

Automated Cranial 4D Flow MRI Analysis

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Synopsis

To develop and validate a software tool for robust cerebrovascular analysis of 4D flow MRI data that includes:

- An interactive visual interface
- Quick and robust automated and reproducible hemodynamic analysis

Background

- 4D flow MRI acquires in-vivo, time-resolved, volumetric data of velocity vectors.
 - Allows for qualitative and quantitative analysis of anatomy and hemodynamics
 - Post-processing can be time consuming
- A previously-established cranial 4D flow centerline processing scheme (CPS)¹ allowed:
 - Repeatable quantitative analysis
 - Automated segmentation
 - Quick post processing times (~15 min)
 - Analysis of +1000 cranial cases

Purpose: Develop a quantitative velocity tool (QVT) that improves 4D flow post-processing:

1. Fully automated quantitative flow computations for all vessels (~3000 locations)
 - Anatomical: vessel segmentation
 - Functional: blood velocity, flow, pulsatility index, etc.
2. Interactive 3D user interface
 - Visualization of color-coded flow parameters
 - Easy vessel selection
3. Reduce memory requirements and post-processing times (~5 min)

Methods

- An in-vivo study was performed to assess CPS and QVT segmentation performances, interobserver variabilities, vessel selection, and post processing times.
- Ten 4D flow brain scans were acquired on healthy volunteers.
 - PC-VIPR² acquisition: field strength = 3T; volume = 22 cm³; spatial res. = 0.69 mm²; V_{ENC} = 80 cm/s; scan time = 7 min.; retrospective cardiac gating (20 frames).
- The QVT graphical interface (Figure 1) was developed in MATLAB18b (Mathworks, MA).

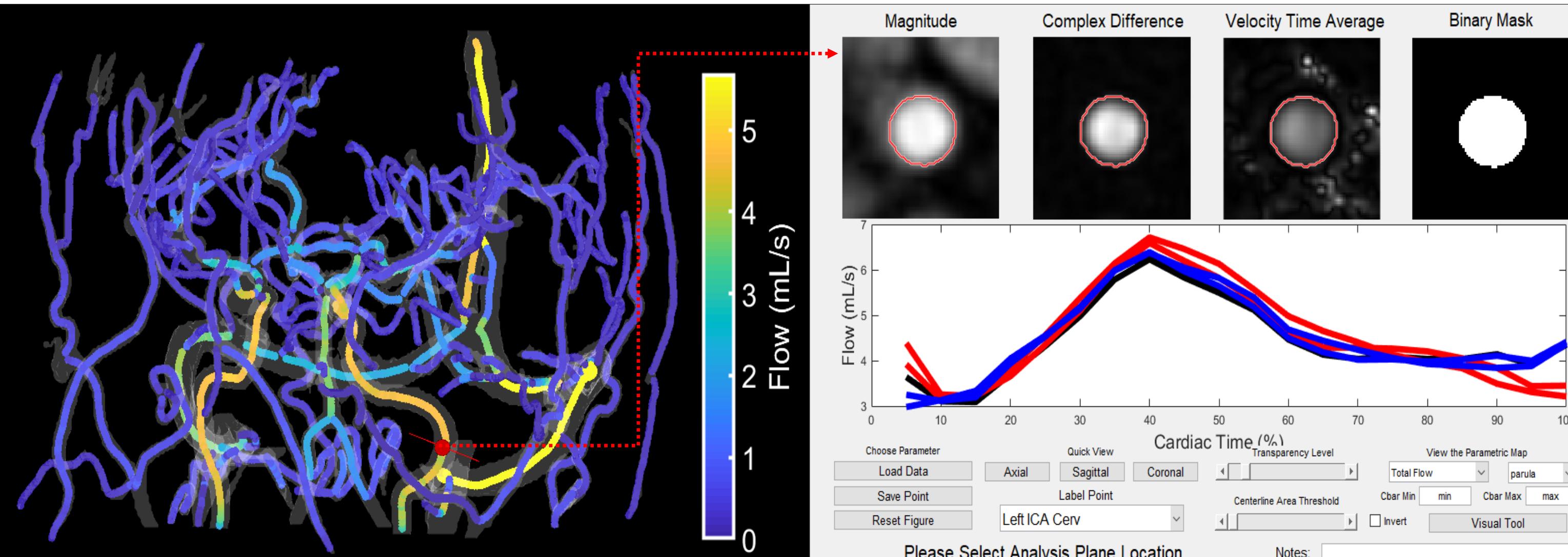


Figure 1: 4D flow post-processing tool with 3D interactive graphical user interface. As the user moves the cursor, the cross-sectional images and flow waveforms are updated in real-time.

- A threshold-based vessel segmentation algorithm was developed (Figure 2).
- Automated calculations of blood flow parameters were completed across the vascular network and color-coded in the 3D interface (Figure 3).

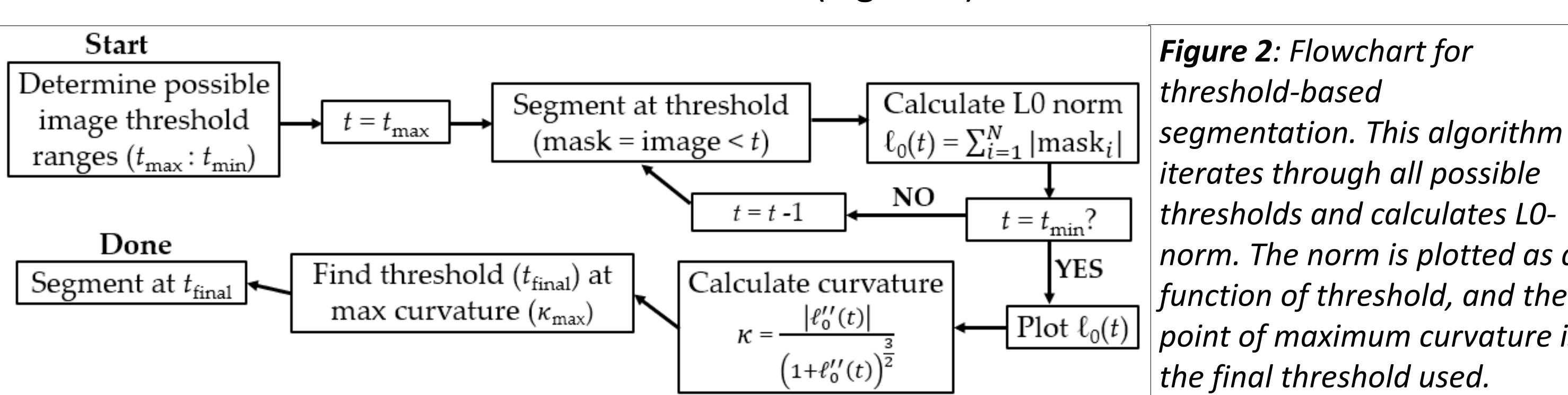


Figure 2: Flowchart for threshold-based segmentation. This algorithm iterates through all possible thresholds and calculates L0-norm. The norm is plotted as a function of threshold, and the point of maximum curvature is the final threshold used.

- Flow analysis was completed in 13 locations: internal carotid arteries (ICA), basilar artery (BA), middle cerebral arteries (MCA), posterior cerebral arteries (PCA), straight sinus (SS), superior sagittal sinus (SSS), and transverse sinuses (TS).
- Automated vessel segmentation methods were evaluated against a manual segmentation:
 - K-means segmentation (CPS) and threshold-based segmentation (QVT)
 - Number of missed (non-selectable) vessels were counted for both tools.
- Intra-observer variability of flow was calculated between two observers.
- Conservation of flow at ICA, MCA, ACA and TS, SS, SSS junctions was assessed with QVT.

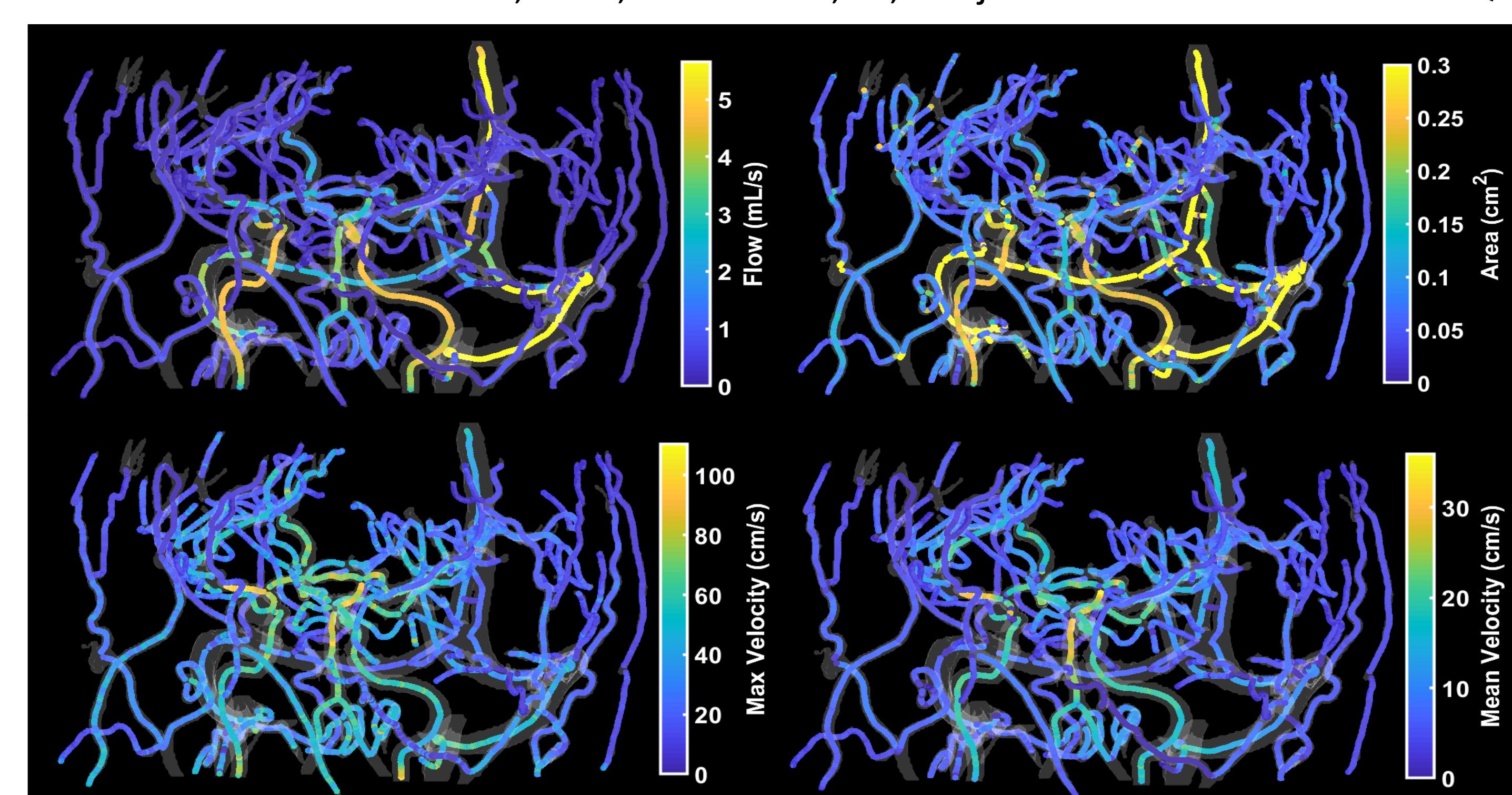


Figure 3: Quantitative parameters (area, flow, velocity) color-coded on automatically-generated centerlines.

Results

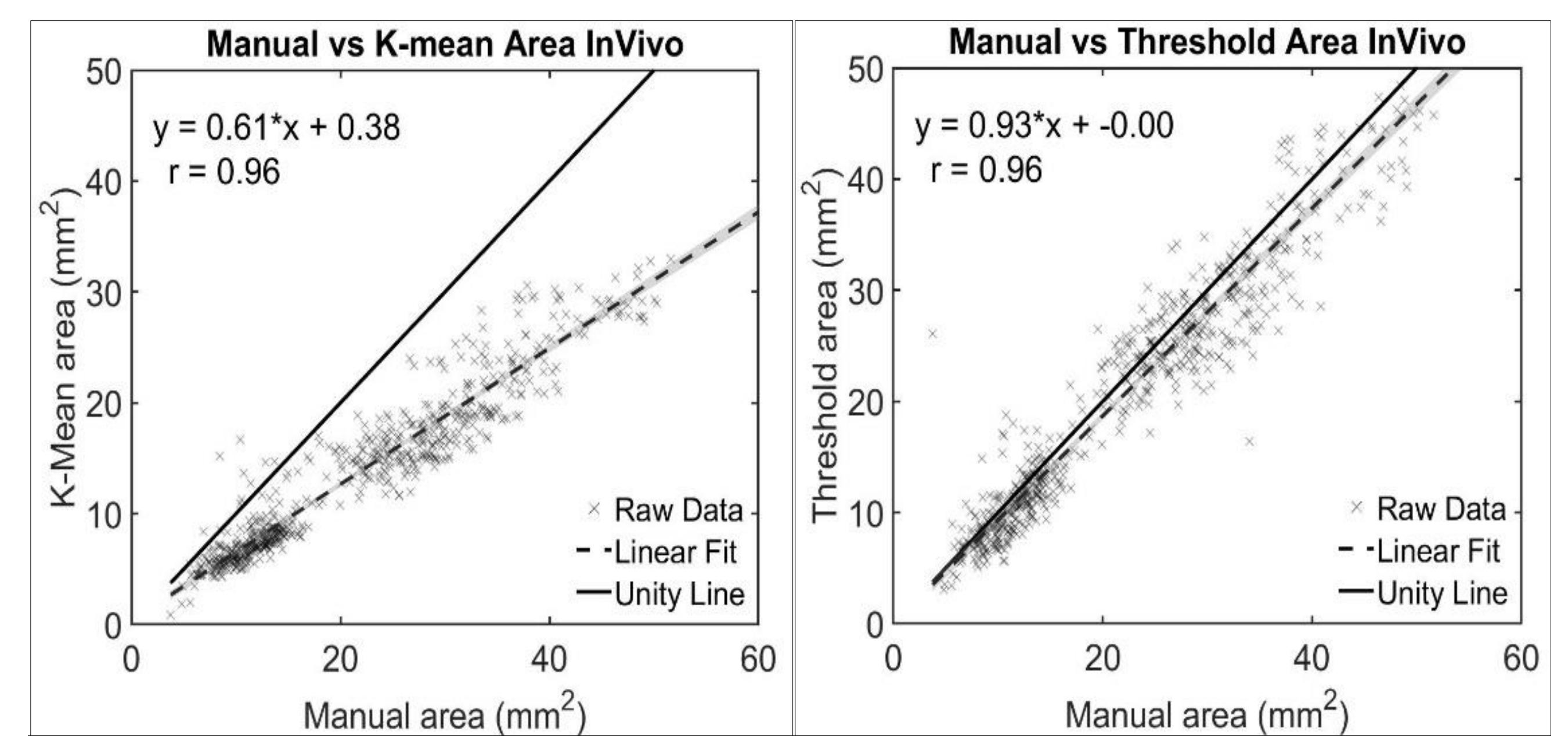


Figure 4: Linear regression (correlation) plots of k-means-derived (CPS) segmented vessel areas and iterative threshold-based (QVT) areas versus ground-truth manual segmentation.

- Regression (Figure 4) for both segmentation methods showed high correlation ($R=0.96$).
- Threshold-based segmentation was more accurate compared to manual segmentation (slope=0.93) than k-means segmentation (slope=0.61).
- Examples of segmentation improvements are shown in Figure 5.
- Number of missed vessels reduced from 16.2% to 3.8% when analyzed in QVT.
 - MCA, PCA, and TS accounted for 25/26 missed vessels.
 - Average variance of flow along vessel segments was similar for CPS ($2.8 \pm 2.6\%$) and QVT ($2.9 \pm 2.2\%$).
- Bland-Altman analysis of flow quantification between users showed a mean difference and 95% CIs of $0.042 [-0.37, 0.45]$ ml/s with the QVT (Figure 6).
- Bias and CIs are reduced when compared to old CPS $-0.1 [-1.88, 1.68]$ ml/s.
- Internal consistency of flow was observed at vessel junctions for the QVT (Figure 7).

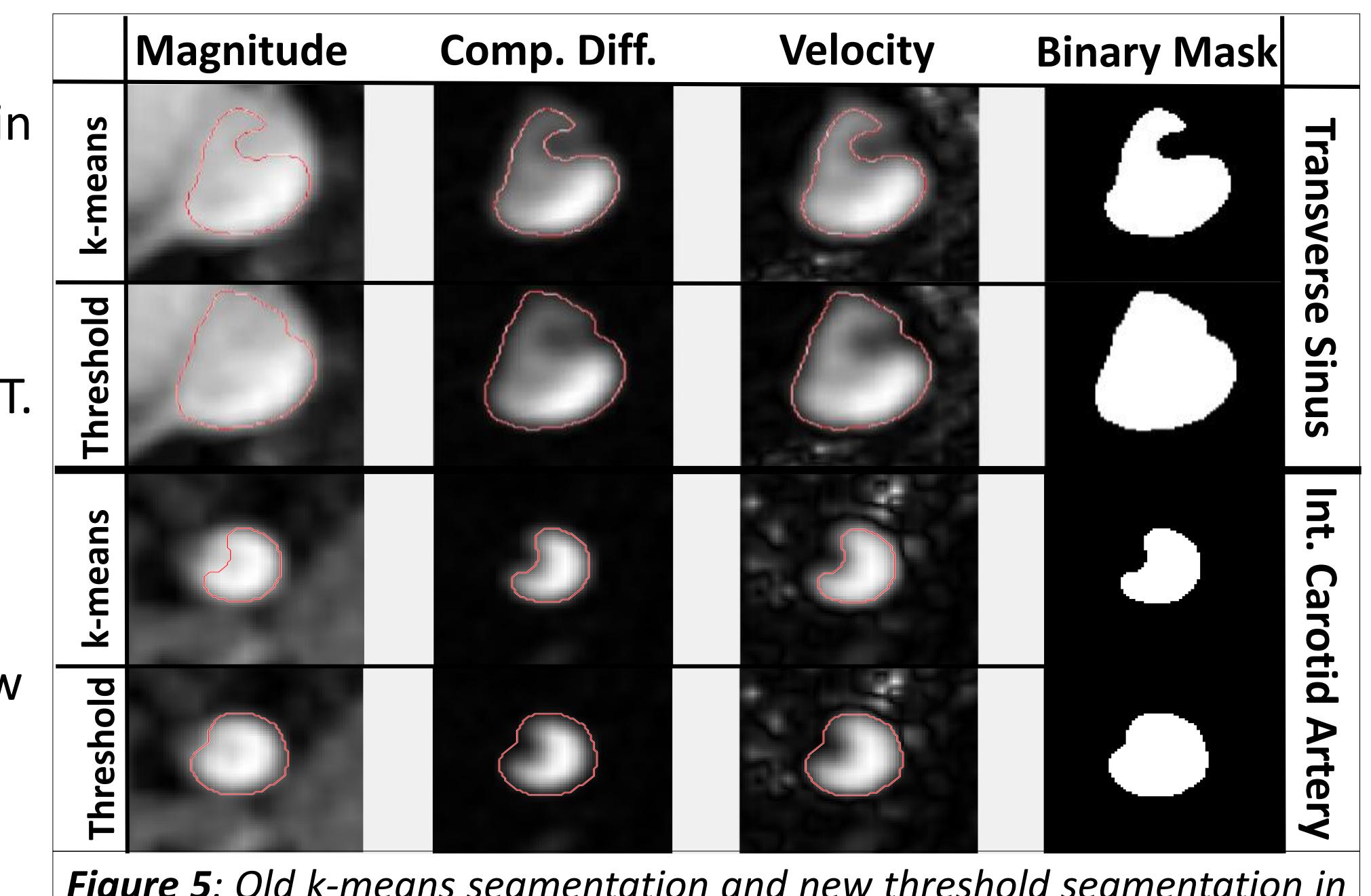


Figure 5: Old k-means segmentation and new threshold segmentation in transverse sinus (top) and internal carotid artery (bottom).

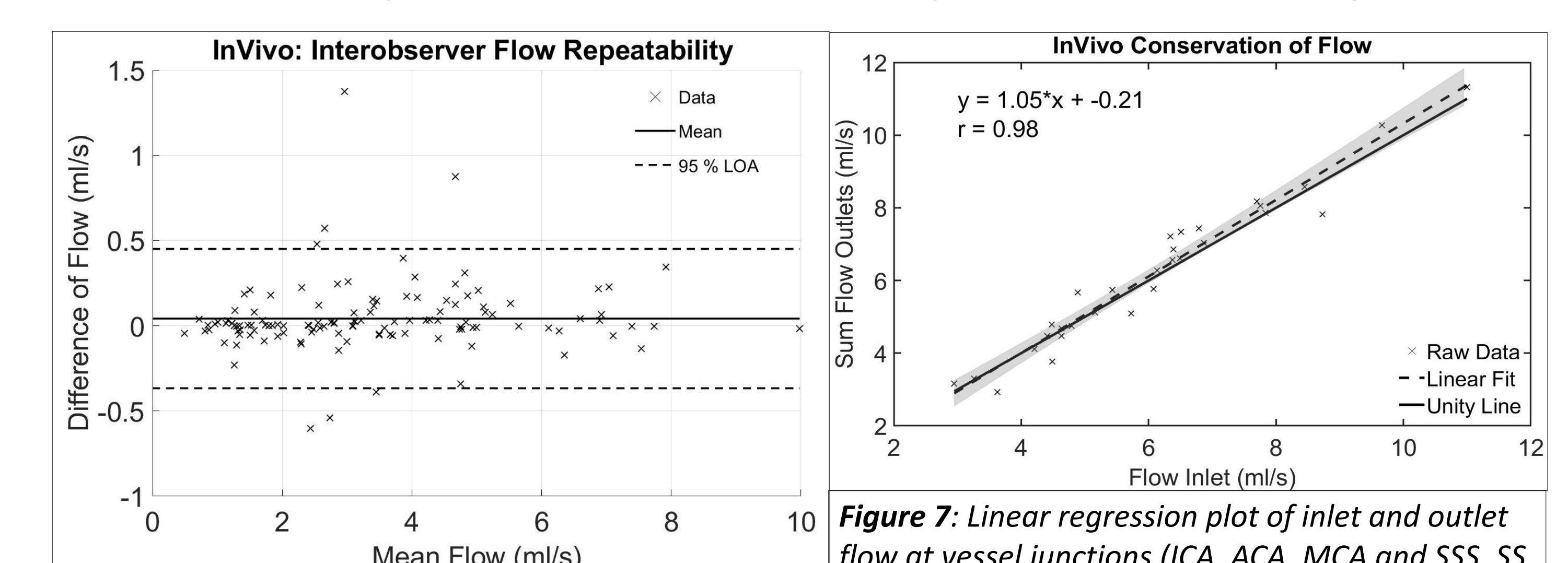


Figure 6: Bland-Altman plot of inter-observer repeatability for QVT.

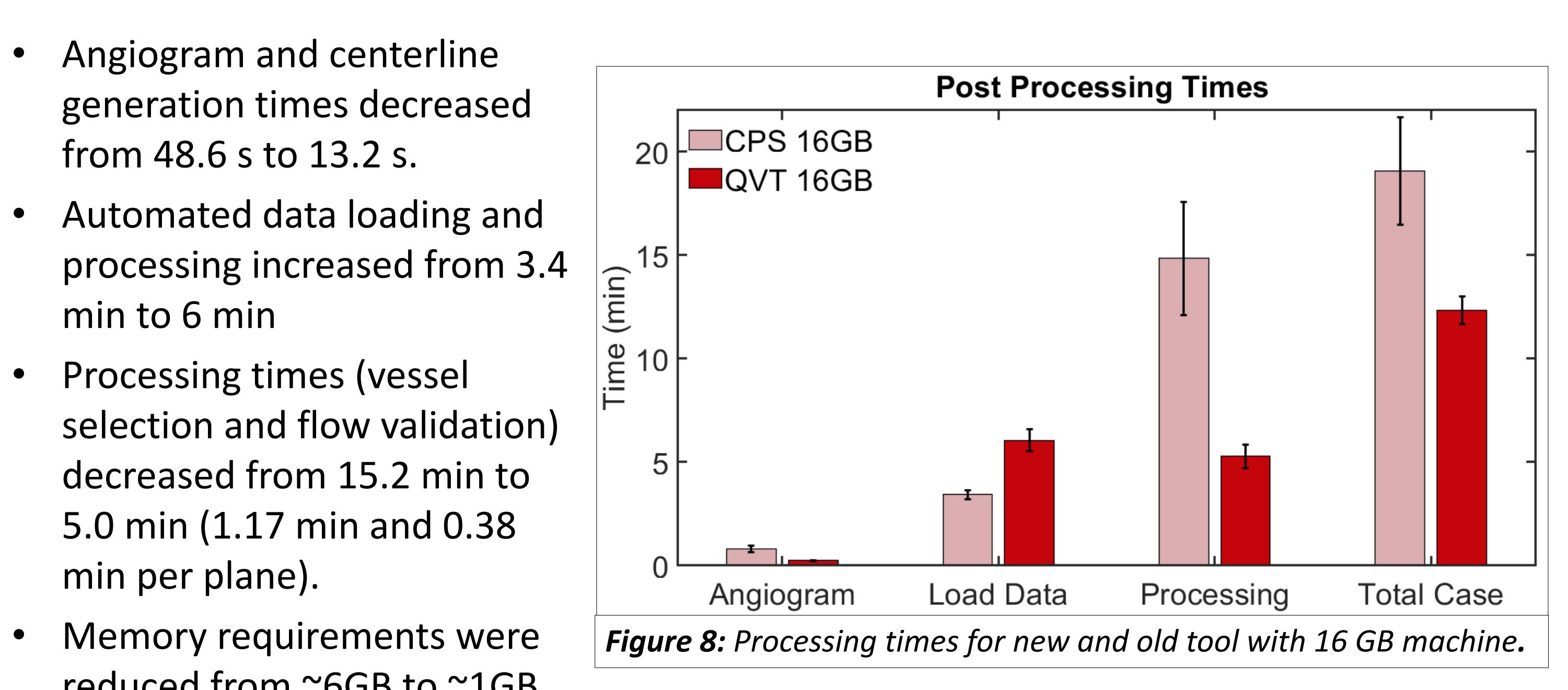


Figure 7: Linear regression plot of inlet and outlet flow at vessel junctions (ICA, ACA, MCA and SS, TS junctions). High correlation reflects good internal flow consistency with the QVT.

- Angiogram and centerline generation times decreased from 48.6 s to 13.2 s.
- Automated data loading and processing increased from 3.4 min to 6 min
- Processing times (vessel selection and flow validation) decreased from 15.2 min to 5.0 min (1.17 min and 0.38 min per plane).
- Memory requirements were reduced from ~6GB to ~1GB

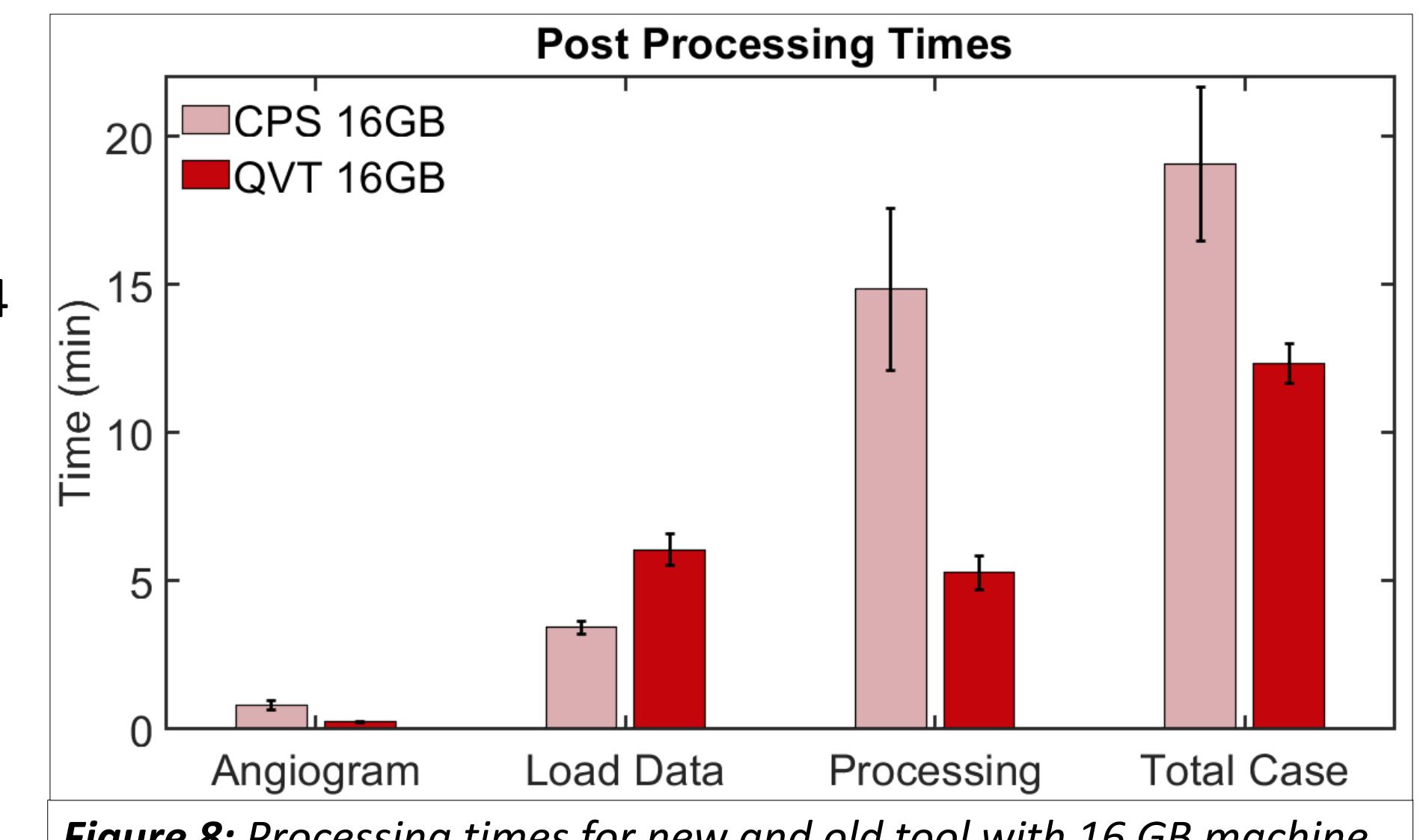


Figure 8: Processing times for new and old tool with 16 GB machine.

Discussion

- Robust analysis tools with high repeatability are extremely desirable for high-volume, long-term 4D flow studies.
- Both tools showed high user repeatability, small local flow measurement variations (<3%), and quick processing times.
- However, we confirmed that the QVT provided improved angiograms, quicker vessel selection, easier visualization, reduced computer memory requirements, and faster processing times.
 - Due to these improvements, we suggest utilizing the QVT moving forward.
- To further verify these findings, more extensive repeatability and flow consistency testing should be performed.

References

- [1] Schrauben, E. et al. JMRI 2015; 42:1458-64. [2] Gu, T. et al. AJNR 2008; 26:743-749