

# Demonstration of Patient-Specific Simulations to Assess Left Atrial Appendage Thrombogenesis Risk

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García-Villalba M, Rossini L, Gonzalo A, Vigneault D, Martinez-Legazpi P, Durán E, Flores O, Bermejo J, McVeigh E, Kahn AM and del Álamo JC (2021) Demonstration of Patient-Specific Simulations to Assess Left Atrial Appendage Thrombogenesis Risk. Front. Physiol. 12:596596. doi: 10.3389/fphys.2021.596596 <sup>1</sup> Bioengineering and Aerospace Engineering Department, Carlos III University of Madrid, Leganés, Spain, <sup>2</sup> Department of Mechanical and Aerospace Engineering, University of California, San Diego, La Jolla, CA, United States, <sup>3</sup> Department of Mechanical Engineering, University of Washington, Seattle, WA, United States, <sup>4</sup> Department of Bioengineering, University of California, San Diego, La Jolla, CA, United States, <sup>5</sup> Hospital General Universitario Gregorio Marañón, Instituto de Investigación Sanitaria Gregorio MaraMarañón, Facultad de Medicina, Universidad Complutense de Madrid, and CIBERCV, Madrid, Spain, <sup>6</sup> Department of Radiology, University of California, San Diego, La Jolla, CA, United States, <sup>7</sup> Division of Cardiovascular Medicine, University of California, San Diego, La Jolla, CA, United States, <sup>8</sup> Center for Cardiovascular Biology, University of Washington, Seattle, WA, United States, <sup>9</sup> Institute for Stem Cell and Regenerative Medicine, University of Washington, Seattle, WA, United States

Atrial fibrillation (AF) alters left atrial (LA) hemodynamics, which can lead to thrombosis in the left atrial appendage (LAA), systemic embolism and stroke. A personalized riskstratification of AF patients for stroke would permit improved balancing of preventive anticoagulation therapies against bleeding risk. We investigated how LA anatomy and function impact LA and LAA hemodynamics, and explored whether patient-specific analysis by computational fluid dynamics (CFD) can predict the risk of LAA thrombosis. We analyzed 4D-CT acquisitions of LA wall motion with an in-house immersed-boundary CFD solver. We considered six patients with diverse atrial function, three with either a LAA thrombus (removed digitally before running the simulations) or a history of transient ischemic attacks (LAAT/TIA-pos), and three without a LAA thrombus or TIA (LAAT/TIA-neg). We found that blood inside the left atrial appendage of LAAT/TIA-pos patients had marked alterations in residence time and kinetic energy when compared with LAAT/TIA-neg patients. In addition, we showed how the LA conduit, reservoir and booster functions distinctly affect LA and LAA hemodynamics. Finally, fixed-wall and moving-wall simulations produced different LA hemodynamics and residence time predictions for each patient. Consequently, fixed-wall simulations risk-stratified our small cohort for LAA thrombosis worse than moving-wall simulations, particularly patients with intermediate LAA residence time. Overall, these results suggest that both wall kinetics and LAA morphology contribute to LAA blood stasis and thrombosis.

Keywords: computational fluid dynamics, cardiovascular hemodynamics, atrial fibrillation, computed tomography, imaged-based simulation, immersed boundary method, thrombosis, left atrium

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## INTRODUCTION

Atrial fibrillation (AF) is the most common cardiac arrhythmia, affecting approximately 35 million people worldwide (Benjamin et al., 2019). During AF, the atria beat irregularly and their function is not coordinated with the ventricles. Each atrium has a morphologically characteristic body, and an appendage where thrombi form preferentially (Goette et al., 2017). Because some of these thrombi can travel to the brain, the risk of stroke of patients with AF is five times higher compared to the general population, and AF causes 15% of all strokes (Benjamin et al., 2019). Anticoagulation drugs reduce the risk of strokes in patients with AF. However, because these drugs are associated with an increased risk of bleeding, they are prescribed only to patients for whom the risk of stroke outweighs the bleeding risk. Current methods to risk-stratify patients are not personalized and contain no information about the patients' cardiac anatomy or blood flow; instead, they are based on demographic and clinical factors such as age, gender, or coexistent hypertension. These factors are derived from large clinical trials and have limited predictive value for a specific patient. Furthermore, there are currently many patients for whom there is clinical uncertainty as to whether anticoagulation is beneficial (January et al., 2014).

Because blood stasis is considered necessary thrombogenesis, we focus on the fluid mechanics of the left atrium (LA) and the left atrial appendage (LAA). The LA is a compliant structure that performs several functions during the cardiac cycle. It operates as a reservoir for the incoming flow through the pulmonary veins (PVs) during ventricular systole, as a conduit during early left ventricle (LV) diastole, and as a booster pump that augments LV filling during late LV diastole (Hoit, 2014). As a result of the LA complex geometry and motion, and its variability among the population, the flow dynamics within this chamber are complex, and thus far have not been fully characterized. Compared to the LV, there are few patient-specific computational fluid dynamics (CFD) analyses of the LA. In fact, early CFD studies of LA hemodynamics were part of whole-left-heart simulations focusing mostly on how LA geometry affects the LV flow features (Chnafa et al., 2014; Vedula et al., 2015; Lantz et al., 2016).

Computational fluid dynamics analyses focusing on the LA are relatively recent (Zhang and Gay, 2008). Koizumi et al. (2015) prescribed synthetic wall motions of the LA of a healthy subject to study how atrial function may contribute to washing out the LAA. Several other works followed a similar approach, offering somewhat contradictory results. Masci et al. (2020) studied two subjects with a history of AF, comparing the effects of LA wall motion in sinus rhythm with weak random motion, and concluding that reduced LAA contractility can result in slow blood flow. By contrast, Otani et al. (2016) studied two subjects with impaired atrial function and a history of AF but no evidence of prior stroke or TIAs, showing that the subject with better atrial function experienced more LAA blood stasis despite having higher flow velocities.

It appears that LAA morphology correlates with the risk of stroke in patients with AF (Di Biase et al., 2012; Bisbal et al., 2013; Yamamoto et al., 2014; Yaghi et al., 2020). However, the

degree of correlation is only moderate for the most commonly observed LAA shapes and the mechanisms responsible for this correlation are not well understood. Several CFD studies have explored the effect of LAA geometry on blood flow but, taken together, have offered ambiguous results. Bosi et al. (2018) found that LAA morphologies that are clinically associated with higher stroke incidence have impaired blood washout, whereas other studies (Garcia-Isla et al., 2018; Feng et al., 2019; Masci et al., 2019) reported that LAA stasis depends critically on additional factors such as PV inflow patterns.

Although the ambiguity in previous CFD analyses of LA hemodynamics could be attributed to their reduced sample sizes, it also suggests that blood transit in the LAA is a complex multi-factorial process that cannot be parameterized easily. Considering these challenges, previous efforts have been limited by their lack of a direct validation endpoint to test whether blood stasis increases the risk of LAA thrombosis. In this study, we performed patient-specific CFD simulations in a group of patients with varying atrial function and LAA geometries. Albeit still small (N = 6), this group is larger than those typically considered in CFD studies so far. Importantly, a subgroup of patients (N = 3) had either mural LAA thrombosis, which was segmented out prior to the simulations, or a history of transient ischemic attacks (TIAs). This approach offered direct insight into the relationship between atrial flow and LAA thrombosis. Moreover, we considered both moving-wall and fixed-wall anatomies to evaluate the dependence of blood stasis predictions on LA wall motion.

### MATERIALS AND METHODS

## Imaging of Human Subjects

We retrospectively studied N=6 patients. All of them were enrolled from a database of patients with existing computed tomography (CT) scans with full RR coverage over the entire LA and LAA. Selection of patients was random with the constraint of covering a wide range of LA ejection fractions as assessed by CT. Cardiac-gated cine CT scans were acquired at the National Institutes of Health (NIH), Bethesda, Maryland (N=3) and at the University of California San Diego (UCSD), CA, United States (N=3). The studies were approved by the Institutional Review Board at both centers.

Imaging was performed following standard clinical protocols at each participating center. Specifically, electrocardiogram (ECG) gating with inspiratory breath-hold was used to sample one entire heartbeat. The images were acquired with 3 different scanner models: 2 studies were obtained on a Siemens Force system (at NIH), 3 studies were obtained on the Toshiba Aquillion ONE (1 and NIH, 2 at UCSD) and one study was obtained on the 256-slice GE Healthcare Revolution CT at UCSD. For each image iodinated contrast was injected followed by saline flush. The doses and injection rates of contrast were chosen based on the patient's weight using standard clinical protocols.

The images were reconstructed using the manufacturers' standard algorithms (Toshiba reconstruction filter AIDR3D, FC08-cfa, GE Healthcare AsirV = 0.5 standard kernel, Siemens

filter kernel br36d). This procedure yielded DICOM files containing z-stacks of images of 512 x 512 pixels with in-plane pixels dimensions of between 0.32 mm and 0.48 mm in the x-y axial plane. Each z-stack contained between 100 to 320 slices with resolutions of 0.5 mm to 1 mm in the z-direction. The multiphase reconstructions were performed at regular time intervals across the heart cycle, with 5% RR and 10% RR being the minimum and maximum intervals.

## **Image Processing**

The 4D CT scans described above were imported into itk-SNAP (Yushkevich et al., 2006) and a segmentation was created using the semi-automatic active contour segmentation method. The PVs were clipped by manually choosing planes for each time frame; similarly, the LV and aorta were clipped off and the LAA was identified. From each 3D segmentation, a surface triangulation was extracted in MATLAB, smoothed and resampled to the resolution required for the CFD using iso2mesh (Qianqian and Boas, 2009) (an example is shown in Supplementary Figure 1). For each subject, the point cloud of vertices of the largest-LA-volume time frame was used as a reference surface and registered to the point clouds from the other time frames using the Coherent Point Drift algorithm (Myronenko and Song, 2010). The location of the centroid of each triangle was established by Fourier interpolation with  $N_{FOU}$  = 6 modes to match the temporal resolution of our CFD solver. This 4D kinematic LA model provides the position and velocity of the centroids of the triangles discretizing the LA surface at all times.

Indices of atrial function were calculated from CT images using standard definitions based on chamber volume, as well as on the waveforms of PV and transmitral flow (Hoit, 2014; **Table 1**). LA sphericity was calculated as described by Bisbal et al. (2013). The morphology of the LAA was categorized as one of the four types previously proposed by Di Biase et al. (2012) (chicken wing, cactus, windsock or cauliflower) by a cardiologist with expertise in cardiac CT interpretation. The LAA segmentations were skeletonized using the MATLAB image processing toolbox at each reconstructed phase of the cardiac cycle, and the branches of the skeleton were pruned, segmented and labeled automatically (**Figure 1**). The number of LAA lobes and the length of each lobe were measured (**Table 2**).

# **Computational Fluid Dynamics**

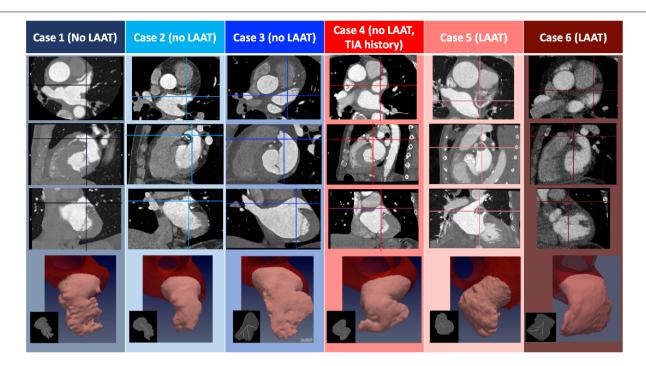
The simulations were performed with the in-house code TUCAN (Moriche, 2017; Moriche et al., 2017; Flores et al., 2020), which solves the incompressible Navier-Stokes equations for a Newtonian fluid of constant kinematic viscosity,  $\nu=0.04~{\rm cm^2/s}$ , using a fractional step method. The discretization used a staggered, Cartesian grid and centered, second-order, finite differences. Time integration was performed with a low-storage, three-stage, semi-implicit Runge–Kutta scheme.

Each simulation was initialized from no flow conditions and run for 10 heartbeats at a low resolution (grid spacing  $\Delta x = 0.090$  cm in all directions), with a constant time step  $\Delta t$  and a Courant number CFL < 0.3. The flow field was then interpolated into a finer grid ( $\Delta x = 0.051$  cm,  $256^3$  grid points) and run for another 10 heartbeats with the same CFL limit. This

TABLE 1 | Summary of anatomical and functional parameters of the left atrium (LA) in the LAAT/TIA-neg and (left three columns) and LAAT/TIA-pos (right three columns) groups.

Subject Number	1	2	3	4	5	6
LAA Thrombus (or history of TIA)	No	No	No	TIAs	Yes	Yes
Sinus Rhythm	Yes	Yes	Yes	No	No	No
Mean LA Volume (ml)	86.6	70.1	115	145	157	180
Min. LA Vol. (VLA,min, ml)	59.6	49.0	87.2	119	150	157
Max. LA Vol. (VLA,max, ml)	108	91.2	145	155	165	205
Pre-A-wave Volume (VLA,pre-A, ml)	94.3	72.3	119	149	153	178
LA Sphericity	0.80	0.82	0.78	0.86	0.83	0.81
Global (LA EF) [(VLA, max- VLA,min)/ VLA,max]	0.45	0.46	0.40	0.23	0.091	0.23
Expansion Index [(VLA,max- VLA,min)/ VLA,min]	0.81	0.86	0.66	0.30	0.10	0.31
Passive EF [(VLA, max- VLA,pre-A)/ VLA,max]	0.13	0.21	0.18	0.039	0.073	0.13
Booster EF [(VLA, pre-A - VLA,min)/ VLA,pre-A]	0.37	0.32	0.27	0.20	0.020	0.12
Conduit Volume / LV Stroke Volume [1- (VLA,max- VLA,min) / SV]	0.33	0.50	0.41	0.46	0.80	0.63
PVs (ml)	43	38	51	36	17	49
PVd (ml)	65	58	60	52	63	94
PVs-PVd ratio	0.66	0.66	0.84	0.69	0.27	0.52
PVa (ml)	17	7.3	12	13	0.14	1.4
PVa duration (cycles)	0.11	0.11	0.13	0.12	0.039	0.068
A-wave duration (cycles)	0.22	0.25	0.30	0.18	0.21	0.23
PVa – A wave duration ratio	0.49	0.43	0.44	0.64	0.18	0.30
E/A ratio	1.64	2.36	3.03	2.05	3.86	3.65

Mean volume indicates time-averaged volume. EF, ejection fraction; PVs, blood volume that enters the LA during LV systole; PVd, blood volume that enters the LA during LV diastole; PVa, blood volume that exits the LA due to reverse flow volume through the pulmonary veins during atrial contraction; E/A ratio, ratio of peak mitral velocities during early diastole (E-wave) and atrial contraction (A-wave). Supplementary Figure 2 provides an example of these parameters.



**FIGURE 1** Anatomical Features of Patient-Specific Subjects. The three first rows display CT images corresponding to the N = 6 subjects studied. The images represent, from top to bottom, axial, sagittal and coronal plane sections showing the left atrium, and including LAA thrombi if present (patients 5 and 6). In each panel, the intersections with the other two view planes are represented with a vertical and a horizontal line. The bottom row displays a 3D rendering of the segmented LAA for each subject, and an inset showing its plane projection (**gray**) and the branches that define the LAA lobes (**white**). Data are shown at an instant corresponding to 50% of the R-R interval.

TABLE 2 | Summary of morphological parameters of the left atrial appendage (LAA) in the LAAT/TIA-neg and (left three columns) and LAAT/TIA-pos (right three columns) groups.

Subject Number		1	2	3	4	5	6
LAA Thrombus or hi	istory of TIAs	No	No	No	TIAs	Yes	Yes
<b>LAA Volume</b> (ml)	Mean	6.94	4.85	14.3	10.7	15.5	22.0
	Minimum	4.32	3.14	10.2	9.10	13.8	19.8
	Maximum	8.97	6.28	17.9	11.6	17.4	24.7
LAA EF		0.52	0.50	0.43	0.22	0.21	0.20
LAA Shape	(CH = chicken, CA = cactus, WS = windsock, see ref. Bisbal et al., 2013)	CH	CA	CH	WS	WS	WS
	Number of lobes	$3 \pm 0.6$	$1\pm0$	$4 \pm 0.8$	$2 \pm 0.3$	$1 \pm 0.4$	$3 \pm 1.4$
	Length (cm)	$3.7 \pm 0.6$	$1.9 \pm 0.4$	$3.1 \pm 0.4$	$3.3 \pm 0.4$	$1.7 \pm 0.2$	$3.0 \pm 0.7$

EF, ejection fraction.

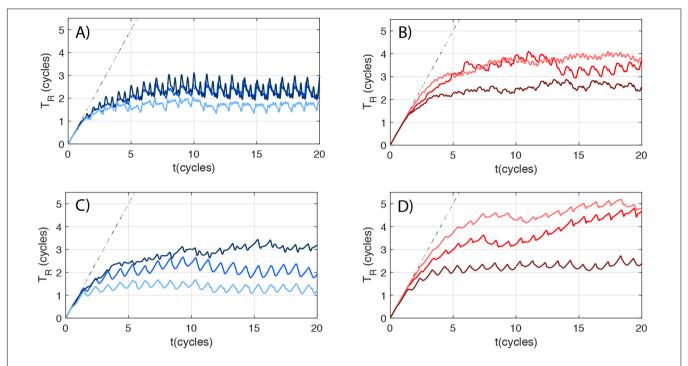
resolution is comparable to the finest of the resolutions employed in previous works (Vedula et al., 2015). In preliminary runs, an even finer grid was employed ( $\Delta x = 0.034$  cm), showing no relevant differences in the results.

The blood residence time ( $T_R$ ) was computed in the whole computational domain by solving the forced passive scalar equation (Rossini et al., 2016),

$$\frac{\partial T_R}{\partial t} + \vec{v} \cdot \nabla T_R = 1, \tag{1}$$

where  $T_R$  was set equal to zero at the start of the simulation and at the PVs. Eq. (1) was discretized like the Navier-Stokes

equations, but spatial fluxes were evaluated with a 3rd order WENO scheme (Shu, 1998). This scheme ensures a robust solution of the residence time, avoiding Gibbs phenomena near steep gradients without sacrificing accuracy elsewhere. Given that  $T_R$  was initialized at zero, we ran the simulations for several heartbeats to let  $T_R$  rise until it converged to a quasiperiodic (i.e., recurrent albeit not exactly periodic) behavior at all points inside the LA (**Figure 2**). This convergence period is basically the washout time of the fluid particle with highest  $T_R$ , and can be particularly large inside the LAA. In our simulations, plotting the volume-averaged residence time inside the LAA suggested that  $\sim$ 15 heartbeats were enough to achieve



**FIGURE 2** | Mean LAA residence time dynamics. Volume-averaged LAA residence time ( $T_R$ ) vs. time (t) for LAAT/TIA-neg (**blue, panels A,C**) and LAAT/TIA-pos subjects (**red, panels B,D**). Each patient's curve is colored following the scheme used in **Figure 1**. (**A,B**) Moving-wall simulations. (**C,D**) Fixed-wall simulations. The dashed curve in each panel represents  $T_R = t$ .

convergence (**Figure 2**). Hence, we ran our simulations for 20 heartbeats and used the last 4 heartbeats to compile statistics.

The LA surface was placed inside a cubic box of side 13 cm, using free-slip boundary conditions at the cube boundaries. The geometry and motion of the LA surface were prescribed using the patient-specific 4D kinematic LA models obtained from CT imaging as described in the previous two sections. To this end, we employed the immersed boundary method (IBM) (Mittal and Iaccarino, 2005). In these methods, the flow is solved in a fixed, Cartesian grid and the position and motion of surfaces are imposed by a localized volumetric force. In the present work, the IBM of Uhlmann (2005) was employed. Because the inlet planes of the PVs are not boundaries of the simulation domain, we also used the IBM to impose flow rates through these inlets. Specifically, the PV velocity was prescribed by applying a volumetric force  $f_{PV}$  in a buffer region upstream of each PV inlet plane. The target inlet velocity at each PV inlet plane was defined as  $v_t = \frac{\nabla_i}{A_i} n$ , where  $A_i$  and n are the cross-sectional area and the vector normal to each inlet's plane, respectively, and  $O_i$  (i = 1,...4) is the flow rate through each PV. The force was set proportional to the difference between the local velocity and the target inlet velocity, i.e.,  $f_{PV} = \frac{1}{\tau} (v_t - v)$ , where the constant  $\tau$  is the characteristic time of the forcing. For all simulations presented here, we chose  $\tau = 10\Delta t = 0.5$ ms, resulting deviations from the target velocity < 5% at the inlet. An analogous scheme was used to set the residence time to zero at the inlet of the PVs.

To specify the flow rates through the PVs, we calculated the combined flow rate through all the PVs,  $Q_{PV}$  from mass

conservation inside the left atrium. We computed the difference between the LA volume change obtained from each patient's segmentation and the flow rate exiting the LA through the mitral valve  $(Q_{MV})$ . The latter was obtained from the change in the ventricular volume during LV diastole, measured using each patient's CT images. Since we did not have patientspecific measurements of the flow rate through each PV,  $Q_{PV}$ was split evenly among all the PVs, i.e.,  $Q_i = Q_{PV}(t)/4$ . CFD analysis with evenly split pulmonary flow rates has been previously shown to produce reasonable LA hemodynamics when compared to phase contrast MRI data (Lantz et al., 2019). To further evaluate this choice of inflow boundary conditions, we performed preliminary simulations for one of the cases comparing this approach to splitting  $Q_{PV}$  proportionally to the area of the pulmonary veins, so that the bulk mean velocity was the same in all PVs. We did not find any significant differences between these two approaches in terms of the results presented in section 3. For the outflow boundary condition, the MV is considered as a plane that instantly opens (when  $Q_{MV}>0$ ) and closes (imposing  $Q_{MV}=0$ ). No specific model is considered for the geometry of the MV leaflets, since they are not expected to affect the flow structure inside the LA (Vedula, 2015).

To explore the influence of the motion of the LA wall, we performed corresponding simulations keeping the position of the LA walls fixed at the zero mode of its Fourier temporal series. For the fixed wall-simulations, the A-wave was digitally removed from the transmitral flow, setting  $Q_{MV}=0$  for the duration of the atrial contraction.

## **RESULTS**

# **Subject Population**

The median age of the study subjects was 65 (range 50 – 92), and two were female. Subjects 1—3 were imaged in sinus rhythm and did not have LAA thrombus. Subjects 5 and 6 were imaged in AF with an LAA thrombus, which was digitally removed prior to segmenting the LA geometry for the CFD simulations. Subject 4 was imaged in atrial fibrillation without LAA thrombus, but had a previous history of transient brain ischemic attacks (TIAs). AF was determined based on ECG rhythm strips and lack of late diastolic LA/LAA ejection ("atrial kick"). Based on this information, we categorized subjects 1 – 3 as LAA-thrombus/TIA negative (LAAT/TIA-neg), and subjects 4 – 6 as LAAT/TIA-pos.

Subjects 1 – 3 had normal LA functions as inferred from CT-derived chamber volumes and PV flows (**Table 1**). Subjects 4 – 6 had enlarged LA with impaired global function. Subject 4 had decreased reservoir function but relatively normal booster function. Subjects 5 and 6 had impaired reservoir and booster function. The LA chamber of subject 5 had severely impaired function, operating mostly as a conduit between the PVs and the LV.

Figure 1 displays cardiac CT images of the 6 subjects including segmentations of their LAAs and a plane projection of the morphological skeletonization of the appendage. These images show that subject 5 had an obvious isolated LAA thrombus in the central cavity of the LAA, while subject 6 had the distal section of the LAA filled with thrombus (which was confirmed on a "late" single acquisition obtained 22 seconds after the dynamic acquisition). The LAA skeletonization was used to determine LAA lobe number (i.e., the number of skeleton branches) and LAA length (i.e., the length of the branch whose end is most distal to the LAA orifice). Table 2 contains detailed morphological information of the LAA for our study subjects. While the small size of the cohort precludes detailed statistical examination, it is worth noting that the LAAT/TIA-pos subjects had more complex LAA shapes (i.e., windsock vs chicken or cactus) and larger LAA volumes than the LAAT/TIA-neg subjects. We did not observe striking differences in LAA lobe count or LAA length (**Table 2**).

## Flow Visualization

We first present 3D velocity vector maps for one of the subjects with normal atrial function (subject 3, Figure 3). The vectors are colored according to their proximity to four anatomical landmarks: the left PVs, the right PVs, the LAA and the mitral annulus. Three representative instants of the cardiac cycle are shown: LA diastole, LV early filling, and LA systole (Figure 3, top, center and bottom row, respectively). We juxtaposed movingwall and fixed-wall simulations (Figure 3, left and right column, respectively) to illustrate how wall motion affects LA flow dynamics. The flow rate profiles in the PVs and mitral annulus, and the volume profiles of the LA and LAA are also plotted.

Starting at atrial diastole, the mitral valve is closed, and LA volume is increasing as blood enters the chamber in the moving-wall simulations (**Figure 3A**, blue arrows). The collision of the left and right PV jets redirects blood towards the mitral

annulus (red arrows) and the LAA (green arrows). In contrast, the fixed-wall simulation shows almost no fluid motion at the same timepoint (**Figure 3B**). Given that the chamber volume does not vary in fixed-wall simulations and the mitral valve is closed, the only possible fluid motion is the remnant flow from ventricular diastole.

During LV relaxation (E-wave, Figure 3C), blood still enters the LA through the PVs forming a jet collision pattern, but LV suction accelerates the blood towards the mitral annulus. Consequently, the flow rate exiting through the mitral annulus  $(O_{MV})$  exceeds the incoming flow rate through the PVs  $(O_{PVs})$ and LA volume decreases. The LAA also contracts during this phase, driving a relatively uniform outward flow in its interior. The fixed-wall simulation (Figure 3D) exhibits stronger flow during LV filling even if it was run with the same  $Q_{MV}$ . In the moving-wall simulation, passive LA contraction leads to a net difference between the outgoing and incoming flow rates  $Q_{MV}$  -  $Q_{PV} \approx 400$  ml/s. However, this difference must be zero in the fixed-wall simulation - thus, PV inflow jets are stronger. The LAA flow pattern in the fixed-wall simulation is reminiscent of a shear-driven cavity flow (Koseff and Street, 1984), consisting of a weak recirculating pattern driven by the external flow in the atrial body. This pattern is notably different to the one observed in the moving-wall simulations.

During atrial systole, chamber contraction drives flow into the ventricle (A-wave, **Figure 3E**) and simultaneously creates backflow into the PVs (Ar-wave). LAA contraction drives a relatively uniform outward flow inside the appendage, similar to the E-wave albeit somewhat slower. In contrast, the fixed-wall simulation displays a weak recirculatory flow corresponding to the late decay stage of the PV and mitral annulus jets generated during LV filling. In particular, the velocities in the LAA are much smaller than in the moving-wall simulation.

# The Effect of Atrial Function on Flow Patterns

The results presented so far reveal that blood flow in the LA and LAA is sensitive to atrial wall motion. However, wall motion can be reduced in different ways depending on whether a patient has impaired reservoir and/or booster function (Hoit, 2014). Thus, we visualized LA and LAA flow patterns across the phases of the cardiac cycle for three patients: one with normal atrial function (subject 3, **Figure 4A**), one with impaired reservoir function (subject 4, **Figure 4B**), and one with both impaired reservoir and booster functions (subject 5, **Figure 4C**). Each snapshot shows blood velocity vectors inside the LAA and iso-surfaces of the second invariant of  $\nabla \vec{v}$ , which reveal the vortical structures of the atrial body. Similar to **Figure 3**, we juxtapose data from moving-wall and fixed-wall simulations.

The subject with normal atrial function exhibits two vortexring systems associated with the atrial filling jets discharging from the left and right PVs (**Figure 4A**). These vortex rings collide into each other and break down into a complex network of fine-scale vortices that fills the whole atrial body, and which is sustained throughout the cardiac cycle. The fixed-wall simulation presents two strong vortex-ring systems that break down into

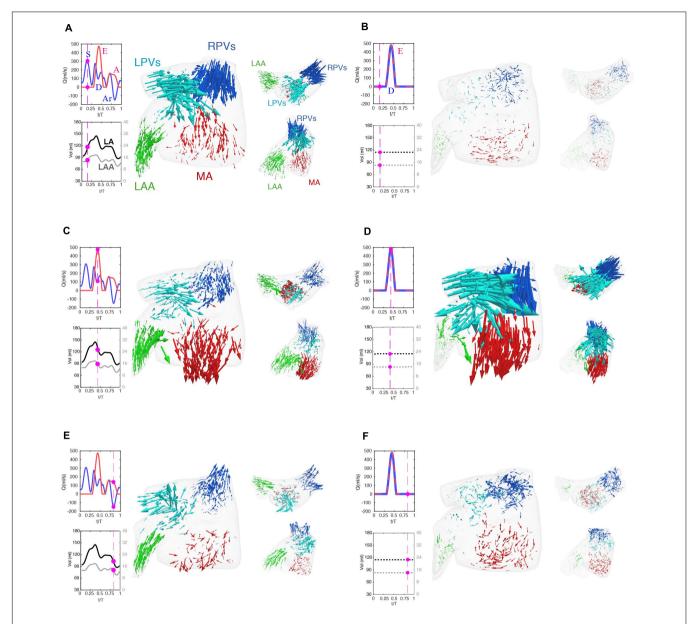
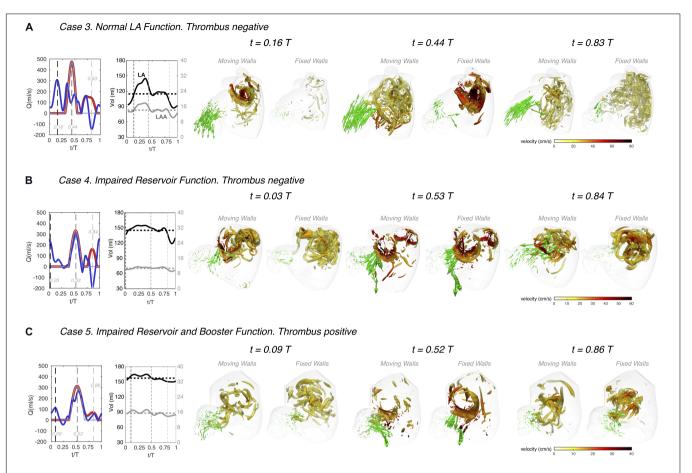


FIGURE 3 | Flow visualization in a left atrium with normally moving and fixed walls. (A,C,E) Simulation with prescribed patient-specific LA wall motion as obtained from time-resolved CT imaging. (B,D,F) Simulation in which the position of the LA walls of the same patient were at its time-averaged position. Each panel displays three views of an instantaneous 3-D vector map of blood flow velocity in the left atrium. Clockwise: larger frontal view, smaller axial view and smaller oblique view emphasizing the flow in the left atrial appendage. The vectors are colored according to their proximity to the right pulmonary veins (blue), left pulmonary veins (cyan), left tarial appendage (green) and mitral valve (red), and are scaled with the magnitude of blood speed. The panels also include time histories of the flow rate through the mitral valve (red) and the cumulative flow rate through the pulmonary veins (blue), and of the volumes of the left atrium (black) and the left atrial appendage (gray). The magenta bullets indicate the instant of time represented in the vector plots of each panel. The main waves of these profiles are indicated in the plots, i.e., the E-wave and A-wave of transmitral flow during early and late LV filling, the S-wave and D-wave of forward PV flow during LV systole and diastole, and the Ar-wave of backward PV flow during LV diastole, respectively. In the fixed-wall simulations, LA volume is constant and therefore the only transmitral and PV flow waves are the E-wave and the D-wave, and these two waves coincide. The flow vectors inside the LA are represented at three instants of time. (A,B) t = 0.16 T, which corresponds to atrial diastole and peak flow rate through the pulmonary veins in the moving-walls subject. (C,D) t = 0.44 T, which corresponds with left ventricular diastole and peak flow rate through the mitral valve. (E,F) t = 0.94 T, which corresponds with atrial systole and peak backflow rate through the pulmonary vein.

finer-scale vortices, but these decay almost completely by atrial diastole of the next cardiac cycle. As shown in **Figure 3**, the magnitude and dynamics of the velocity vectors in the LAA of this normal subject vary markedly between the moving-wall and fixed-wall simulations.

In patients with impaired atrial function, moving-wall and fixed-wall simulations yield qualitatively similar LA vortical structures, while the velocity fields inside the LAA exhibit interesting differences. In patient 4 (**Figure 4B**), who had severely impaired reservoir function (LA expansion index = 0.3,



**FIGURE 4** | Left atrial vortex patterns and left atrial appendage flow velocity in subjects with normal and impaired atrial function. Three subjects are shown, one with normal atrial function (**panel A**, same subject as in **Figure 3**), one with impaired reservoir function (**panel B**), and one with both impaired reservoir and booster functions (**panel C**). For each subject, instantaneous 3D flow visualizations are shown in the frontal view of the LA using data the moving-walls and fixed-walls simulations. In each visualization, we rendered the flow velocity vectors in the LAA vicinity and vortex cores in the whole LA (as detected by thresholding the second invariant of the velocity gradient tensor,  $Q = 1000 \text{ s}^{-2}$ ). The LAA velocity vectors were amplified by a factor of 4 with respect to those in **Figure 3**, and the surface of the vortex cores was colored according to the local absolute value of blood velocity, as indicated in the colorbars. Each panel includes plots of the time histories of flow rate through the mitral valve (**red**) and the cumulative flow rate through the pulmonary veins (**blue**), both for the moving-walls (dark color, thick continuous lines), and fixed-walls (light color, thin dashed lines) simulations. The panels also include each patient's LA volume (**black**) and LAA volume (**gray**) vs. time. For each subject, flow visualizations are shown at three instants of time. The first one corresponds to atrial diastole [i.e., peak flow rate through the pulmonary veins in the moving wall simulations, t/T = 0.16, 0.03 and 0.09 in panels (**A-C**)]. The second instant of time corresponds to atrial systole [i.e., the A-wave of left ventricular filling and peak backflow rate through the pulmonary veins, t/T = 0.83, 0.84, and 0.86 in panels (**A-C**)]. These three instants of time are indicated with black, dark gray and light gray dashed vertical lines in the flow-rate plots at the left-hand-side of each panel.

see Table 1), the LA wall was akinetic during atrial diastole and LV relaxation. Consequently, the moving-wall and fixed-wall simulations produced similar LAA velocity patterns during these phases. However, this patient had a relatively normal booster function (booster LA ejection fraction = 0.2) with normal LA wall kinetics during atrial systole. As a result, blood velocity in the LAA showed and outward flow pattern for the moving-wall simulations whereas it showed a weak recirculating motion for the fixed-wall simulations. In subjects with decreased reservoir and booster function, e.g., subject 5 in Figure 4C, the moving-wall and fixed-wall simulations produced qualitatively similar LAA velocity distributions across the whole cardiac cycle. Consistent with patient 5 having more passive than booster LA function, the most appreciable difference between fixed-wall and

moving-wall simulations was observed for atrial diastole – during this phase, the moving-wall LAA has a weak inward-flowing pattern in addition to the swirl that is found in the fixed-wall LAA.

# The Effect of Atrial Function on Residence Time and Kinetic Energy

To visualize the effect of impaired atrial function on blood stasis inside the LA chamber, we mapped the blood residence time in two oblique intersecting plane sections of the LA body and the LAA. These maps are represented in **Figure 5** for the same three study subjects of **Figure 4**, both for moving and fixed walls. Once converged ( $t \ge 15$  cycles, see **Figure 2**), the  $T_R$  maps show

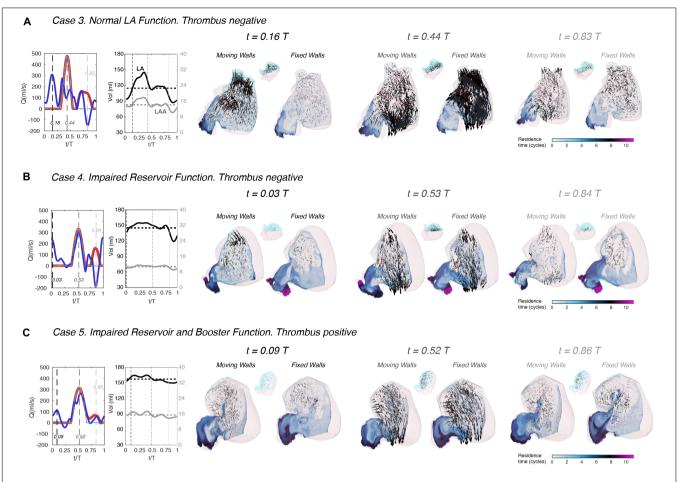


FIGURE 5 | Left atrial blood residence time in subjects with normal and impaired atrial function. Three subjects are shown, one with normal atrial function (panel A), one with impaired reservoir function (panel B), and one with both impaired reservoir and booster functions (panel C). Each subject includes data from both moving-walls simulations and fixed-walls simulations. For each subject, instantaneous 2D fields of the absolute value of the residence time  $T_R$  (see main text for definition) are plotted in two oblique plane sections of the LAA and the LA body that separate the cyan and pink parts of the sagittal view insets. Additionally, 3D blood velocity vectors are rendered behind the sectioning planes (i.e., inside the cyan part of the LAA and the LA body). Similar to Figure 4, each panel includes plots of the time histories of flow rate through the mitral valve (red) and the cumulative flow rate through the pulmonary veins (blue), both for the moving-walls (dark color, thick continuous lines) and fixed-walls (light color, thin dashed lines) simulations. The panels also include each patient's LA volume (black) and LAA volume (gray) vs. time. The data are represented at three instants of time. The first one corresponds to atrial diastole [i.e., peak flow rate through the pulmonary veins in the moving wall simulations, t/T = 0.16, 0.03, and 0.09 in panels (A-C)]. The second instant of time corresponds to left ventricular rapid filling [i.e., the A-wave of left ventricular filling and peak backflow rate through the pulmonary veins, t/T = 0.83, 0.84, and 0.86 in panels (A-C)]. These three instants of time are indicated with black, dark gray and light gray dashed vertical lines in the flow-rate plots at the left-hand-side of each panel.

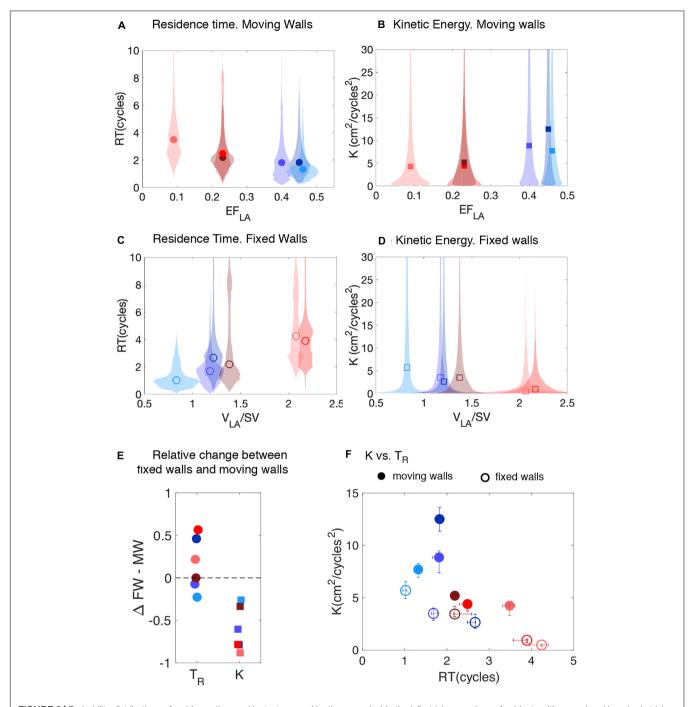
that blood is systematically more stagnant inside the LAA than in the LA body. Residence time gradients are also significantly sharper inside the LAA leading to elevated TR at the LAA apex, consistent with flow stirring being lower in the appendage than in the atrial body.

Visual inspection of the  $T_R$  maps suggests that LAA blood stasis in LAAT/TIA-neg patients (**Figure 5A**) was less prominent than in LAAT/TIA-pos patients (**Figures 5B,C**). To confirm this result, we compiled the statistical distributions of LAA  $T_R$  for all of our subjects using 200 time instants equally distributed over the cardiac cycle. We also compiled statistics of the kinetic energy K = 1/2 ( $u^2 + v^2 + w^2$ ), since this hemodynamic variable is more accessible via medical imaging than  $T_R$ . The data from moving-wall simulations show that LAA residence time increases

as  $EF_{LA}$  worsens across different subjects (**Figure 6A**). Although less clear, we also found a trend for K to decrease as  $EF_{LA}$  worsens (**Figure 6B**). In the fixed-wall simulations all subjects have zero ejection fraction, but the data suggests that subjects with higher LA volume tend to have higher  $T_R$ , and lower K (**Figures 6C,D**).

# Hemodynamic Parameters and Thrombosis Risk on Moving-Wall and Fixed-Wall Simulations

**Figures 6A,B** demonstrates that blood inside the left atrial appendage of LAAT/TIA-pos patients had marked alterations in hemodynamic parameters when compared with LAAT/TIA-neg patients. Specifically, LAAT/TIA-pos patients had 57% higher



**FIGURE 6** | Probability distributions of residence time and instantaneous kinetic energy inside the left atrial appendage of subjects with normal and impaired atrial function. **(A,B)** Violin plots of the probability density function (shaded patches) of  $T_R$ , and K inside the LAA of the N=6 subjects studied, together with their median (symbols). The data come from the moving-wall simulations and are plotted vs. the LA ejection fraction of each subject. **(C,D)** Violin plots of the probability density function of the same variables as obtained from fixed-wall simulations. The data are plotted vs. the volume of the LA normalized by stroke volume. **(E)** Relative change in mean values of  $T_R$ , and K between the fixed-wall (FW) and the moving-wall (MW) simulations, defined as  $\Delta T_R = \frac{T_{R,PW} - T_{R,MW}}{T_{R,MW}}$ , etc. **(F)** Scatter plot of K vs  $T_R$  for moving-wall (solid symbols) and fixed-wall (open symbols) simulations. The error bars in each panel span the 95% confidence intervals of the mean. Solid and hollow symbols come, respectively, from moving-wall and fixed-wall. In all the panels, each subject's data is colored using the same scheme used in **Figure 1** (blue shades indicate LAAT/TIA-neg patients and red shades indicate LAAT/TIA-pos patients).

mean residence time and 69% lower mean kinetic energy inside their LAA than LAAT/TIA-neg patients. Given that these trends seemed to hold for the fixed-wall simulations (94% higher mean

residence time, 216% mean kinetic energy, **Figures 6C,D**), we examined the variation in LAA  $T_R$  and K between fixed-wall and moving-wall simulations (**Figure 6E**). This analysis showed

that K decreased in the fixed-wall simulations across all study subjects; however, the variations in  $T_R$  were not systematic. In particular, fixed-wall simulations yielded equal or lower LAA  $T_R$  than moving-wall simulations for two of the three LAAT/TIA-neg subjects. While these results are not easy to reconcile with the dependence of  $T_R$  on  $EF_{LA}$  (Figure 6A), they are consistent with our finding that fixed-wall simulations tend to represent the LA and LAA flow patterns better in atria with impaired function than in normal atria (Figure 3). Finally, we examined whether fixed-wall simulations were less accurate in separating the LAAT/TIA-pos and LAAT/TIA-neg groups in our small study population. Scatter plots of LAA  $T_R$  and K (Figure 6F) showed some overlap between the LAAT/TIA-pos and LAAT/TIA-neg groups in the fixed-wall simulations. In contrast, moving-wall simulations provided good clustering of the groups for either  $T_R$  or K.

## DISCUSSION

Thrombogenesis is most likely to occur with the concurrence of endothelial damage or dysfunction, blood stasis, and the presence of procoagulatory factors in blood. Blood flow governs the transport of the enzymes, chemical factors, and platelets to and from endothelial damage sites (Fogelson and Neeves, 2015). Blood stasis permits the reactive species to remain close enough to each other to initiate clot formation (Seo et al., 2016). The goal of this study was to investigate how LA function influences LAA blood stasis. More specifically, we hypothesized that CFD analyses of LA/LAA hemodynamics could identify patients at increased risk of LAA thrombosis.

# Atrial Function, Flow Patterns, and Blood Stasis

By selecting subjects with a wide range of LA ejection fractions ( $EF_{LA}$ ), we characterized how LA and LAA flow patterns vary with atrial function. Consistent with previous MRI and CFD reports of flow in the LA (Kilner et al., 2000; Fyrenius et al., 2001; Vedula et al., 2015), we found that blood from the PVs flows mostly towards the mitral annulus during ventricular diastole and systole. In subjects with impaired reservoir function, reduced wall motion during ventricular systole caused slower flow and quicker dissipation of fine-scale vortices.

In the LAA, normal wall kinetics drove inward and outward flow jets. By contrast, flow inside the LAA was similar to a shear-driven cavity flow in subjects with impaired booster and reservoir function, consistent with Masci et al. (2019) data. Additionally, our subject with impaired reservoir function and normal booster function experienced hybrid LAA flow dynamics that oscillated between lid-driven cavity flow during LA diastole, and a weak outflow jet during LA systole. Taken together, our results indicate that flow patterns inside the LAA are not as complex as in the LA body or the LV, which we attribute to the strong geometrical confinement of the flow inside the appendage. These flow patterns led to elevated blood residence in the LAA compared to the LA body, in agreement with previous CFD results (Koizumi et al., 2015; Otani et al., 2016; Bosi et al., 2018; Garcia-Isla et al., 2018; Masci et al., 2019, 2020), and the

consensus that thrombi form preferentially in the LAA. Patients with impaired atrial function also showed a trend for blood stasis to appear near the mitral annulus, since their weaker LA body vortical structures caused decreased stirring compared to normal subjects.

# **Blood Transit and LAA Thrombogenesis**

While previous CFD studies suggested that reduced wall motion impairs LA blood transit (Otani et al., 2016; Masci et al., 2019, 2020), there is a lack of systematic data about how functional parameters such as  $EF_{LA}$  affect this transit. We demonstrated that LAA  $T_R$  has a clear dependence on  $EF_{LA}$ . We also showed how LAA  $T_R$  is related to the risk of thrombosis, leveraging that our database includes two subjects with LAA thrombi and one with a history of TIAs. We and others have recently shown that LV blood stasis in the setting of myocardial infarction precedes LV mural thrombosis (Martinez-Legazpi et al., 2017; Garg et al., 2019) and cerebral microembolism (Delgado-Montero et al., 2019). The present study extends this idea to the LAA showing that LAA thrombosis could occur when the mean residence time inside the LAA exceeds  $\sim$ 2 s (i.e., 2 cycles). While the number of subjects in our pilot study is too low to make statistically significant conclusions, this threshold agrees with Martinez-Legazpi et al. (2017) threshold for LV thrombosis derived from N = 72 patients. This agreement is particularly notable considering the different etiologies of myocardial infarction and AF, and the different methodologies used in our two studies.

In addition to residence time, we show that kinetic energy could be potentially useful to predict the risk of LAA thrombosis. This knowledge could be useful because, while  $T_R$  is a cumulative variable whose calculation requires solving a partial differential equation, calculating K is significantly less involved. In particular, K should be accessible by medical imaging modalities with moderate time resolution, such as phase-contrast MRI. Nevertheless, the observed predictive ability of K in the case of the LAA could be in large part due to the unique geometrical characteristics of the appendage and its particular location within the LA. We note that K can yield poor predictions of blood stasis in other scenarios (Hendabadi et al., 2013).

## Fixed-Wall vs. Moving-Wall Simulations

Fixed-wall simulations provided flow fields and stasis maps that differed from those obtained in moving-wall simulations, especially in subjects with normal atrial function. This discrepancy is not surprising; in addition to neglecting LA wall motion, fixed-wall simulations have oversimplified inflowoutflow boundary conditions and ignore LV - LA interactions. On the other hand, fixed-wall simulations only require static images which are more commonly available in clinical practice, and are less involved from the point of image segmentation and CFD analysis. Previous studies considered fixed walls (Bosi et al., 2018; Garcia-Isla et al., 2018) but the effect of this simplification on the simulation results had not been evaluated before. Furthermore, given the difficulty of accurately modeling wall motion and hemodynamics in fibrillating atria (Boyle et al., 2020), comparing fixed-wall and moving-wall simulations can provide useful information about how to interpret CFD results.

In our reduced study cohort,  $T_R$  and K in the fixed-wall LAA were less accurate predictors of LAA thrombus or TIAs than their moving-wall counterparts. These findings warrant further evaluation in larger patient cohorts and reflect the influence of LA and LAA geometry on blood flow. Morphological analysis of the LA chamber in our subject cohort (Table 1) revealed that the LAAT/TIA-pos subjects had the largest LA volumes and showed a trend for higher LA sphericity, consistent with clinical reports of stroke rates vs. LA/LAA geometry in AF patients (Di Biase et al., 2012; Bisbal et al., 2013; Yamamoto et al., 2014). Moreover, our LAAT/TIA-pos subjects, which had the highest LAA residence time in both our moving-wall and fixed-wall simulations, also had LAA morphologies that are ranked more pro-thrombogenic according to Di Biase et al. (2012) categorization. Although the number of LAA lobes measured by 3D ultrasound imaging and stroke risk have been shown to correlate AF patients (Yamamoto et al., 2014), we did not observe a relationship between LAA lobe number and  $T_R$ , K or presence of LAAT/TIAs (**Table 2**). This discrepancy could be caused by our small cohort size and differences in spatial resolution between CT and ultrasound.

# Strengths and Limitations of the Present Study

The number of patients considered in this study was small (N=6), even considering that we performed simulations with both moving and fixed walls for each subject. However, our patient cohort was diverse enough to explore how different aspects of LA function (e.g., reservoir, booster pump, and conduit) affect the flow inside this chamber with an unprecedented level of detail. Also, while small, the number of subjects considered here is higher than in most previous CFD studies (Otani et al., 2016; Bosi et al., 2018; Masci et al., 2020). More importantly, by including patients without LAA thrombus and with LAA thrombus or history of TIAs, we were able to demonstrate that CFD analysis of LA flow could predict the risk of LAA thrombogenesis.

We did not consider a model of the coagulation cascade in this study. Biochemical modeling of thrombogenesis could be incorporated into the CFD framework (Seo et al., 2016). However, the additional models would markedly increase the cost of the simulations. Also, the accuracy of these models is unclear, given their dependence on many reaction constants whose values are difficult to obtain in a patient-specific basis.

We considered the same heart rate and blood viscosity for all patients to eliminate these parameters as independent variables in our analysis, even if these parameters may vary from patient to patient. Besides, we note that heart rate in particular can vary significantly for each individual due to physical activity, stress, circadian rhythm and, of course, rhythm disorders such as AF. Analyzing how these intra-patient variations affect CFD predictions was beyond the scope of this study. In LAAT-pos patients, we segmented the thrombus and removed it before performing the CFD analyses in an effort to recapitulate the conditions that caused thrombosis. However, the presence of a thrombus could constrain the motion of the LAA walls – thus, pre-thrombus LAA wall motion might have not been fully

recovered. One of our subjects had a history of TIAs of unknown origin, which we presumed to be cardioembolic from the LA/LAA, but we cannot exclude the possibility of another etiology.

Similar to previous simulations (Koizumi et al., 2015; Otani et al., 2016; Bosi et al., 2018; Garcia-Isla et al., 2018; Masci et al., 2020), our LA model did not consider the mitral valve in the outflow of the chamber. This choice is justified by whole left heart CFD studies showing that the mitral valve does not significantly affect LA flow (Vedula, 2015; Vedula et al., 2015). Mitral regurgitation, which can affect 15-25% of AF patients (Delgado and Bax, 2017; Owens et al., 2017), would require modeling the mitral valve. Our work improves over most previous computational studies of the LA flow (Bosi et al., 2018; Garcia-Isla et al., 2018; Masci et al., 2020) by considering patient-specific inflow/outflow boundary conditions, similar to Otani et al. (2016). One limitation of our simulations is that PV flow rates were evenly distributed (Lantz et al., 2019). More patient-specific modeling of inflow boundary conditions would involve measuring the flow rate through each pulmonary vein (e.g., by transesophageal echocardiography) or modeling the flow impedance of each pulmonary vein based on imaging data such as vein diameter (Shi et al., 2011).

As discussed above, fixed-wall simulations have severe limitations as a model of AF, including but not limited to their inability to capture LA motion and deformation mediated by the chamber's passive response to incoming PV flow and LV-LA interactions. The rationale for including fixed-wall simulations in our study was to assess how these limitations affect the prediction of patient-specific hemodynamic parameters such as residence time, by comparing with moving-wall simulations. This assessment is warranted considering that several previous studies used fixed-wall simulations (Bosi et al., 2018; Garcia-Isla et al., 2018).

Finally, in patients with paroxysmal AF, imaging performed while patients are in sinus rhythm might have to be used to predict LAA thrombosis risk. Multi-scale models of atrial electro-mechanics coupled to CFD analysis (reviewed, e.g., in Boyle et al., 2020) could tackle this problem. However, these models combine patient-specific data (e.g., anatomical images) with data that is not patient-specific (e.g., fiber orientation, LA wall mechanical properties, channel gating kinetics, and fibrosis distribution). Furthermore, we note that, while there is an unequivocal correlation between AF and embolic stroke, the long-held paradigm that thrombi form in LAA during AF episodes, embolize, and travel to the brain causing stroke is now under scrutiny, after a recent clinical trial revealed absence of AF episodes in the month preceding thromboembolic events (Brambatti et al., 2014).

# CONCLUSION

This work combined 4D-CT acquisitions of left atrial wall motion with immersed-boundary CFD analyses of LA blood flow to establish proof of principle of personalized risk stratification of LAA thrombogenesis for patients with AF that could be

translated to the clinical setting. For the large number of AF patients for whom there is currently uncertainty regarding optimum medical treatment, the proposed risk stratification framework may allow clinicians to better balance the benefits of chronic anticoagulation therapy (or LAA occlusion therapy) with the hemorrhagic risks of therapy. Larger studies are warranted to evaluate the robustness and potential clinical utility of this methodology.

### **DATA AVAILABILITY STATEMENT**

The original contributions presented in the study are included in the article/**Supplementary Material**, further inquiries can be directed to the corresponding author/s.

## **ETHICS STATEMENT**

The studies involving human participants were reviewed and approved by the National Institutes of Health, Bethesda, MD, United States and the University of California San Diego, CA, United States. The patients/participants provided their written informed consent to participate in this study.

## **AUTHOR CONTRIBUTIONS**

MG-V, OF, EM, AK, and JÁ designed the research. MG-V, LR, AG, DV, ED, OF, EM, AK, and JÁ performed the research.

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## SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fphys. 2021.596596/full#supplementary-material

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**Conflict of Interest:** The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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