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Whole Heart Coronary Imaging with Flexible Acquisition Window and Trigger Delay

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Abstract

Coronary magnetic resonance imaging (MRI) requires a correctly timed trigger delay derived from a scout cine scan to synchronize k-space acquisition with the quiescent period of the cardiac cycle. However, heart rate changes between breath-held cine and free-breathing coronary imaging may result in inaccurate timing errors. Additionally, the determined trigger delay may not reflect the period of minimal motion for both left and right coronary arteries or different segments. In this work, we present a whole-heart coronary imaging approach that allows flexible selection of the trigger delay timings by performing k-space sampling over an enlarged acquisition window. Our approach addresses coronary motion in an interactive manner by allowing the operator to determine the temporal window with minimal cardiac motion for each artery region. An electrocardiogram-gated, k-space segmented 3D radial stack-of-stars sequence that employs a custom rotation angle is developed. An interactive reconstruction and visualization platform is then employed to determine the subset of the enlarged acquisition window for minimal coronary motion. Coronary MRI was acquired on eight healthy subjects (5 male, mean age = 37 ± 18 years), where an enlarged acquisition window of 166–220 ms was set 50 ms prior to the scout-derived trigger delay. Coronary visualization and sharpness scores were compared between the standard 120 ms window set at the trigger delay, and those reconstructed using a manually adjusted window. The proposed method using manual adjustment was able to recover delineation of five mid and distal right coronary artery regions that were otherwise not visible from the standard window, and the sharpness scores improved in all coronary regions using the proposed method. This paper demonstrates the feasibility of a whole-heart coronary imaging approach that allows interactive selection of any subset of the enlarged acquisition window for a tailored reconstruction for each branch region.

Introduction

High-resolution MRI of the heart for applications such as coronary artery imaging provides unique challenges among most anatomical imaging due to both cardiac and respiratory motion. Recent advances in hardware and software MR technologies have allowed whole-heart...
imaging, specifically for the visualization of the coronary arteries, to be performed in a clinically feasible scan time at both 1.5T and 3T [1–7]. In practice, high-resolution whole-heart and targeted cardiac imaging utilize an ECG-gated k-space segmented acquisition strategy, which samples the 3D k-space volume only during the quiescent period of the cardiac cycle. For all of these methods, a correctly timed trigger delay (TD) to synchronize the segmented k-space acquisition with the quiescent period of the cardiac cycle is essential. Therefore, a prior scout cine acquisition is typically employed prior to the coronary imaging sequence to determine the optimal TD value for each individual [8]. However, subjects demonstrating heart-rate variability may present image quality degradation due to inaccurate TD timing errors [9]. An accurate timing determination for whole-heart coronary imaging can be difficult in such subjects, as the heart rate can vary from beat to beat over the duration of a long scan, or change significantly in the presence or absence of breath-holding. In addition to the per-patient and physiological variability, the optimal period quiescent period with minimal motion for each artery branch may vary by anatomical position [10,11]. Therefore, a single acquisition window may not necessarily be optimally timed for each coronary artery branch and region.

To overcome these challenges, multiphase acquisition strategies with non-Cartesian sampling schemes have been incorporated into coronary imaging. Stehning et al. have proposed a segmented 3D radial stack-of-stars acquisition approach that generated four consecutive volumes at each temporal sub-window [12], combining these sub-images to improve image quality. Additionally, several other studies have explored coronary acquisition over multiple temporal phases using k-space segmented strategies [6,13] or high-resolution 3D cine approaches [14–17] with radial or spiral sampling patterns. In all of the aforementioned multiphase 3D volumetric imaging approaches, the temporal window resolution is defined as the duration of each equally spaced temporal phase, and is not inherently designed for retrospective processing of any desired temporal subset within the acquisition window.

In this work, we propose an alternative acquisition scheme whose reconstruction window can be retrospectively selected by the operator in an interactive manner as any subset of the acquisition window (with a minimum window size of ~30 ms) and temporal resolution of a single readout (~4 ms). A radial stack-of-stars sampling based on golden angle (GA) radial interleaving is processed retrospectively to determine the period of minimal coronary artery motion within the acquired temporal window, which additionally allows tailoring of the optimal reconstruction window to individual branches and regions of each coronary artery.

**Materials and Methods**

**3D Stack-of-Stars K-space Segmented Sequence and its Optimal K-space Distribution**

Fig. 1 shows the schematics of the proposed sampling scheme. Our proposed approach uses an ECG-gated, k-space segmented 3D balances steady-state free precession (b-SSFP) enlarged acquisition window that utilizes a radial stack-of-stars sampling strategy with a customized rotation angle $\theta$. We acquire each $k_z$-plane one at a time in a centric order, instead of all $k_z$ slices for a particular radial spoke position at once prior to applying a rotation angle.

Typically, for radial imaging strategies that yield a single volume, radial spokes are acquired in such way that that the resulting $k_r$ plane is uniformly distributed; thus, if $n$ radial spokes are acquired on each plane, then a rotation angle that ensures uniform distribution satisfies the equation $\theta = j \cdot 180^\circ / n$, where $j$ is an integer such that the greatest common divisor of $j$ and $n$ is 1. $\theta$ is often set to $180^\circ / n$, (i.e. $j = 1$), where a sequential linear sweep is achieved through the plane.
An alternate scheme uses a golden angle (GA) rotation angle where $\theta$ is set to 111.246°. This scheme allows for retrospective reconstruction of the data with different temporal resolution, and has been utilized in dynamical imaging approaches with both radial and spiral trajectories in 2D for cardiac real-time imaging [18,19]; in 3D for the liver [20–22], brain [23], and the heart [24–26]; as well as in both 2D and 3D for imaging the vocal chords [27–31].

In this study, we seek to incorporate an acquisition strategy that enables a flexible reconstruction from any subset of the acquisition window. In a 3D stack-of-stars sequence acquired in an ECG-gated and k-space segmented scheme, a constant $\theta$ rotation is applied between each spoke, and this holds true between the final radial spoke of the acquisition window and the first spoke in the subsequent heartbeat. Therefore, the acquired k-space distribution becomes sensitive to the number of projection lines, or the Turbo Field Echoes (nTFE) in each heart beat as the acquired radial spokes over multiple heartbeats are temporally indexed modulo of nTFE. Furthermore, to achieve flexible retrospective selection and reconstruction, any subset of this temporal window must have the sorted k-space spokes to be distributed as uniformly as possible for minimal artifacts; and this is not necessarily achieved by using $\theta = \text{GA}$, as shown in Fig. 2. In this figure, the GA interleaving results in clustering of the spokes within a small temporal window when the acquisition window is set to 48 TFEs (left column). We also show that an improved distribution is feasible by using a different rotation angle (right column), and this hold true whether we use a reconstruction window of a single TR, or with 10 consecutive TRs, both containing spokes acquired over multiple repeated heartbeats ($n = 14$). In the subsequent sections, we will show how this is achieved.

The optimal k-space distribution over any temporal window subset can be examined as a function of 4 parameters; the rotation angle $\theta$ between consecutive radial spokes, the number of projection lines per heart beat (nTFE), the number of heartbeats per k$_z$ slice (nHB), and the number of projections in each heartbeat used for reconstruction, i.e. the reconstruction window (nRCW).

An empirical simulation-based approach was developed to assess and optimize the radial coverage over a flexible subset of the temporal window. In this study, the repeated number of heartbeats was fixed to nHB = 14, as this value corresponded to a 100% Nyquist sampling density over an 80 ms temporal window. In this optimization, we seek an optimization to enforce

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**Fig 1.** This diagram demonstrates the sampling scheme with the enlarged acquisition window: a) is the pulse sequence diagram that acquires an enlarged acquisition window. b) shows each k$_z$ plane being repeatedly acquired over a fixed number of heart beats. c) shows a centric k$_z$ ordering of the radial stacks to fill the 3D k-space volume.

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both even spacing of the radial spokes over a flexible range of the reconstruction window (nRCW), and to minimize the clustering of spokes in any sector of the reconstruction window. Accordingly, we define the objective function to minimize the following:

$$C(n_{TFE}, n_{HB}) = \arg \min_{\theta} \sum_{n_{RCW}=N_{min}}^{N_{max}} \| \Delta \Theta_{n_{RCW}}^{\theta, n_{TFE}} - \Delta \Theta_{n_{RCW}}^{n_{HB}, n_{RCW}} \|_2^2 + \lambda \cdot \max(\| \Theta_{n_{RCW}}^{\theta, n_{TFE}} - \Theta_{n_{RCW}}^{n_{HB}, n_{RCW}} \|)$$

where $\Theta_{n_{RCW}}^{\theta, n_{TFE}}$ is a vector corresponding to the angles of the nHB · nRCW radial spokes acquired within the reconstruction window over multiple heartbeats when a $\theta$ rotation is used between subsequent spokes, $\Theta_{n_{RCW}}^{n_{HB}, n_{RCW}}$ is the same vector of radial angles for the linear case with a $180^\circ/(n_{HB} \cdot n_{RCW})$ rotation between subsequent spokes, and $\Delta \Theta$ is the angle gap vector, which is the first differential of the $\Theta$ vector. We note that in this formulation, the reconstruction window is a subset of the acquisition window, hence $n_{RCW} \leq n_{TFE}$. Additionally, $\Theta_{n_{RCW}}^{n_{HB}, n_{RCW}}$ is invariant to the reconstruction window position within the acquisition window, and its derivation is shown in the Appendix. For the final objective function determination, all range of nRCW values was accounted for by taking the summation of nRCW between Nmin and Nmax.

In this optimization, the first (left) term enforces uniform spacing, while the second (right) one minimizes clustering. We note the right term is not rotation invariant, thus the first element of angle of $\Theta_{n_{RCW}}^{\theta, n_{TFE}}$ was chosen such that the $\Theta_{n_{RCW}}^{n_{HB}, n_{RCW}}$ was best aligned with $\Theta_{n_{RCW}}^{n_{HB}, n_{RCW}}$, by minimizing $\| \Theta_{n_{RCW}}^{\theta, n_{TFE}} - \Theta_{n_{RCW}}^{n_{HB}, n_{RCW}} \|_2^2$.

In this simulation, for each candidate $\theta$ at a fixed (nTFE, nHB) pair, the reconstruction window size nRCW was varied between Nmin = 7 (98 spokes; 44% sampling density) and Nmax = 25 (350 spokes; 156% sampling density), which corresponded to a temporal window of duration 32–115 ms at TR = 4.6 ms. The $\lambda$ value was empirically set to 0.7, which yielded the least clustering across all examined nTFEs.
In order to determine how well the GA/noptimal distributed the radial spokes among all possible input angles, an exhaustive numerical optimization using equation [1] was also performed. For this optimization, all $\theta$ values between [0.01, 179.99] were examined for the even nTFEs between 32 and 64 and nHB = 14, and were sorted according to their corresponding cost values from the objective function. The $\theta = \text{GA/noptimal}$ was expressed as a percentile rank among all possible angles.

Real-Time Visualization of Platform Design

An in-house accelerated reconstruction platform was developed in MATLAB for the visualization, and in C++ and NVIDIA CUDA for real-time non-Cartesian radial k-space gridding reconstruction. The raw k-space data and scanner parameters are used to automatically prepare the data for visualization on the platform, which consists of the following four steps: 1) temporal indexing and k-space look-up table generation, 2) coil compression, and 3) 3D volume data loading onto the CPU memory.

In step 1, the raw k-space data is first indexed into the following coordinate system $[k_x, k_y, k_z, t]$, and a look-up table is generated for rapid referencing of k-space from any desired temporal window. In step 2, a coil compression method proposed by Zhang et al. was employed to load the 3D k-space volume from 32 channels onto our standard workstation running the visualization platform; data was compressed from 32 to 8 compressed channels [32]. This compression method, which was originally proposed for Cartesian sampling, was feasible in our approach as the $k_z$-direction is fully sampled and thus was used as the spatially transformed direction to generate the hybrid k-space domain required for this approach. An additional Fourier transform in the $k_z$ direction is applied during this step, yielding an equivalent set of k-space values and their coordinates in the $[k_x, k_y, z, t]$ space, where $z$ corresponds to the axial slice. In step 3, the processed data is finally mounted onto the CPU memory for visualization. Mounted data is stored as temporally indexed and channel-compressed k-space values, and coordinate position. The operator uses the GUI to select the $z$-slice and temporal window, and the developed GPU reconstruction engine performs the gridding operation of the selected non-Cartesian k-space data, and outputs the reconstructed image for immediate display onto the GUI panel. The gridding operation is parallelized and performed on a per-point basis, where gridded data from each GPU kernel is accumulated onto the final Cartesian grid using a dedicated GPU atomic addition operation [33]. Image reconstruction is performed as a sequence of four sub-steps: a) zero-padding, b) inverse FFT, c) Cropping to the correct FOV, and d) Roll-off correction. After completing these steps, the reconstructed image is mounted back to the CPU memory, and is returned to the MATLAB GUI.

For the GPU-gridding step, a fixed Kaiser-Bessel gridding kernel size of $W = 5.5$ and a default oversampling factor of 1.25 were chosen in this study, for which the corresponding gridding shape $\beta$ parameter set to 9.9981 per approximation method proposed by Beatty et al. [34].

The performance of each GPU reconstruction substep was evaluated using different input k-space parameters typically acquired in high-resolution coronary MRI, and the measured times were averaged over 100 operations. For comparison, the reconstruction was also performed and evaluated using the NUFFT package [35] provided in MATLAB, which requires an initial preparation step.

In Vivo Imaging

All scans were performed on a 1.5T Philips Achieva MR system using a 32-channel cardiac array under an institutional review board (IRB) approved protocol (IRB No. 2001-000793;
approved by the Committee on Clinical Investigation of Beth Israel Deaconess Medical Center, Boston, MA). Written informed consents were obtained from all subjects prior to imaging, allowing use of all acquired MR data in our studies. Eight healthy subjects (5 male, mean age = 37 ± 18 years) were scanned. The ECG-gated 3D radial stack-of-stars SSFP coronary MRI was performed approximately 2 minutes post-contrast injection (0.2 mmol/kg, Gadobenate Dimeglumine) with the following parameters: TR = 4.6 ms, TE = 1.8 ms, FA = 100°, FOV = 300 x 300 x 80–120 mm³. Spatial resolution = 1.3 x 1.3 x 2.0 mm³. The subject-specific gating delay was determined from a breath-held 2D cine, and was adjusted 50 ms prior to the determined quiescent period. The prolonged acquisition window size was set to 220 ms for nTFE = 48 for subjects with a heart rate (HR) < 65, while a smaller window of 166 ms (nTFE = 36) was used for HR that exceeded 65. Accordingly, a custom radial angle of θ = 15.89 (nTFE = 48) and θ = 9.27 (nTFE = 36) was used (from Table 1), respectively. The coronary MRI sequence used a T2Prep pulse with an echo time of 50 ms, a diaphragmatic navigator with a 7 mm gating window with slice-tracking (tracking factor = 0.6), fat saturation, and a single half-alpha ramp up pulse, which was selected to minimize the time between all magnetization preparation pulses with volumetric imaging over the prolonged acquisition window to maximize the fat saturation over the enlarged acquisition window. Physiological parameters were extracted from the cine and the coronary imaging sequences. The optimal temporal reconstruction window for the right coronary artery (RCA) and the left anterior descending artery (LAD) were retrospectively determined manually by an experienced cardiologist using the developed visualization tool on a clinical workstation. The interactive assessment for the selection of the optimal coronary vessel visualization allowed selection of: 1) z-slice in the image display for scrolling through the 3D volume, 2) zoom, 3) manipulation of the window level, and 4) adjustment of both the start and end of the temporal window. A typical workflow was as follows: the standard 120 ms

| nTFE | GA/n optimal with λ = 0.7 | Percentile (%) rank per 18000 |
|------|-------------------------|-----------------------------|
| 32   | 55.62 (GA/2)            | 5                           |
| 34   | 13.91 (GA/8)            | 10                          |
| 36   | 9.27 (GA/12)            | 0.5                         |
| 38   | 55.62 (GA/2)            | 12                          |
| 40   | 7.42 (GA/15)            | 3                           |
| 42   | 37.08 (GA/3)            | 3                           |
| 44   | 18.54 (GA/6)            | 12                          |
| 46   | 13.91 (GA/8)            | 10                          |
| 48   | 15.89 (GA/7)            | 5                           |
| 50   | 13.91 (GA/8)            | 3                           |
| 52   | 10.11 (GA/5)            | 7                           |
| 54   | 8.56 (GA/13)            | 0.01                        |
| 56   | 10.11 (GA/11)           | 3                           |
| 58   | 7.95 (GA/14)            | 5                           |
| 60   | 13.91 (GA/8)            | 7                           |
| 62   | 12.36 (GA/8)            | 1                           |
| 64   | 9.27 (GA/2)             | 5                           |

In Bold—nTFE (36 = 166 ms window, 48 = 220 ms window) and angles used in this study.

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window at the TD was first examined; the temporal window was then adjusted by the operator to yield improved visualization of the examined artery branch region. The window that yielded a notably better vessel region was then examined for its sharpness.

All image reconstruction at the desired temporal window was performed offline on the in-house visualization platform described previously. For assessment of the coronary vessel sharpness, 3D volumes were reconstructed to 0.65 x 0.65 x 2.0 mm³ resolution with zero-filled interpolation using the platform. Vessel sharpness was measured using the SoapBubble Tool [36], a semi-automated two-step approach that first performs software-assisted delineation of a multi-planar reformat of the visualized artery region, followed by an fully automated sharpness calculation on the reformatted plane using Deriche edge detection [37]. Sharpness measurements were measured by an independent operator using the SoapBubble tool. The measurements were repeated twice for intra-observer variability. This tool was also used to measure the luminal signal-to-noise ratio (SNR), surrounding myocardium SNR, and the lumen-to-myocardium contrast-to-noise ratio (CNR), where the lumen ROI was prescribed on the targeted artery region after the multi-planar reformat, and the myocardium ROI was prescribed on the myocardial tissue adjacent to the lumen ROI. Background air was used for noise measurement.

All statistics were performed using Microsoft Excel (Microsoft, Redmond, WA). The Student’s t-test was used for the comparison of all measurements.

Results
Simulation of K-space distribution optimization

Fig. 3 shows the best GA/n reconstructions for nTFEs between 32 and 64 (in step sizes of 4) when λ was set to 0.7; which was manually determined to provide a de-clustered selection of the spokes in all tested cases for all even nTFEs between 32 and 64.

Compared to other GA/n cases, these cases yielded both evenly distributed and de-clustered results after the proposed optimization. The look-up table was generated for all examined nTFEs = [32, 34, 36, ..., 64], and is reported in Table 1, along with the percentile rank out of 18000. These cases were within the top 12 percentile. A look-up-table-derived GA/n angle reported in Table 1 provided a reasonable coverage of the kₓ-kᵧ planes with good image quality even under undersampled conditions, as the proposed objective function employed in this study accounts for both the optimal angle gap spacing and clustering of radial spokes within a sector of the kₓ-kᵧ plane. The reported rotation angles offer reasonable approximations for a range of nTFEs when nHB is fixed to 14, and images were acquired using the reported table values for nTFE = 36 and 48 in the subsequent sections.

Visualization Platform Evaluation

Fig. 4A) shows the front-end GUI display developed for this software, and 4B) shows the block diagram of the reconstruction steps performed for an operator-selected temporal window,
including the timing calculations for each reconstruction operation from the operator instructions to visualization. The total time required for the combined transfer of input and output data between MATLAB and GPU was constant at 0.18 ± 0.01 seconds. Table 2 shows the average computation times at different reconstruction window sizes. The GPU implementation required a fixed 0.2 seconds for MATLAB-GPU interaction, 0.2–0.35 seconds for gridding—for a total of 0.4–0.55 seconds for visualization of the desired 2D slice from the operator-specified temporal window and slice. For CPU gridding with NUFFT, an initial preparation step of the appropriate k-space look-up table took 1.5–10.5 seconds before performing the gridding and visualization steps in a single step (1.2–1.8 seconds), which required ~3 times than the GPU implementation.

Significant speed-up in total processing time is observed compared to the NUFFT-based preparation implemented in MATLAB, which required as much as 13 seconds (10.5 for prep + 1.8 for gridding), and is therefore unsuitable for a flexible and retrospective reconstruction workflow for interactive visualization.

In Vivo Imaging

All scans were completed successfully. The average navigator efficiency for the whole-heart coronary imaging across the eight subjects was 41 ± 11%. Table 3 reports the normalized vessel sharpness scores from the operator assessments, as well as its difference between the score

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**Diagram:**

**Table 2. Average Processing Times GPU vs CPU Methods.**

| Test Methodology                      | Total Computation Time (seconds) | GPU-TRANSFER | GPU-GRID | GPU-TOTAL | CPU-PREP | CPU-GRID TOTAL |
|---------------------------------------|---------------------------------|--------------|----------|-----------|----------|----------------|
| 1.3 mm resolution Recon               |                                 |              |          |           |          |                |
| RecWin Size = 7; (464 × 98 × 8) = 363776 pts | 0.18 ± 0.01                  | 0.19 ± 0.02  | 0.37 ± 0.04 | 1.5 ± 0.1 | 1.2 ± 0.1 |
| RecWin Size = 16; (464 × 224 × 8) = 831488 pts | 0.18 ± 0.01                  | 0.23 ± 0.02  | 0.41 ± 0.04 | 3.5 ± 0.4 | 1.4 ± 0.1 |
| RecWin Size = 48; (464 × 672 × 8) = 2494464 pts | 0.18 ± 0.01                  | 0.35 ± 0.03  | 0.53 ± 0.04 | 10.5 ± 1.1 | 1.8 ± 0.2 |
| Submillimeter resolution Recon        |                                 |              |          |           |          |                |
| RecWin Size = 48; (600 × 864 × 8) = 4147200 pts | 0.19 ± 0.01                  | 0.78 ± 0.08  | 0.96 ± 0.10 | 17.9 ± 1.8 | 2.8 ± 0.29 |

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| RecWin Size = 48; (600 × 864 × 8) = 4147200 pts | 0.19 ± 0.01                  | 0.78 ± 0.08  | 0.96 ± 0.10 | 17.9 ± 1.8 | 2.8 ± 0.29 |
Table 3. Vessel Wall Sharpness Scores.

| Subj. | RCA (Root) | RCA (Mid) | RCA (Distal) | LAD |
|-------|------------|-----------|--------------|-----|
|       | [Standard Window Score; Manually Derived Window Score] (%) | Repeated measurement in next row. |
| 1     | 0.37; 0.43 | ND; 0.36* | ND; 0.22* | 0.29; 0.30 |
|       | 0.37; 0.43 | ND; 0.37* | ND; 0.24* | 0.30; 0.32 |
| 2     | 0.37; SAME | 0.37; 0.42 | 0.24; 0.30 | 0.30; SAME |
|       | 0.33; SAME | 0.38; 0.39 | 0.24; 0.30 | 0.27; SAME |
| 3     | 0.34; 0.39 | ND; 0.38* | ND; 0.31* | 0.30; 0.40 |
|       | 0.32; 0.37 | ND; 0.39* | ND; 0.30* | 0.32; 0.42 |
| 4     | 0.37; 0.39 | 0.31; 0.31 | ND; ND | 0.42; 0.39 |
|       | 0.31; 0.38 | 0.28; 0.29 | ND; ND | 0.40; 0.38 |
| 5     | 0.36; 0.40 | ND; 0.44* | 0.21; 0.29 | 0.27; 0.37 |
|       | 0.35; 0.38 | ND; 0.42* | 0.22; 0.28 | 0.27; 0.38 |
| 6     | 0.42; SAME | 0.45; 0.49 | 0.29; SAME | 0.41; SAME |
|       | 0.43; SAME | 0.44; 0.46 | 0.30; SAME | 0.41; SAME |
| 7     | 0.28; SAME | — | — | ND; ND |
|       | 0.28; SAME | — | — | ND; ND |
| 8     | 0.52; SAME | 0.41; 0.47 | ND; ND | 0.42; SAME |
|       | 0.53; SAME | 0.42; 0.49 | ND; ND | 0.44; SAME |

(*)—Vessel was non-descript with the standard, but visualized with the manually window.
(SAME)—The standard window yielded good image quality after visual assessment.
(ND)—Vessel was non-descript.
(—)—Subject 7 exhibited having a small rudimentary right with left-dominance, thus the RCA-mid to distal end was not present. BOLD—reconstructions are shown in Fig. 5.

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from the standard window, while Table 4 reports the adjusted temporal window selected by each operator. Table 5 summarizes the SNR and CNR measurements.

Fig. 5 demonstrates examples of motion corrected vessel at the root of the RCA, and a distal RCA vessel recovered by the operator using the proposed approach. Fig. 6 shows operator’s workflow to optimize the visualization at the mid-RCA region by manipulating the reconstruction’s temporal window in an interactive manner. In this case, the operator was able to fully recover the RCA that was non-descript with the initial window (left) by gradually adjusting the reconstruction window to that yielding optimal visualization with minimal blurring motion.

Out of 30 examined artery regions, 9 were sufficiently well-visualized using the standard window, and operator interaction was performed in 19 regions. Of these, 14 regions (4 each in RCA-root, mid, and LAD, and 2 in RCA-distal) yielded a visible vessel with the standard window that was manipulated by the operator for better visualization, while 5 were fully visualized using the proposed method when it was non-descript for the standard window (eg. Fig. 5–2, right). The lumen-to-myocardium CNR increased by 30±25% (n = 14; p < 0.005) after manual adjustment compared to the vessels depicted in the standard 120 ms window reconstructions. Three regions were not examined due to any visible artery region throughout the entire acquisition window. For the RCA root and LAD, the average sharpness scores were (manually selected vs standard 120 ms windows): RCA (Root) 0.40 ± 0.06 vs 0.38±0.07; (LAD) 0.36±0.06 vs 0.34±0.07. Repeated measurements yielded: RCA (Root) 0.39 ± 0.07 vs 0.37±0.08; (LAD) 0.37±0.06 vs 0.34±0.07. The vessel sharpness assessments were statistically significant and
improved (P < 0.0005 for both measurements) in the 14 regions where the vessel region was visualized in the standard window, but necessitated interactive adjustment by the operator. The average sharpness score improvements in these regions were both 0.05 ± 0.04 for measurements 1 and 2.

Table 4. Manually Selected vs Standard Temporal Window.

| Subj. | Standard Window | RCA (Root) | RCA (Mid) | RCA (Distal) | LAD |
|-------|-----------------|------------|-----------|--------------|-----|
|       | [Start of Window, End of Window.] (milliseconds) with respect to ECG trigger delay. |
| 1     | [620, 740]      | [583, 670] | [579, 630] | [592, 651]   | [611, 717] |
| 2     | [620, 740]      | SAME       | [575, 639] | [583, 657]   | SAME          |
| 3     | [510, 630]      | [497, 571] | [492, 556] | [469, 543]   | [573, 660]    |
| 4     | [600, 720]      | [559, 679] | [559, 679] | ND           | [609, 715]    |
| 5     | [550, 670]      | [528, 625] | [509, 564] | [545, 632]   | [505, 597]    |
| 6     | [580, 700]      | SAME       | [566, 658] | SAME         | SAME          |
| 7     | [670, 890]      | SAME       | —          | —            | ND            |
| 8     | [650, 720]      | SAME       | [650, 714] | ND           | SAME          |

(SAME)—The standard window yielded good image quality after visual assessment. (ND)—Vessel was non-descript. 
——Subject 7 exhibited having a small rudimentary right with left-dominance, thus the RCA-mid to distal end was not present.

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Table 5. SNR and CNR Measurements.

| Subj. | RCA (Root) | RCA (Mid) | RCA (Distal) | LAD |
|-------|------------|-----------|--------------|-----|
|       | Myocardium SNR; Lumen SNR; Lumen-to-Myocardium CNR |
| 1     | (Standard) | 8.6; 22.0; 13.4 | ND* | ND* | 5.6; 13.9; 8.3 |
|       | (Adjusted) | 6.7; 23.8; 17.1 | 5.6; 21.8; 16.2 | 7.3; 14; 6.7 | 5.4; 17.6; 12.2 |
| 2     | (Standard) | 7.2; 16.7; 9.6 | 10.2; 21.9; 11.7 | 8.4; 20.8; 12.4 | 5.7; 15.2; 9.5 |
|       | (Adjusted) | SAME       | 4.4; 18.4; 13.9 | 6.3; 16.5; 10.2 | SAME          |
| 3     | (Standard) | 4.5; 15.1; 10.6 | ND* | ND* | 5.6; 17.9; 12.3 |
|       | (Adjusted) | 5.1; 20.8; 15.7 | 7.1; 17.1; 10.0 | 7.2; 16.4; 9.2 | 6.0; 20.6; 14.6 |
| 4     | (Standard) | 10.1; 23.8; 13.7 | 9.0; 27.5; 18.5 | ND | 6.2; 22.1; 15.9 |
|       | (Adjusted) | 7.6; 28.6; 21.0 | 12.9; 30.0; 17.1 | ND | 6.5; 24.7; 18.3 |
| 5     | (Standard) | 13.0; 26.4; 13.3 | ND* | ND* | 7.0; 14.9; 7.9 | 4.1; 11.8; 7.7 |
|       | (Adjusted) | 6.6; 22.2; 15.6 | 4.5; 15.7; 11.3 | 6.7; 16.8; 10.1 | 6.2; 17.6; 11.4 |
| 6     | (Standard) | 6.8; 36.3; 29.5 | 12.8; 38.3; 25.5 | 7.7; 19.7; 12.1 | 5.5; 17.8; 12.3 |
|       | (Adjusted) | SAME       | 15.9; 56.2; 40.3 | SAME | SAME          |
| 7     | (Standard) | 9.4; 32.8; 23.5 | — | — | ND            |
|       | (Adjusted) | SAME       | — | — | ND            |
| 8     | (Standard) | 10.5; 41.1; 30.6 | 13.5; 34.3; 20.8 | ND | 8.8; 34.8; 26.0 |
|       | (Adjusted) | SAME       | 10.2; 46.0; 35.8 | ND | SAME          |

(*)—Vessel was non-descript with the standard, but visualized with the manually window.
(SAME)—The standard window yielded good image quality after visual assessment.
(ND)—Vessel was non-descript.
——Subject 7 exhibited having a small rudimentary right with left-dominance, thus the RCA-mid to distal end was not present.
BOLD—Lumen-to-Myocardium CNR.

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Discussion

In this study, we have demonstrated the feasibility of acquiring an enlarged acquisition window using a 3D radial stack-of-stars sampling scheme with a GA-derived customized angle for optimal k-space distribution to enable reconstruction of any flexible temporal subset of the acquisition window in a retrospective manner, allowing tailored reconstruction for different coronary artery branches and regions within the acquired 3D volume. An optimized rotation angle for the 3D k-space segmented sequence was derived from the GA radial sampling trajectory, allowing a flexible reconstruction with a minimal subset of at least 7 k-space spokes per heart-beat and a temporal resolution of 32 ms. A simulation was performed to report the optimal rotation angle for a fixed nTEF and nHB pair. A look-up-table was derived for the optimal GA/n angle as reported in Table 1, which provides a reasonable coverage of the kx-ky planes with good image quality even under undersampled conditions accounting for both the optimal angle gap spacing and clustering of radial spokes within a sector of the kx-ky plane. For rapid and interactive visualization of the data, a GPU-based reconstruction engine was combined with the GUI developed in MATLAB to enable an interactive assessment. When compared with the MATLAB-based NUFFT reconstruction method that is employed for offline non-Cartesian methods, the GPU-based radial reconstruction requires no dedicated preparation, and is

![Image](image1.png)

**Fig 5.** Examples of the impact of interactive vs. fixed selection of the trigger delay and window for visualization of RCA in two different subjects. In Subject #1, using the interactively selected window mitigates the temporal blurring seen in RCA images reconstructed using fixed trigger delay and window. In Subject #2, the distal region of RCA can be fully recovered by interactively adjusting the trigger delay and window size.

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![Image](image2.png)

**Fig 6.** The operator’s workflow for optimal visualization of the mid-RCA region. In A) the images reconstructed from the fixed pre-determined window of 120 ms and trigger delay of 620 ms. The RCA cannot be visualized presumably due to cardiac motion. In B), the reconstruction window was adjusted to 87 ms with a TD of 593 ms and in C) a 65 ms reconstruction window with a TD of 602 ms, which results in a motion-free clear visualization of the RCA. The operator scrolled through multiple slices (not shown) to verify the optimal reconstruction for the RCA.

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therefore more suitable for on-the-fly visualization within 1 second even under computationally heavy conditions, which matches previously reported ranges [38]. While the gridding reconstruction using the NUFFT package in MATLAB performs sufficiently well for most offline radial gridding reconstructions at our site that are acquired with lower spatial resolutions, a more robust GPU gridding implementation enabled the real-time visualization approach required for the clinical assessment tool in this study.

The strength of the proposed combined acquisition and reconstruction approach lies in its ability to enable visualization of any slice of the whole-heart 3D volume in real-time from the time-sorted k-space data. The additional consideration of the temporal dimension in the reconstruction step enables a retrospective examination of any flexible subset of the enlarged acquisition window. Further scan time reduction can be gained by undersampling the radial stack-of-stars sequence by incorporating iterative reconstruction approaches such as parallel imaging [39] or compressed sensing [40].

There are several limitations to this study. First, the contrast mechanism throughout the enlarged acquisition window is not optimized in this study with respect to: a) fat suppression, and b) contrast optimization throughout the enlarged acquisition window. For fat suppression, we employ a conventional fat saturation preparation prior to the radial SSFP acquisition of the enlarged acquisition window. Thus, this approach can suffer from fat signal recovery, and was subsequently used with contrast enhancement. Combination with fat water separation methods, such as Dixon techniques [41], or using water selective imaging approaches, such as ATR [42] or wideband SSFP [43], can potentially address this at the expense of prolonged scan time. Finally, in this study, we only examined the RCA and the LAD in healthy subjects. A more detailed examination of our technique in 1) a larger study population, 2) in patients with suspected or known coronary artery disease, and 3) assessing all coronary vessels are warranted.

In conclusion, this study demonstrates the feasibility of a whole-heart coronary imaging approach with an enlarged acquisition window by which the trigger delay and acquisition window can be adjusted interactively to tailor the reconstruction of different coronary branch regions. The sub-optimal distribution of k-space spokes in a segmented radial stack-of-stars acquisition with different acquisition and reconstruction windows is first addressed by the proposed optimization approach, and a GUI tool is developed to enable a flexible GPU-based offline radial k-space reconstruction for interactive assessment of any desired temporal window from the 4D coronary k-space data.

**Appendix**

Let $\Theta_0$ denote $\Theta_{\theta,n\text{TFE},n\text{HB},n\text{RCW}}$ for a fixed ($\theta$, $n\text{TFE}$, $n\text{HB}$, $n\text{RCW}$) in a reconstruction window between $[0, n\text{RCW}]$. Then, any subsequent reconstruction window with an offset of $k$ TRs can be denoted as $[k, k + n\text{RCW}]$, denoted by $\Theta_k$. As the rotation angle is a fixed constant $\theta$ during the acquisition, $\Theta_k = (\Theta_0 + k\theta) \mod \pi$, an element-wise addition, which is equivalent to a global rotation of the radial spokes by $k\theta$. Subsequently, it can be seen that the angle gap $\Delta \Theta_k = \Delta \Theta_0$, as this gap does not change over a global rotation. Hence, $\Delta \Theta_{\theta,n\text{TFE},n\text{HB},n\text{RCW}}$ is invariant to the position of the reconstruction window, and can be simplified as a function that depends solely on $n\text{RCW}$ regardless of its relative position within the acquisition window.

**Author Contributions**

Conceived and designed the experiments: KK RN. Performed the experiments: KK MF. Analyzed the data: KK MF. Contributed reagents/materials/analysis tools: KK SR MA. Wrote the paper: KK SR MA RN.
References

1. Weber O, Martin A, Higgins C (2003) Whole-heart steady-state free precession coronary artery magnetic resonance angiography. Magnetic Resonance in Medicine 50: 1223–1228. PMID: 14648570

2. Sakuma H, Ichikawa Y, Suzawa N, Hirano T, Makino K, et al. (2005) Assessment of Coronary Arteries with Total Study Time of Less than 30 Minutes by Using Whole-Heart Coronary MR Angiography. Radiology 237: 316–321. PMID: 16126921

3. Santos JM, Cunningham CH, Lustig M, Hargreaves BA, Hu BS, et al. (2006) Single breath-hold whole-heart MRA using variable-density spirals at 3T. Magn Reson Med 55: 371–379. PMID: 16408262

4. Bi X, Carr JC, Li D (2007) Whole-heart coronary magnetic resonance angiography at 3 Tesla in 5 minutes with slow infusion of Gd-BOPTA, a high-relaxivity clinical contrast agent. Magn Reson Med 58: 1–7. PMID: 17659629

5. Yonezawa M, Nagata M, Kitagawa K, Kato S, Yoon Y, et al. (2013) Quantitative Analysis of 1.5-T Whole-Heart Coronary MR Angiograms Obtained with 32-Channel Cardiac Coils: A Comparison with Conventional Quantitative Coronary Angiography. Radiology: 122491.

6. Wu HH, Gurney PT, Hu BS, Nishimura DG, McConnell MV (2013) Free-breathing multiphase whole-heart coronary MR angiography using image-based navigators and three-dimensional cones imaging. Magn Reson Med 69: 1083–1093. doi:10.1002/mrm.24346 PMID: 22648856

7. Piccini D, Monney P, Sierro C, Coppo S, Bonanno G, et al. (2014) Respiratory Self-navigated Postcontrast Whole-Heart Coronary MR Angiography: Initial Experience in Patients. Radiology 270: 378–386. doi:10.1148/radiol.13132045 PMID: 24471387

8. Bi X, Park J, Deshpande V, Simonetti O, Laub G, et al. (2007) Reduction of flow- and eddy-currents-induced image artifacts in coronary magnetic resonance angiography using a linear centric-encoding SSFP sequence. Magnetic Resonance Imaging 25: 1136–1147. PMID: 17905246

9. Shechter G, Resar JR, McVeigh ER (2006) Displacement and velocity of the coronary arteries: cardiac and respiratory motion. IEEE Trans Med Imaging 25: 369–375. PMID: 16524092

10. Liu G, Qi XL, Robert N, Dick AJ, Wright GA (2012) Ultrasound-guided identification of cardiac imaging windows. Med Phys 39: 3009–3018. doi: 10.1118/1.4711757 PMID: 22755685

11. Stehning C, Börnert P, Nehrke K, Dönser O (2005) Free breathing 3D balanced FFE coronary magnetic resonance angiography with prolonged cardiac acquisition windows and intra-RR motion correction. Magnetic Resonance in Medicine 53: 719–723. PMID: 15723401

12. Warmuth C, Schnorr J, Kaufels N, Wagner S, Pilgrimm H, et al. (2007) Whole-heart coronary magnetic resonance angiography: contrast-enhanced high-resolution, time-resolved 3D imaging. Invest Radiol 42: 550–557. PMID: 17620937

13. Kressler B, Spincemaille P, Prince MR, Wang Y (2006) Reduction of reconstruction time for time-resolved spiral 3D contrast-enhanced magnetic resonance angiography using parallel computing. Magn Reson Med 56: 704–708. PMID: 16892189

14. Lai P, Larson AC, Park J, Larson AC, Zhang Q, Simonetti O, et al. (2005) Contrast-enhanced 4D radial coronary artery imaging at 3.0T within a single breath-hold. Magn Reson Med 54: 470–475. PMID: 16032681

15. Feng L, Grimm R, Tobias Block K, Chandarana H, Kim S, et al. (2013) Golden-angle radial sparse parallel MRI: Combination of compressed sensing, parallel imaging, and golden-angle radial sampling for fast and flexible dynamic volumetric MRI. Magn Reson Med.
22. Agrawal MD, Spincemaille P, Mennitt KW, Xu B, Wang Y, et al. (2013) Improved hepatic arterial phase MRI with 3-second temporal resolution. J Magn Reson Imaging 37: 1129–1136. doi:10.1002/jmri.23920 PMID: 23197440

23. Song HK, Yan L, Smith RX, Xue Y, Rapacchi S, et al. (2013) Noncontrast enhanced four-dimensional dynamic MRA with golden angle radial acquisition and k-space weighted image contrast (KWIC) reconstruction. Magn Reson Med.

24. Liu J, Spincemaille P, Codella NC, Nguyen TD, Prince MR, et al. (2010) Respiratory and cardiac self-gated free-breathing cardiac CINE imaging with multiecho 3D hybrid radial SSFP acquisition. Magn Reson Med 63: 1230–1237. doi:10.1002/mrm.22306 PMID: 20432294

25. Lai P, Huang F, Li Y, Nielles-Vallespin S, Bi X, et al. (2010) Contrast-kinetics-resolved whole-heart coronary MRA using 3DPR. Magn Reson Med 63: 970–978. doi:10.1002/mrm.22246 PMID: 20373398

26. Paul J, Divkovic E, Wundrak S, Bernhardt P, Rottbauer W, et al. (2014) High-resolution respiratory self-gated golden angle cardiac MRI: Comparison of self-gating methods in combination with k-t SPARSE SENSE. Magn Reson Med.

27. Kim YC, Hayes CE, Narayanan SS, Nayak KS (2011) Novel 16-channel receive coil array for accelerated upper airway MRI at 3 Tesla. Magn Reson Med 65: 1711–1717. doi:10.1002/mrm.22742 PMID: 21590804

28. Kim YC, Narayanan SS, Nayak KS (2011) Flexible retrospective selection of temporal resolution in real-time speech MRI using a golden-ratio spiral view order. Magn Reson Med 65: 1365–1371. doi: 10.1002/mrm.21500262

29. Kim YC, Proctor MI, Narayanan SS, Nayak KS (2012) Improved imaging of lingual articulation using real-time multislice MRI. J Magn Reson Imaging. 35: 943–948. doi:10.1002/jmri.23510 PMID: 22127935

30. Kim YC, Lebel RM, Wu Z, Ward SL, Kho MC, et al. (2013) Real-time 3D magnetic resonance imaging of the pharyngeal airway in sleep apnea. Magn Reson Med.

31. Zhu Y, Kim YC, Proctor MI, Narayanan SS, Nayak KS (2013) Dynamic 3-D visualization of vocal tract shaping during speech. IEEE Trans Med Imaging 32: 838–848. doi:10.1109/TMI.2012.2230017 PMID: 23204279

32. Zhang T, Pauly JM, Vasanawala SS, Lustig M (2013) Coil compression for accelerated imaging with Cartesian sampling. Magn Reson Med 69: 571–582. doi:10.1002/mrm.22467 PMID: 22488589

33. Nam S, Akçakaya M, Basha T, Stehning C, Manning WJ, et al. (2012) Compressed sensing reconstruction for whole-heart imaging with 3D radial trajectories: A graphics processing unit implementation. Magn Reson Med.

34. Beatty PJ, Nishimura DG, Pauly JM (2005) Rapid gridding reconstruction with a minimal oversampling ratio. IEEE Trans Med Imaging 24: 799–808. PMID: 15959939

35. Fessler JA (2007) On NUFFT-based gridding for non-Cartesian MRI. J Magn Reson 188: 191–195. PMID: 17689121

36. Etienne A, Botnar RM, Van Muyswinkel AM, Boesiger P, Manning WJ, et al. (2002) “Soap-Bubble” visualization and quantitative analysis of 3D coronary magnetic resonance angiograms. Magn Reson Med 48: 658–666. PMID: 12353283

37. Deriche R (1990) Fast algorithms for low-level vision. IEEE Transactions on Pattern Analysis and Machine Intelligence 12.

38. Nielsen J (1999) User interface directions for the Web. Commun ACM 42: 65–72.

39. Pruessmann KP, Weiger M, Börnert P, Boesiger P (2001) Advances in sensitivity encoding with arbitrary k-space trajectories. Magn Reson Med 46: 638–651. PMID: 11590639

40. Lustig M, Donoho D, Pauly J (2007) Sparse MRI: The application of compressed sensing for rapid MR imaging. Magnetic Resonance in Medicine 58: 1182–1195. PMID: 17969013

41. Börnert P, Koken P, Nehrke K, Eggers H, Ostendorf P (2014) Water/fat-resolved whole-heart Dixon coronary MRA: an initial comparison. Magn Reson Med 71: 156–163. doi: 10.1002/mrm.24648 PMID: 23400853

42. Leupold J, Hennig J, Scheffler K (2006) Alternating repetition time balanced steady state free precession. Magn Reson Med 55: 557–565. PMID: 16447171

43. Nayak KS, Lee HL, Hargeaves BA, Hu BS (2007) Wideband SSFP: alternating repetition time balanced steady state free precession with increased band spacing. Magn Reson Med 58: 931–938 PMID: 17969129