW (kHz) | 41.67 | 41.67 |
| Field of view, FOV (cm) | 11 | 11 |
| Thickness (mm) | 2 | 2 |
| Matrix | 224×224 | 224×224 |
| Number of excitations, NEX | 3 | 3 |
| Scan time (min) | 8:48 | 8:48 |
| Fat saturation | Yes | Yes |
| Spacing overlap | 0 | 0 |
| Number of slices | 16 | 16 |
| Pulse sequence | FSE | FSE |

images. Wong et al.\(^{53}\) reported that the average adventitia thickness in US images is 0.5 ± 0.1 mm, and thus the authors expanded the segmented media-adventitia boundary in US images by 0.5 mm before the distance-based registration error calculation described in the following.
US–US carotid images. The resulting 3D distributions of registration error values were also used for better visualizing and analysis of the registration accuracy by mapping the color-coded distance $d_{i,j,k}$ onto the segmented carotid surfaces.

The mean registration error (MRE) was defined for each carotid section, $k$, for individual image sets as described in the following:

$$\text{MRE}_k = \frac{1}{90N_k} \sum_{j=1}^{N_k} \sum_{i=1}^{90} d_{i,j,k}.$$  \hspace{1cm} (10)

The mean registration error (MRE) was also calculated for the whole vessel, to express the registration error by a single numerical value as given in the following:

$$\text{MRE} = \frac{1}{90\sum_{k=1}^{3} N_k} \left( \sum_{k=1}^{3} \sum_{j=1}^{N_k} \sum_{i=1}^{90} d_{i,j,k} \right).$$  \hspace{1cm} (11)

The maximum registration errors (MAXE) were also calculated for each carotid section, $k$, for individual image sets as described in the following:

$$\text{MAXE}_k = \max_{j \in [1,N_j]} \left( \max_{i \in [1,90]} d_{i,j,k} \right).$$  \hspace{1cm} (12)

The registration error $d_{i,j,k}$ between the segmented vessel surfaces in the target image and the registered image includes the error due to the variability in surface segmentations as well. To reduce the effect of the segmentation errors in registration error calculations, the mean surface of five repeated segmentations of the expert user was used, with at least 5 days between the segmentation operations as was done in the US–US registration error calculations.

II.B.3. Nonrigid registration accuracy

MR images were registered at field strengths of 1.5 and 3.0 T separately with corresponding 3D US images using the authors’ twisting and bending model-based nonrigid registration algorithm. The MRE was calculated separately for each carotid section, $k$, as well as the MRE for the whole vessel (all three sections together) in all the samples. The authors also calculated the average MRE and MRE and the standard deviation of all 12 carotid vessels using the values of MRE and MRE of individual samples.

II.B.4. Comparison of rigid and nonrigid registration results

The registration error calculations as described earlier in Sec. II B 3 for the rigid registration were repeated and the errors were compared to those of nonrigid registration to analyze the capabilities of the nonrigid registration method in improving the registration accuracy. The significance of the improvement was analyzed statistically using paired $t$-tests separately for each carotid section, as well as for the whole vessel together. In addition, the MAXE was calculated separately for each carotid section, $k$, for all rigid and nonrigid registrations to examine the ability of nonrigid registration in correcting large localized errors of rigid registration.

II.B.5. Registration error as a function of the distance from carotid bifurcation

Distribution of registration error along the carotid artery is an important factor in evaluating the effectiveness of the registration results. For a example, since most of the carotid plaque occurs around the bifurcation, a low registration error close to the bifurcation is preferable. To study this, the authors calculated the average registration error on each plane perpendicular to the $z$ axis as a function of the distance, $s$, from the carotid bifurcation to that plane, in each carotid section, $k$, for both field strengths as described in Eq. (13). The distance $s$ was defined by the distance along the $z$ axis from the user-defined carotid bifurcation.

$$\text{MRE}_{s,k} = \frac{1}{90} \sum_{i=1}^{90} d_{i,s,k}.$$  \hspace{1cm} (13)

II.B.6. Comparison of US–MR registration errors at two MR field strengths

Recent studies have shown that SNR and CNR ratios were higher in carotid MR images acquired at 3.0 T as compared to 1.5 T. The authors studied the effect of these changes on US–MR carotid nonrigid registration by comparing the registration errors at the two field strengths using a paired $t$-test at each carotid section, $k$, as well as all three carotid branches together for all image samples.

III. RESULTS

The tests were performed on a personal computer running Windows 2000 OS with a Pentium 4, 2.4 GHz processor, and 1 gbytes of RAM. The registration program was implemented with ITK without optimizing it for processing time. The registration time after initialization ranged from 10 to 40 min for one pair of images for those reported in this paper.

III.A. Nonrigid registration accuracy

The twisting and bending model-based nonrigid registration algorithm was applied to register each of 12 US images with corresponding 1.5 and 3.0 T MR images separately. Figure 5 shows a sample result, where Figs. 5(a) and 5(b) show the segmented vessel boundaries for 1.5 and 3.0 T, respectively, while Figs. 5(c) and 5(d) show the corresponding distance-based registration errors calculated and mapped onto the carotid surfaces on MR images. Table II shows the registration errors of twisting and bending model-based, intersubject, nonrigid registrations of US with 1.5 and 3.0 T MR separately for 3D images of carotid artery for all data sets, as well as the average values for all the samples. It is important to note that the internal and common carotid arteries showed a lower mean registration error than the corre-
FIG. 5. Examples of registration results showing vessel boundaries after twisting and bending model-based nonrigid registration using: (a) 3D US with 1.5 T MR, (b) 3D US with 3.0 T MR. The color-coded registration errors, $d_{ij}$ (mm), have been mapped on the registered MR surfaces using: (c) 3D US with 1.5 T MR and (d) 3D US with 3.0 T MR.

TABLE II. Distance-based mean registration error (MRE$_k$) for twisting and bending model-based nonrigid registration of US with 1.5 T MR and US with 3.0 T MR for the $k=1$: common, 2: internal, and 3: external carotid artery segments, as well as MRE for the whole vessel. Note that the 3D images of the whole vessel here used in the registration procedure. The last row shows the overall mean (and standard deviation in parentheses) for all of the image sets.

<table>
<thead>
<tr>
<th></th>
<th>Mean registration error, MRE$_k$ (mm)</th>
<th>MRE (mm)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$k=1$: Common</td>
<td>$k=2$: Internal</td>
</tr>
<tr>
<td></td>
<td>1.5 T</td>
<td>3.0 T</td>
</tr>
<tr>
<td>Subject1-L</td>
<td>0.8</td>
<td>1.5</td>
</tr>
<tr>
<td>Subject1-R</td>
<td>0.9</td>
<td>1.0</td>
</tr>
<tr>
<td>Subject2-L</td>
<td>1.1</td>
<td>3.0</td>
</tr>
<tr>
<td>Subject2-R</td>
<td>1.0</td>
<td>0.9</td>
</tr>
<tr>
<td>Subject3-L</td>
<td>2.1</td>
<td>1.9</td>
</tr>
<tr>
<td>Subject3-R</td>
<td>2.5</td>
<td>2.8</td>
</tr>
<tr>
<td>Subject4-L</td>
<td>1.0</td>
<td>0.7</td>
</tr>
<tr>
<td>Subject4-R</td>
<td>2.1</td>
<td>2.1</td>
</tr>
<tr>
<td>Subject5-L</td>
<td>1.5</td>
<td>1.3</td>
</tr>
<tr>
<td>Subject5-R</td>
<td>1.2</td>
<td>0.7</td>
</tr>
<tr>
<td>Subject6-L</td>
<td>1.5</td>
<td>1.1</td>
</tr>
<tr>
<td>Subject6-R</td>
<td>0.8</td>
<td>1.3</td>
</tr>
<tr>
<td>Overall mean</td>
<td>1.4</td>
<td>1.5</td>
</tr>
<tr>
<td>(s.d.)</td>
<td>(0.6)</td>
<td>(0.8)</td>
</tr>
</tbody>
</table>
sponding external artery for both 1.5 and 3.0 T MR images. This is an important result, since carotid plaques are found in the common and internal carotid arteries.

**III.B. Comparison of rigid and nonrigid registration results**

Each of the 12 US images was registered with corresponding 1.5 and 3.0 T MR images using only rigid registration to analyze the improvement in registration accuracy for twisting and bending model-based nonrigid registration. Figure 6 shows the average MRE of all image samples for different carotid sections, and average MRE for the whole vessel. Results of paired t-tests ($α=0.05$, $1−β=0.8$, $P<0.05$) in Table III showed that the registration error decreased significantly for the whole vessel as well as in the internal carotid artery for nonrigid registration of US images with MR images of both field strengths, but the change was not significant in the external carotid artery. Registration error in the common carotid artery showed a small but significant improvement over rigid registration for 1.5 T MR, while the change in registration error is not significant for 3.0 T MR. The normal distribution of all MRE values was tested using the Kolmogorov–Smirnov normality test.

The maximum registration error, MAXE, was also calculated for each carotid section separately for rigid and nonrigid registration of US carotid images with 1.5 and 3.0 T MR images and results are shown in Table IV. These results indicated that twisting and bending model-based nonrigid registration was able to reduce the maximum registration error in 52 of the 72 cases tested.

**III.C. Registration error as a function of the distance from carotid bifurcation**

Figure 7 shows the average MRE, $s$, of planes perpendicular to the $z$ axis for rigid registration as well as twisting and bending model-based nonrigid registration of 3D US with 1.5 and 3.0 T MR images separately. For all cases, registration error increased on planes away from the bifurcation, and could be justified since the registration process was started by aligning the bifurcations of target and source images. Therefore, a high registration accuracy was observed around the carotid bifurcation. The other important finding was that the nonrigid registration errors were lower than the rigid registration errors for all cases except the external carotid artery in the 3.0 T MR images.

**III.D. Comparison of US–MR registration errors at two MR field strengths**

The effect of MR field strength on registration error was investigated by performing paired t-tests on the nonrigid registration errors of US carotid images with MR images at different field strengths of 1.5 and 3.0 T and the results are shown in Table V. The results obtained did not show statistically significant differences between nonrigid registration accuracy of 3D US images with 1.5 and 3.0 T MR images.

<table>
<thead>
<tr>
<th>Carotid section</th>
<th>Mean difference (mm)</th>
<th>$P$ value</th>
<th>95% CI (mm)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1.5 T</td>
<td>3.0 T</td>
<td>1.5 T</td>
</tr>
<tr>
<td>Common</td>
<td>0.1</td>
<td>0.2</td>
<td>0.0380</td>
</tr>
<tr>
<td>Internal</td>
<td>0.2</td>
<td>0.4</td>
<td>0.0097</td>
</tr>
<tr>
<td>External</td>
<td>0.2</td>
<td>−0.1</td>
<td>0.0995</td>
</tr>
<tr>
<td>Overall</td>
<td>0.2</td>
<td>0.2</td>
<td>0.0016</td>
</tr>
</tbody>
</table>

*Statistically significant (paired t-test, $α=0.05$, $1−β=0.8$, $P<0.05$) reduction in registration error.
Even though the SNR and CNR are higher for 3.0 T MR images, when other parameters are kept at the same values, the authors did not observe any significant changes in non-rigid registration error after registration with 3D US images.

### IV. DISCUSSION

Atherosclerotic plaques that are vulnerable to disruption, fracture, or fissuring, are at higher risk for embolization, occlusion, and consequent ischemic strokes. Noninvasive imaging tools that better characterize arterial wall, atherosclerotic plaque structure and composition, and tissue strain and wall shear stress may help to determine the factors that lead to the development of unstable lesions, and identify patients at risk of plaque disruption and stroke. Combining the complementary information provided by 3D MR and US carotid images may lead to a better understanding of the underlying compositional and textural factors that define plaque and wall vulnerability, which may lead to better and more effective stroke prevention strategies and patient management. In addition, it may be advantageous to combine noninvasive MR plaque composition information with 3D US images, since it would provide a tool for in vivo validation of US plaque characterization. Although carotid MR imaging allows characterization of plaque composition, it is more expensive and less accessible for use as a monitoring tool, compared to ultrasound. Therefore, development...

### TABLE IV. Maximum registration error (MAXE<sub>k</sub>) of different carotid sections for both rigid (R) and twisting and bending model-based nonrigid (NR) registration and MR field strengths of 1.5 and 3.0 T.

<table>
<thead>
<tr>
<th>Carotid section</th>
<th>1.5 T</th>
<th>3.0 T</th>
</tr>
</thead>
<tbody>
<tr>
<td>Subject1-L</td>
<td>4.8</td>
<td>5.2</td>
</tr>
<tr>
<td>Subject1-R</td>
<td>5.6</td>
<td>4.0</td>
</tr>
<tr>
<td>Subject2-L</td>
<td>3.5</td>
<td>6.5</td>
</tr>
<tr>
<td>Subject2-R</td>
<td>2.7</td>
<td>2.5</td>
</tr>
<tr>
<td>Subject3-L</td>
<td>5.1</td>
<td>5.3</td>
</tr>
<tr>
<td>Subject3-R</td>
<td>4.5</td>
<td>5.6</td>
</tr>
<tr>
<td>Subject4-L</td>
<td>2.8</td>
<td>2.1</td>
</tr>
<tr>
<td>Subject4-R</td>
<td>4.6</td>
<td>4.0</td>
</tr>
<tr>
<td>Subject5-L</td>
<td>4.0</td>
<td>3.5</td>
</tr>
<tr>
<td>Subject5-R</td>
<td>2.9</td>
<td>3.5</td>
</tr>
<tr>
<td>Subject6-L</td>
<td>6.1</td>
<td>4.3</td>
</tr>
<tr>
<td>Subject6-R</td>
<td>3.1</td>
<td>5.0</td>
</tr>
</tbody>
</table>

### TABLE V. Comparison of the difference in MRE<sub>k</sub> (mm) and MRE for US–MR twisting and bending mode-based nonrigid registration results with MR images of field strengths 1.5 and 3.0 T. There were no statistically significant paired t-test, \( \alpha=0.05, 1-\beta=0.8, P<0.05 \) changes in registration error.

<table>
<thead>
<tr>
<th>Carotid section</th>
<th>Mean of MRE difference (mm)</th>
<th>( P ) value</th>
<th>95% confidence interval (mm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Common</td>
<td>−0.1</td>
<td>0.5636</td>
<td>−0.6–0.3</td>
</tr>
<tr>
<td>Internal</td>
<td>0.2</td>
<td>0.3352</td>
<td>−0.3–0.7</td>
</tr>
<tr>
<td>External</td>
<td>−0.2</td>
<td>0.6511</td>
<td>−1.0–0.6</td>
</tr>
<tr>
<td>Overall</td>
<td>−0.04</td>
<td>0.7764</td>
<td>−0.3–0.3</td>
</tr>
</tbody>
</table>
of noninvasive, less-expensive US imaging tools may be advantageous, since US is already used for measuring carotid stenosis and IMT in clinical settings.

This paper presented an intrasubject nonrigid registration technique to register 3D carotid US and MR images to correct the nonlinear misalignments between them, which is essential for combining the two imaging modalities. A twisting and bending model was used to deform images within the natural movements of the neck. Thus, any unnatural warping of the images during the registration could be avoided. As an added benefit, the number of nonrigid transformation parameters could be limited to six, which is a low number compared to other nonrigid transformations, such as the free form deformation. The twisting and bending model can be changed if a need arises in the future to make it more comprehensive. For example, if the authors change the twisting model from the “cylindrical shaft” to an “elliptical shaft,” warping can be introduced to some degree in the z axis, but then the model will have one additional registration parameter. More testing, using either a phantom or different head positions in MR and US image acquisitions, is then necessary to determine if a more complex model would reduce the registration error significantly. The nonrigid registration algorithm presented was tested with 3D US and MR images acquired from subjects with total plaque area >0.5 cm². The effects of cardiac pulsation in carotid images, if present, can be a problem in image registration. However, this problem has a very small effect, if any, in patients with carotid atherosclerosis, since the stiffness of the carotid artery increases with progression of the disease, thereby reducing the appearance of cardiac pulsation effects in images.

Since the ground truth was unknown, evaluation of the registration accuracy was challenging. It was very difficult to find a sufficient number of corresponding landmarks in 3D carotid US and MR images to calculate the target registration error. The authors previously used segmented media-adventitia boundary surfaces for evaluating 3D US to US carotid image registration. However, the media-adventitia boundary was not clearly visible in MR images, and therefore the segmented surfaces in US images was expanded by an average adventitia thickness of 0.5 mm (Ref. 53) to calculate the registration errors between adventitia-tissue boundary surfaces. In addition, registration error calculations using segmented vessel boundaries also contain uncertainty due to variability in segmentation. Mean standard error in segmentation has been reported to be about 0.2 mm for similar 3D US images of the carotid arteries.

3D US images with MR images were registered at two field strengths. As shown in Table II, nonrigid registration errors, at both field strengths, were lower in the common and internal carotid arteries than in the external. Similarly, as shown in Fig. 6, nonrigid registration errors were significantly lower than the rigid registration errors in the internal carotid artery, while the change was not statistically significant in the external carotid artery. This was an important observation since the internal carotid artery is more important in managing patients at risk of stroke as it is in the direct path of blood supply to the brain. The diameter of the external carotid is small and sometimes difficult to image with 3D US, making it difficult to register, and optimization process could also be biased toward optimizing the alignment of two larger branches during the registration process. In addition, the external carotid artery is closer to the skin and, therefore, can be moved with respect to the internal carotid in US image acquisition due to the pressure from the probe, and this movement cannot be corrected by the twisting and bending model. Thus, the registration of the external carotid artery was not consistent and therefore improvements were not significant. Since the external carotid artery mainly supplies blood to the external organs in the face, this would not present a significant problem in monitoring carotid plaque. Furthermore, comparing the twisting and bending model-based nonrigid registration technique with other unconstrained nonrigid registration algorithms may be useful to analyze the ability of the technique. However, use of unconstrained nonrigid transformations in registration of 3D US and MR images of the carotid arteries must be avoided to prevent the warping of different plaque components visible in the two modalities.

While nonrigid registration errors in the common carotid artery showed a small but significant improvement over rigid registration for 1.5 T MR, the change in registration error was not significant for 3.0 T MR. The authors did not expect a large difference between rigid and nonrigid registration errors in the common carotid artery as it was anticipated that the most significant nonrigid deformations would be present in the upper part of the carotid arteries since the different head positions deform the upper part of the neck more than the base where the common carotid artery is located. In addition, as shown in Fig. 7, nonrigid registration errors in the internal and common carotid arteries were low near the bifurcation, which is the area usually monitored for carotid plaque. Furthermore, vulnerable plaques may show larger localized changes, mainly ulcerations and fissuring, which are large enough to be detected with registration errors reported in this paper. For an example, Sitzer et al. obtained a mean ulceration diameter of 3770 ± 2130 µm (3.7 ± 2.1 mm) for a symptomatic patient population.

Although 3.0 T MR images have a higher SNR and CNR than the 1.5 T images, the authors did not observe statistically significant change in registration errors between them, as shown in Table V. An important factor may be due to the use of MI as the image similarity measure, which has been shown to be robust in the presence of noise and intensity value variations. It may also be beneficial to study the differences in segmentations in 1.5 and 3.0 T MR images after aligning them by rigid registration. However, a number of previous studies for other applications has not shown significant differences in physical measurements and diagnostic accuracy.

The nonrigid registration technique presented in this paper required some user interaction in preprocessing and initialization. In preprocessing, a user-selected ROI must be defined but the effect of variability in ROI selection would have a minimal effect on the final results due to the use of...
NMI as image similarity measure. NMI is insensitive to the changes of overlap of ROI defined in the two images.\textsuperscript{61} In addition, the user must select four points on each of the two images for initializing the registration. One point each in each image is used to identify the carotid bifurcation to be used to calculate both initial transformations and rotations and it has been shown that the carotid bifurcation can be located in 3D US images with a standard deviation of 0.56 mm.\textsuperscript{27} The other three points together with the bifurcation are used to calculate the initial rotation. Since this initial coarse alignment was allowed to change during the registration, it was assumed that the variability in point selections was relatively insignificant to the final registration results.

Processing time of 10–40 min for registration of one pair of images can be considered higher than normal for a registration technique with only 12 parameters (6 rigid and 6 nonrigid). The authors have not optimized the program yet and have used a user interface implemented in Python programming language. Therefore, speed may be improved by optimizing the program and implementing the whole program in C++. In addition, large 3D US images (about 400\( \times \)350\( \times \)750 voxels) could have contributed to the high processing time. Thus, the speed can be improved by downsampling the 3D US images, but more tests would be required to find a balance between processing time and registration accuracy.

Speckle in US images provides useful information to the radiologist, but together with shadowing and other artifacts, makes registration prone to errors. The authors minimized the effect of this problem by using MI and nongradient-based Powell optimizer. Tissue mimicking flow artifacts of MR images\textsuperscript{62} might create some problems with the registration process and more research is needed to find ways to overcome these limitations. Effects on the registration process by other anatomical structures in the carotid images were reduced by masking out the area outside the carotid artery and its surrounding. Although effects due to the presence of cardiac pulsation in carotid images were small for patients with stiff arterial wall,\textsuperscript{63} they may be further minimized by using cardiac gating image acquisitions.

V. CONCLUSION

A nonrigid registration technique to register intra-subject 3D US and MR carotid images has been developed using a twisting and bending model, normalized mutual information, and Powell optimization method. The technique was tested with 3D US and corresponding 1.5 and 3.0 T MR images acquired on the same day and overall errors of 1.4±0.3 mm for 1.5 T and 1.5±0.4 mm for 3.0 T were obtained. However, these values also include the uncertainties from expansion of media-adventitia boundary in the US images to establish the adventitia-tissue boundary and variability in segmentation. Therefore, the actual registration errors may be less than the reported values. More important, lower registration errors were obtained in the internal and common carotid arteries, especially close to the carotid bifurcation, where most of the plaque accumulates due to the arterial remodeling in the presence of atherosclerosis.

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\textsuperscript{19}T. S. Hatsukami, R. Ross, N. L. Polissar, and C. Yuan, “Visualization of fibrous cap thickness and rupture in human atherosclerotic carotid plaque in vivo with high-resolution magnetic resonance imaging.” Circulation