A Methodology to Detect Abnormal Relative Wall Shear Stress on the Full Surface of the Thoracic Aorta Using Four-Dimensional Flow MRI

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Purpose: To compute cohort-averaged wall shear stress (WSS) maps in the thoracic aorta of patients with aortic dilation or valvular stenosis and to detect abnormal regional WSS.

Methods: Systolic WSS vectors, estimated from four-dimensional flow MRI data, were calculated along the thoracic aorta lumen in 10 controls, 10 patients with dilated aortas, and 10 patients with aortic valve stenosis. Three-dimensional segmentations of each aorta were coregistered by group and used to create a cohort-specific aortic geometry. The WSS vectors of each subject were interpolated onto the corresponding cohort-specific geometry to create cohort-averaged WSS maps. A Wilcoxon rank sum test was used to generate aortic $P$-value maps ($P<0.05$) representing regional relative WSS differences between groups.

Results: Cohort-averaged systolic WSS maps and $P$-value maps were successfully created for all cohorts and comparisons. The dilation cohort showed significantly lower WSS on 7% of the ascending aorta surface, whereas the stenosis cohort showed significantly higher WSS on 34% of the ascending aorta surface.

Conclusions: The findings of this study demonstrated the feasibility of generating cohort-averaged WSS maps for the visualization and identification of regionally altered WSS in the presence of disease, compared with healthy controls. Magn Reson Med 00:000–000, 2014. © 2014 Wiley Periodicals, Inc.

Key words: wall shear stress; aorta; dilation; valve stenosis

INTRODUCTION

Clinical management of aortopathy employs anatomical and hemodynamic measurements, such as aortic diameter and peak transvalvular blood velocity. In the presence of aortic dilation or aortic valve disease, these parameters are simple to acquire, yet have limited predictive ability to identify subjects who may experience progressive aortic enlargement, dissection, or rupture (1–3). Current guidelines focus on consensus-based thresholds for aortic diameter (4, 5) and are known to lead to divergent outcomes in similarly classified patients in terms of symptom status, cardiovascular events, or aortic valve replacement (6–8). Furthermore, changes in aortic geometry are often detected late in the disease process, and management does not consider the underlying markers suspected to drive wall remodeling and progressive aortic dilatation. This is an important consideration given recent longitudinal studies finding independent associations between hemodynamic markers and aortopathy (9, 10). In this context, wall shear stress (WSS), the tangential force exerted by blood flow on the vessel wall (11), may be a promising prognostic marker associated with the expression of transcriptional factors responsible for extracellular matrix degradation and vascular smooth muscle cell apoptosis (12–14).

In the last decade, considerable progress has been made regarding the estimation of WSS. The development of four-dimensional (4D) flow MRI (time-resolved three-dimensional [3D] phase contrast MRI with three-directional velocity encoding) (15–17) with full volumetric coverage of the thoracic aorta has made the assessment of in vivo WSS a reality (18–27). Nonetheless, a number of challenges remain to determine whether abnormal WSS presents a definitive risk for adverse aortic remodeling in a large, well-controlled population-based study. The current challenges include the complexity of measuring WSS along the entire aortic surface, a lack of established baseline WSS values (for healthy individuals), spatial resolution effects, and the ability to consistently identify and describe aortic locations with abnormal WSS. As a consequence, there is currently no standardized method for comparing 3D aortic WSS between individuals and patient cohorts.

In this study, we present a novel methodology capable of quantifying 3D WSS over the entire aorta lumen surface and thereby provide a method to derive cohort-averaged 3D WSS vector maps. The technique allows for compact visualization of 3D WSS assessed across multiple subjects and enables a quantitative comparison of the 3D WSS environment on the surface of the vessel between cohort groups. The goal of this study was to present the methodology and test the hypothesis that the technique can statistically compare regional 3D WSS differences as...
Abbreviations: MAA, midascending aorta; NA, not applicable; SOV, sinus of Valsalva; VENC, velocity encoding.

Differences across cohorts were evaluated using the Kruskal-Wallis test.

expressed on the entire vessel surface between patient groups according to the nature of the aortic disease.

METHODS

Study Cohort

Thirty subjects were retrospectively enrolled according to the following subgroup criteria: 1) 10 age-appropriate healthy control subjects (59 ± 10 years) with no history of cardiovascular disease, normal aortic valve function, and normal thoracic aortic geometry; 2) 10 patients with normal aortic valve function and aortic dilation (59 ± 12 years); and 3) 10 patients determined to have moderate to severe aortic stenosis and concomitant aortic dilation (65 ± 14 years, referred to as the stenosis cohort). Patients were excluded if Marfan syndrome or Ehlers-Danlos syndrome was present. All subjects had trileaflet aortic valve morphology. Aortic dilation was defined as a sinus of Valsalva or midascending aorta diameter >4.0 cm. Aortic stenosis was graded according to the absolute systolic peak velocities at the level of the aortic valve, as is recommended for continuous wave Doppler ultrasound guidelines (moderate stenosis = 3–4 m/s; severe stenosis = >4 m/s) (4). Aortic insufficiency was graded as mild, moderate, or severe according to a regurgitant fraction of <30%, 30%–49%, or >50%, respectively (4).

The demographics of the study subjects are summarized in Table 1. The study was approved by the local Institutional Review Board. Nine controls, 2 patients with aortic dilation, and 2 patients with aortic valve stenosis provided informed consent. The tenth control presented normal findings on a clinical scan. The remaining subjects were enrolled using an Institutional Review Board–approved protocol permitting retrospective chart review.

MRI

4D flow MRI measurements were performed on 1.5T and 3T scanners (Espree, Avanto, Skyra, Aera, Siemens, Erlangen, Germany). All patients underwent a standard-of-care thoracic cardiovascular MRI including electrocardiography-gated time-resolved cardiac MRI for the evaluation of cardiac function and valve morphology as well as contrast-enhanced MR angiography for the quantification of aortic dimensions. For valve imaging, a two-dimensional (2D) imaging plane was carefully positioned orthogonal to the aortic root at the level of the aortic valve. In addition, 4D flow MRI of the thoracic aorta was performed in a sagittal oblique volume using prospective electrocardiography gating and free-breathing with a respiratory navigator placed on the lung–liver interface (28). 4D flow pulse sequence parameters were as follows: spatial resolution = 1.7–3.6 × 1.8–2.4 × 2.2–3.0 mm³; temporal resolution = 37–42 ms (14–25 cardiac time frames); echo time/repetition time = 2.2–2.8/4.6–5.3 ms; flip angle = 7°–15°; field of view = 144–380 × 120–285 × 67–116 mm³; velocity sensitivity = 150 cm/s, 150–250 cm/s, 150–450 cm/s for the healthy controls, patients with aortic dilation, and patients with aortic stenosis, respectively (manually specified by the technologist to minimize velocity aliasing).

Data Analysis

4D flow MRI measurements were corrected for eddy currents, Maxwell terms, and velocity aliasing using custom-built software programmed in MATLAB (Mathworks, Natick, Massachusetts, USA) as described previously (29). Voxels with persistent velocity aliasing were corrected manually. 3D phase contrast MR angiography data were created by voxel-wise multiplication of the magnitude data with absolute velocities averaged over all cardiac time frames (29). The 3D phase contrast MR angiography images were semiautomatically segmented using a commercial software package (MIMICS, Materialise, Leuven, Belgium). To obtain a smooth surface of the aortic wall, the segmentation was smoothed using a Laplacian filter (30). Peak systole was defined as the cardiac time frame with the highest average aortic velocity.
Cohort-Specific Aortic Geometry and WSS Maps

The data analysis workflow for 3D WSS estimation, calculation of cohort-specific aortic geometry and WSS maps, and quantification of intercohort differences is shown in Figure 1. All analyses were performed in MATLAB. Visualization of the results was partially performed using commercial software (Ensight, CEI Inc, Apex, North Carolina, USA).

3D WSS Estimation for Individual Aortas

As summarized in Figure 2, WSS vectors were calculated by:

$$\tau = 2 \eta \dot{e} \cdot \tilde{n}$$  \[1\]

where $\tau$ is the WSS vector, $\eta$ is the dynamic viscosity (Newtonian: $3.2 \cdot 10^{-3}$ Pa s), $\dot{e}$ is the rate of deformation.
tensor, and \( \vec{n} \) is the normal vector orthogonal to the vessel wall (Fig. 2a).

By rotating the coordinate system such that the z-axis aligns with the normal vector of the vessel wall (Fig. 2b), it holds that: \( \vec{n} = (0, 0, 1) \). Since no flow occurs through the wall \( (\vec{n} \cdot \vec{v} = 0 \text{ at the wall}) \), the inner product of the rate of deformation tensor and the normal vector is reduced to:

\[
2\dot{\varepsilon} \cdot \vec{n} = \begin{pmatrix}
\frac{\partial \nu_x'}{\partial z'} & \frac{\partial \nu_y'}{\partial z'} & 0
\end{pmatrix}
\]

where the shear rates \( \frac{\partial \nu_x'}{\partial z'} \) and \( \frac{\partial \nu_y'}{\partial z'} \) are the spatial velocity gradients at the wall in the rotated coordinate system. The rotated WSS vector \( \vec{t}' \) is then defined as:

\[
\vec{t}'_x = \eta \frac{\partial \nu_x'}{\partial z'}, \quad \vec{t}'_y = \eta \frac{\partial \nu_y'}{\partial z'}, \quad \vec{t}'_z = 0
\]

The shear rates were derived from one-dimensional smoothing splines (31) fitted through the rotated x- and y-velocity values along the inward normal vector (32) Fig. 2c, Fig. 2d. The WSS vector was transformed to the original coordinate system by inverse rotation Fig. 2e, Fig. 2f. WSS was averaged over five cardiac time frames centered at peak systole to reduce noise. Systole was considered as it is the most hemodynamically active portion of the cardiac cycle. A previous study showed that important differences in hemodynamic behavior were diluted when the diastolic period was included (33).

**Cohort-Averaged WSS map**

A cohort-specific aorta WSS map was created using the following three-step process: 1) the individual aorta 3D segmentations from each cohort were rigidly coregistered, and a map quantifying the amount of shared geometry was generated (i.e., an “overlap” map); 2) the maximal overlap was used to create a cohort-specific, idealized aorta geometry; and 3) the individual 3D WSS vectors were projected onto the cohort-specific aorta geometry and a 3D WSS map representative of the entire cohort was calculated.

**Rigid Coregistration and Generation of the Overlap Map.**

As shown in Figure 3 (top row), two aortic 3D segmentations were rigidly coregistered (six degrees of freedom) using FMRIB’s Linear Image Registration Tool (FLIRT) (34). To create the initial overlap map, two aortic segmentations were summed in a voxel-wise fashion, and volume containing either a 1 (no overlap) or 2 (both aortas overlapping) was created. Next, a third aorta was coregistered, and the overlap map was calculated to yield values ranging from 1 to 3 (1 = no overlap, 2 = two aortas overlapping, and 3 = three aortas overlapping [Fig. 3, lower row]). This process was continued for all 10 subjects in each cohort such that the final map included values representing the amount of geometry shared by each segment, ranging from an overlap value of 1 to 10. Two sequence orders for aorta registration were used to test for reproducibility of the overlap map creation. This process can be expressed as:

Consecutive registration:

\[
\sum_{n=1}^{10} A_n
\]

Random registration:

\[
\sum_{m}^{10} A_m \text{ with } m = \{5, 9, 6, 4, 2, 10, 1, 8, 3, 7\}
\]

**Identification of the Cohort-Specific, Idealized Aorta Geometry.** For each subject, the original aorta 3D
segmentation was coregistered to multiple potential cohort-specific aorta geometries as defined by overlap thresholds \(O_{\text{thresh}}\) with a range of \(1 \leq O_{\text{thresh}} < 10\). Each \(O_{\text{thresh}}\) defined a potential cohort-specific aorta geometry with at least \(n = O_{\text{thresh}}\) overlapping aorta regions \((O_{\text{thresh}}\) map\). Each aorta in the cohort was then rigidly coregistered to the threshold \(O_{\text{thresh}}\) map, and an individual registration error (RE, relative number of voxels not shared by the aorta and \(O_{\text{thresh}}\) map) was calculated as:

\[
RE = \frac{N_{O_{\text{thresh}}-Aorta}}{\left(N_{O_{\text{thresh}}} + N_{Aorta}\right)/2} \times 100 \tag{6}
\]

where \(N_{O_{\text{thresh}}-Aorta}\) is the number of voxels not shared between the overlap map and the individual aorta, \(N_{O_{\text{thresh}}}\) the number of voxels of the \(O_{\text{thresh}}\) map and \(N_{Aorta}\) the number of voxels of the individual aorta. Finally, the \(O_{\text{thresh}}\) map with the lowest RE averaged over all aortas in the cohort was chosen as the cohort-specific aorta geometry. Figure 4 illustrates the optimization process for examples with \(O_{\text{thresh}} \geq 4\) and \(O_{\text{thresh}} \geq 6\).

**Cohort-Specific 3D WSS Maps.** To project the 3D WSS vectors onto the final cohort-specific aortic geometry, affine registration (FLIRT, 12 degrees of freedom) was used, followed by nearest-neighbor interpolation of the 3D WSS vectors (Fig. 5). To investigate the influence of the interpolation process, each individual aorta and the cohort-specific aorta geometry were separated into three regions: ascending aorta (AAo), aortic arch (Arch), and descending aorta (DAo), as shown in Figure 5. The interpolation error (IE) was defined as the relative difference between the mean WSS of the cohort-averaged aorta geometry and the individual aorta:

\[
IE = \frac{|\text{mean WSS geometry} - \text{mean WSS aorta}|}{(\text{mean WSS geometry} + \text{mean WSS aorta})/2} \times 100 \tag{7}
\]

**Averaging Over Cohort.** Finally, cohort-averaged 3D WSS vector maps as well as standard deviation (SD) maps reflecting interindividual differences in WSS were calculated for each of the three cohorts.
Analysis of WSS Differences Between Cohorts

To enable comparison between cohorts, the 3D WSS vectors of the dilation and stenosis cohort were interpolated (nearest neighbor interpolation [see Figure 5]) to the aorta geometry of the control cohort. For this process, the registration error and interpolation error were calculated.

To test the dependence of the comparison between cohorts on the choice of aorta geometry for comparison between cohorts, the subjects in the control cohort were registered and interpolated to the aorta geometry of the dilation cohort.

Statistical Analysis

A Kruskal-Wallis test was used to evaluate differences in age, aortic diameter, and interpolation error between cohorts.

A Wilcoxon rank sum test was used to assess whether differences existed due to the sequence of the aorta registration for the overlap map creation. To identify regional differences in WSS between two cohorts, a Wilcoxon rank sum test was performed for each location on the control aorta geometry, resulting in a P-value map. Differences were considered statistically significant if $P < 0.05$. The resulting P-values were mapped onto the aorta geometry of the control cohort to create aorta P-value maps in order to visualize significant regional differences of WSS between cohorts. To test for reproducibility, a P-value map was created of the individual controls registered and interpolated to the aortic geometry of the dilations.

A Wilcoxon rank sum test was performed to investigate differences between the sinus of Valsalva and mid-ascending aorta diameter of the dilation and stenosis cohort.
RESULTS

3D WSS Estimation for Individual Aortas

In Figure 6, examples of measured systolic blood flow velocities and derived 3D WSS maps for an individual aorta from each cohort are shown. The velocities along the outer curvature of the ascending aorta of the patient with valve stenosis (Fig. 6c) were higher compared with both the control (Fig. 6a) and the dilated aorta (Fig. 6b). As a result, differences in regional velocity profiles (e.g., in the distal ascending aorta in Figure 6d–6f) resulted in altered velocity gradients at the wall in patients with aortic valve stenosis. This elevated velocity gradient resulted in regionally increased WSS for the aorta with valve stenosis (Fig. 6i) compared with the control (Fig. 6g) and dilated aorta (Fig. 6h).

Cohort-Averaged WSS Map

**Cohort-Specific Aorta Geometry**

For all three cohorts, aorta geometries were successfully created and an overlap threshold, $O_{\text{thresh}} \geq 4$ (i.e., an overlap of four or more aorta regions), showed a minimum RE of $23\% \pm 3.0\%$, $20\% \pm 4.7\%$, and $23\% \pm 9.3\%$ for the control, dilation, and stenosis cohorts, respectively. For the creation of the cohort-specific aorta geometries, a consecutive registration sequence starting at 1 to 10 aortas was used. When the random sequence was
applied, a minimum RE of $23\% \pm 4.0\%$, $20\% \pm 4.8\%$, and $23\% \pm 8.9\%$ was seen for the control, dilation, and stenosis cohorts, respectively. The differences between the consecutive and random registration errors were not significant for all cohorts ($P = 0.70$, $P = 0.94$, and $P = 0.91$ for the control, dilation and stenosis cohorts, respectively).

**Cohort-Specific 3D WSS Maps**

The mean interpolation error in the control cohort was $3.2\% \pm 3.0\%$, $2.2\% \pm 1.3\%$, and $4.7\% \pm 6.4\%$ for the AAo, Arch, and DAo, respectively. For patients with aortic dilation, the IE was $1.8\% \pm 1.2\%$, $2.9\% \pm 1.8\%$, and $4.7\% \pm 4.2\%$ for the AAo, Arch, and DAo, respectively. For the stenosis cohort, the IE was $4.1\% \pm 2.8\%$, $3.6\% \pm 4.2\%$, and $1.4\% \pm 0.8\%$. The difference in IE between cohorts was not statistically significant (Kruskal-Wallis test).

Figures 7a–7c display a right-anterior oblique and posterior view of the cohort-averaged 3D WSS maps for healthy controls, patients with dilated aortas, and aortic valve stenosis. The SD maps show that WSS values varied substantially between subjects in the AAo of the stenosis cohort, compared with the SD maps of the controls and patients with aortic dilation, which showed smaller intersubject WSS variability.

**Analysis of WSS Differences Between Cohorts**

The cumulative results of the intergroup comparison of aortic WSS are summarized in Figure 8 and Table 2. P-values for the intergroup comparisons in Figure 8a show that WSS in patients with aortic dilation was significantly reduced in the distal outer curvature (arrow 1) and proximal inner curvature (arrow 2) of the ascending aorta compared with controls (significantly lower WSS in 7% of all AAo voxels [see Table 2]). For registration and interpolation to the dilation geometry, the location and extent of regions with significant differences in WSS between cohorts were similar when compared with interpolation on the control geometry (see Figure 8a and 8b).

In contrast, WSS was significantly elevated in almost the entire outer AAo curvature and a fraction of the inner AAo (34% of AAo voxels [see Table 2]) for patients with aortic stenosis compared with controls (Fig. 8c, arrow 3). In similar regions, WSS was significantly elevated for patients with aortic stenosis compared with the aortic dilatation subjects (41% and 20% of AAo and arch voxels, respectively [Fig. 8d and Table 2]).

The registration errors for the affine registration to the aorta geometry of the control cohort were similar in scale to the registration errors for the cohort-specific idealized geometry: $18\% \pm 1\%$ for controls, $19\% \pm 5\%$ for dilations, and $22\% \pm 4.7\%$ for the stenosis cohort. The interpolation errors for the interpolation of the dilation cohort to the aorta geometry of the control cohort were similar to the interpolation error to the cohort-specific dilation geometry: $2.7\% \pm 2.6\%$, $2.4\% \pm 1.6\%$, and $4.0\% \pm 4.9\%$ for the AAo, Arch, and DAo, respectively. For the stenosis cohort, the interpolation errors were similar for the AAo and Arch ($4.2\% \pm 2.0\%$, $3.1\% \pm 3.0\%$) but higher for the DAo ($8.4\% \pm 6.8\%$).

**DISCUSSION**

The use of cohort-averaged 3D WSS maps derived from healthy or patient cohorts has the potential to serve as a means of comparing individual patient measurements with reference norms, or to compare measurements at specific anatomic locations between groups of subjects. This methodology is an improvement over previous methods which have used 4D flow-derived WSS at regions limited to manually positioned 2D analysis planes (18–22, 24–27, 36). In contrast to the single slice approach, the strategy presented here creates a comprehensive cohort-averaged 3D WSS map covering the thoracic aorta, which allows for the visualization of regional WSS variations between healthy and disease cohorts. The methods described here can be modified to function with maps of other biomarkers, such as regional diameter, oscillatory shear index, velocity vector magnitude/direction, helicity/vorticity (37), or blood residence times. Furthermore, the comparison of single-subject measurements with the accompanying cohort-averaged maps is possible. The possibility exists to use the method to form a type of “aortic atlas,” allowing for the

![FIG. 7. Right-anterior oblique and posterior view of the cohort-specific 3D WSS map for healthy controls (a), dilated aortas (b), and aortas with valve stenosis (c). Insets show the SD maps. The average regional WSS direction on the inner curvature of the AAo is shown by white arrows. The black arrow in (c) indicates elevated WSS at the outer curvature of the AAo. A, anterior; H, head; R, right; P, posterior.](image-url)
determination of whether (and where) a single subject expresses an abnormal biomarker, as defined by confidence intervals created from a large population control group.

The finding of significantly lower WSS in the dilated AAo group compared with the healthy control group is in good agreement with previously published results (23, 25). In contrast, patients with aortic valve stenosis exhibited significantly elevated WSS in the AAo group compared with healthy controls. Previous studies have speculated that this may occur (38), and this has been shown in cases with bicuspid aortic valve (26), but to our knowledge, this is the first study to demonstrate the extent and regional involvement of elevated WSS in patients with tricuspid stenotic valves. By means of the $P$-value maps, the location of significantly altered WSS for the disease groups is easily visualized (Fig. 8). It is known that WSS estimates can vary substantially between studies due to methodological choices, including variables such as viscosity, spatial resolution, and velocity fitting techniques (39). Therefore, WSS differences between cohorts are emphasized by the use of $P$-value maps, rather than absolute WSS values. Noticeably, WSS values for 40% of the ascending aorta surface in patients with aortic dilation were significantly different compared with subjects with aortic valve stenosis. This was despite similar sinus of Valsalva (4.1 $\pm$ 0.6 cm versus 4.0 $\pm$ 0.4 cm; $P = 0.38$ [Wilcoxon rank sum test]) and mid-ascending aorta (4.0 $\pm$ 0.4 cm versus 4.2 $\pm$ 0.3 cm; $P = 0.17$ [Wilcoxon rank sum test]) diameters for both cohorts. Furthermore, intercohort differences in WSS direction were readily apparent. The deviation in WSS direction was highest in the AAo for the stenosis cohort compared with healthy controls. These findings illustrate the complex nature of hemodynamic changes that are associated with aortopathy and that simple metrics such as aortic diameter do not directly correlate with the underlying physiologic changes in blood flow or WSS. In this context, the proposed concept of creating cohort-averaged maps has the potential to provide a better understanding of the role hemodynamic forces may play when considering endothelial cell dysfunction, and thus, potential risk for aortic remodeling (40).

The primary motivation for the development of the cohort-averaged maps is to better understand what

<table>
<thead>
<tr>
<th>Region</th>
<th>Dilations versus Controls</th>
<th>Stenosis versus Controls</th>
<th>Stenosis versus Dilation</th>
</tr>
</thead>
<tbody>
<tr>
<td>AAo (%)</td>
<td>2/7</td>
<td>34/2</td>
<td>41/2</td>
</tr>
<tr>
<td>Arch (%)</td>
<td>1/2</td>
<td>18/0</td>
<td>20/0</td>
</tr>
<tr>
<td>DAo (%)</td>
<td>0/0</td>
<td>3/0</td>
<td>4/0</td>
</tr>
</tbody>
</table>

$+/− = $ higher/lower than controls (columns 1 and 2) or dilation (column 3).
constitutes normal and abnormal parameter ranges, and to identify whether a single individual exhibits values within or outside these cohort-averaged ranges. For example, past studies have used cohort-averaged values to detect differences in WSS (23, 25, 26, 33); however, these studies have primarily investigated WSS at 2D slice locations or on regionally averaged surfaces. Thus, it is possible to miss regional variations in WSS if the 2D slice was placed elsewhere, or if the regional surface average did not constitute the complete abnormal region.

Based on the technique proposed here, abnormal WSS in a single subject (rather than cohort averaged values) may be detected at a singular surface location with coverage over the entire thoracic aorta lumen. Regional outliers (i.e. the abnormal WSS) are detectable by comparing WSS measurements in an individual patient to the mean and standard deviation WSS map for a healthy age/sex-matched WSS group. Abnormal WSS may be found by simply identifying regional deviations from the map by 2 SDs. Future studies with larger cohorts and appropriate control groups (matched for age, sex, aortic size, etc.) are warranted to use this technique to investigate the association of abnormal WSS with risk for aortopathy development (41). Ultimately, with a sufficient number of subjects (and thus sufficient statistical power) cohort-specific WSS atlases could be created. In addition, aortic tissue resected during aortic graft procedures may also be evaluated to investigate the correlation between abnormal WSS behavior (determined by the cohort-averaged WSS map) and molecular expression of biomarkers postulated to be associated with aortic remodeling (42).

It is important to emphasize that many studies investigating WSS, including those from our group, have been based on a limited number of manually placed 2D planes in the thoracic aorta (18–22, 24–27, 36). The WSS algorithm used in this study replaces the need for manual slice placement with a 3D segmentation step based on the systolic portion of the cardiac cycle. We have shown previously that WSS differences exist during the systolic portion of the cardiac cycle, whereas WSS in the diastolic portion is less active (33). This new methodology provides the opportunity to obtain a systolic WSS calculation over the entire thoracic aortic lumen, which reduces manual interaction to a single segmentation step and allows for comparisons of the entire aorta surface between cohorts. As a result, a comprehensive yet compact visualization of complex WSS patterns in multiple cohort-specific subjects is obtained, and WSS differences between patient groups can be easily interpreted, visualized, and quantified. In the future, these features may be beneficial for studies investigating risk stratification in patients with aortopathy and/or aortic valve disease.

A possible limitation of the study is the registration error of 20% found for the registration of the individual aortas to the cohort-specific aorta geometries or to the aorta geometry of the control cohort. This error metric was chosen to illustrate the differences of each individual aorta shape with the cohort-averaged geometry, and is designed to include every single voxel of the aortas. Even for very similar aortas, this error metric can generate large values. An important contribution to the error is the start and end point of each individual aorta. Not all aortas show the same signal enhancement in the left ventricle or the distal descending aorta. Therefore, the start and end point of the aorta segmentations is moderately variable. Furthermore, the location and length of the supra-aortic arteries and celiac trunk are different for each subject, contributing to the error. The main contributors to the error, however, are small differences in the aortic cross-section that cause substantial differences. This can be illustrated by eroding a specific 3D aorta segmentation by 1 voxel, which leads to a registration error of 40% with the same aorta. Therefore, it is quite remarkable that registration of different aortas with different diameters, supra-aortic arteries, and the celiac trunk lead to a registration error of only 20%. Note also that the percentage of the difference is determined by the average number of voxels between the individual aorta and the cohort-specific geometry (Eq. [4]). An alternate error metric, the percentage defined by the total number of voxels (individual aorta + cohort-specific geometry), would result in a reduced error of 11% ± 2%, 11% ± 3%, and 13% ± 5% for the control, dilation, and stenosis cohort, respectively.

Interpolation of the 3D WSS vectors from the individual aorta shape to the cohort-specific aorta geometry can introduce errors. However, it was found that the error was much smaller than the errors reported for registration: up to 5% for interpolation compared with 25% for registration. Therefore, the interpolation step is robust.

The aorta geometry used to create P-value maps can introduce errors in the comparison between cohorts. In this study, minor differences were found by comparing 3D WSS of the control and dilation cohorts on the control geometry and the dilation geometry. Therefore, the statistical results are largely independent of the reference geometry.

Another limitation of the study is the lack of a comparison of 3D WSS with previously used WSS algorithms calculated in 2D slices, manually placed perpendicular to the aorta (18–22, 24–27, 36). Such a comparison was outside the scope of this study, as the goal of the study was to describe the methodology and show the use of creating cohort-averaged 3D WSS maps. Previous studies have shown that both planar (18, 24) and volumetric (32) WSS analysis can be sensitive to differences in resolution and vessel lumen definition. A systematic evaluation of both planar and volumetric WSS analysis is thus warranted to better understand the performance of both techniques. Comparison of the 3D WSS algorithm with previously developed WSS algorithms (based on 2D slice placement) is part of ongoing work.

It is possible that WSS calculations are influenced by displacement artifacts (43) in the 4D flow MRI data related to rapid accelerations of blood flow, mainly present in the stenosis cohort. This implicates that the absolute WSS values calculated may be subject to error. By reporting relative WSS differences between different cohorts, errors in absolute WSS were minimized. Furthermore, it was demonstrated that the algorithm used for 3D WSS calculation based on spline fitting is robust in the presence of complex flow (44).
WSS underestimation as a function of resolution has been carefully studied and quantified (18, 39). In aorta phantoms with perfect parabolic flow and a spatial resolution as used in this study, an underestimation in WSS of 5% of the theoretical value was found (32). Segmentation errors, however, could result in errors up to 30% of the theoretical WSS values. Therefore, future work will elaborate on intersubject variability of WSS due to segmentation of aortic lumen. Nonetheless, computational fluid dynamics have demonstrated good agreement with 4D flow data, when discretization effects are considered (44). Given that there is no gold standard for 3D WSS measurement, we have chosen to emphasize the relative differences between cohorts as examined by the same imaging protocol.

We chose not to examine 3D WSS in different size aortas or differing grades of stenosis given the relatively low number of subjects in these pilot cohorts. However, when we assume a difference of 0.22 Pa and an SD of 0.16 Pa on the distal outer curvature (the region assumedly mostly prone to remodeling) between the controls and dilations cohorts (23), only seven more subjects are needed to obtain a power of 0.8.

The majority of the subjects were included via retrospective chart review rather than using prospective randomized enrollment; however, the primary goal was not to perform a longitudinal study, but rather to present a methodology to create cohort-averaged WSS surface maps over the entire thoracic aorta, and demonstrate the feasibility in a small pilot study to detect differences in normal physiologic and disease biomarkers. The addition of subjects and cohorts is part of ongoing work.

In conclusion, the methodology and application of aortic geometry and WSS maps in a range of subject cohorts was demonstrated. In this pilot study, the technique facilitated the identification of regionally altered WSS in the presence of aortopathy and aortic valve disease compared with healthy controls. This technique may prove useful for the creation of large cohort atlases representing hemodynamic biomarkers. The insights provided by this technique, combined with large scale randomized trials, may help clarify the role of WSS in vessel wall remodeling.

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