Quantitative collagen assessment in right ventricular myectomies from patients with tetralogy of Fallot

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Aims

Patients with tetralogy of Fallot (TOF) are often affected by right ventricular fibrosis, which has been associated with arrhythmias. This study aimed to assess fibrosis distribution in right ventricular outflow tract (RVOT) myocardium of TOF patients to evaluate the utility of single histology-section analyses, and to explore the possibility of fibrosis quantification in unlabelled tissue by second harmonic generation imaging (SHGI) as an alternative to conventional histology-based assays.

Methods and results

We quantified fibrosis in 11 TOF RVOT samples, using a tailor-made automated image analysis method on Picrosirius red-stained sections. In a subset of samples, histology- and SHGI-based fibrosis quantification approaches were compared. Fibrosis distribution was highly heterogeneous, with significant and comparable variability between and within samples. We found that, on average, 67.8 mm\(^2\) of 100 mm\(^2\) thick, histologically processed tissue per patient had to be analysed for accurate fibrosis quantification. SHGI provided data faster and on live tissue, additionally enabling quantification of collagen anisotropy.

Conclusion

Given the high intra-individual heterogeneity, fibrosis quantification should not be conducted on single sections of TOF RVOT myectomies. We provide an analysis algorithm for fibrosis quantification in histological images, which enables the required extended volume analyses in these patients.

Keywords

Fibrosis • Histology • Right ventricle • Congenital heart disease • Tetralogy of Fallot • Second harmonic generation microscopy

Introduction

Cardiac fibrosis—excess accumulation of extracellular matrix in the myocardium—is associated with significant functional impairment and enhanced morbidity and mortality, both in acquired and in congenital heart diseases.\textsuperscript{1} One of the most common cyanotic congenital heart defects is tetralogy of Fallot (TOF).\textsuperscript{2} TOF presents as a combination of a non-restrictive ventricular septal defect, an overriding aorta, right ventricular (RV) outflow tract (RVOT) obstruction, and RV hypertrophy. Following successful operative repair, patients usually have good long-term survival and high quality of life.\textsuperscript{3} However, they face a significantly increased risk for severe ventricular...
arrhythmic complications and sudden cardiac death, with associated sub-clinical ECG changes both before and after operative repair. Fibrosis has been shown previously to be a common feature in right and left ventricles of TOF patients. In RV myocardium of TOF patients, different types of fibrosis have been described, including compact/focal, perivascular, and diffuse interstitial fibrosis. Importantly, all types of fibrosis are thought to contribute to ventricular arrhythmogenesis in repaired TOF patients, making accurate fibrosis assessment an important diagnostic and therapeutic target. Previous studies have employed traditional, mostly single two-dimensional (2D) section-based, histopathological evaluation of TOF myocardium to assess fibrosis in unpaired and repaired TOF patients. Most studies, while indicating significant RV collagen deposition in TOF, are of limited interpretability due to non-quantitative fibrosis characterization and/or restriction of analyses to a small number of sections or fields of view. In order to assess the utility of single histological section analyses for fibrosis quantification, we evaluated intra-individual heterogeneity of fibrosis distribution in RVOT myocardium, resected intraoperatively from TOF patients. We established an automated, observer-independent image analysis method for histology-based fibrosis quantification. The method has been designed to include all forms of intramyocardial fibrosis, while excluding areas of non-myocardial collagen-rich structures, such as thickened endocardium.

While efficient fibrosis quantification can reliably be achieved using the here-described method, the preparation of samples (fixation, dehydration, wax-embedding, serial sectioning, mounting, staining, and coverslipping) remains time-consuming and prone to artefacts such as non-homogeneous shrinkage of the tissue, which can affect absolute values of downstream fibrosis evaluation. Moreover, collagen assessment in histological sections is inherently confined to a pseudo-2D space, which can be error-prone if fibrosis is not isotropically distributed. Therefore, we compared serial histology assessment in stacks of sections to single-section manual fibrosis quantification, and to collagen assessment using second harmonic generation imaging (SHG) in live, non-fixed and in chemically fixed TOF RVOT tissue. SHG is based on the principle that interaction of two photons with an optically nonlinear medium, such as collagen, leads to the formation of one new photon with twice the energy (or half the wavelength) of the initial photon. This permits three-dimensional (3D) high-resolution imaging of fibrillar collagen in unlabelled, native tissue. While SHG-based data acquisition is restricted with respect to imaging depth, it offers the feasibility of collagen fibre orientation analysis in 3D volumes, and identified differences in collagen fibre anisotropy in RVOT samples across TOF patients.

**Methods**

**Tissue collection**

RVOT tissue samples, resected during repair operation or operative valve replacement from TOF patients, were collected in the CardioVascular BioBank (CVBB) of the University Heart Center Freiburg—Bad Krozingen. The CVBB, as well as the use of tissue and data from the CVBB for this project, have been approved by the ethics committee of the Albert-Ludwigs University Freiburg (ethical approval number 393/16 for the CVBB and 589/17 for this study). The study complies with the Declaration of Helsinki and informed consent was obtained from all patients or their legal guardians.

**Histological tissue processing and analysis**

RVOT tissue was fixed in 4% paraformaldehyde for 12–16 h, dehydrated, and embedded in wax (Tissue Processor TP1020, HistoCore Arcadia H, HistoCore Arcadia C, all Leica Biosystems, Wetzlar, Germany) according to standard protocols (see Supplementary material online, Methods section). Tissue samples were microtome-cut into 10 μm thick sections (Rotary Microtome RM2235, Leica) and every second section was batch-stained using an automated stainer (Autostainer XL ST5010, Leica) with Picrosirius red (for detailed protocol, see Supplementary material online), allowing for clear distinction of collagen (red) vs. cytoplasm and nuclei (yellow; see Figure 1A). Stained sections were automatically coverslipped (Glass Coversliper CV5030, Leica), before being scanned and digitized with a native pixel size of 0.442 μm × 0.442 μm (Axio Scan.Z1, Carl ZEISS AG, Oberkochen, Germany). While we used Zeiss hard- and software for image acquisition, our fibrosis quantification algorithm is not sensitive to the hard- and software used, as long as the colours of the sections after staining are preserved and the resolution is sufficient to enable distinction of small collagen fibres (usually >1 μm).

**Automated fibrosis quantification in histological sections**

Images were computationally processed and analysed using custom Python scripts based on open-source packages numpy and scipy. All programme codes developed for the methods described here are available in Supplementary material online. Image data were read from Zeiss’ proprietary CZI format, using Bio-Formats through the python-bioformats package. Each CZI file contained the image at full resolution and at downsized resolutions (different zoom levels). Zoom level 2 of the slide scanner image files (corresponding to a square of 3 × 3 native pixels, resulting in a pixel size of 1.325 μm × 1.325 μm) was used for automated fibrosis quantification.

Images were transformed from the standard additive red–green–blue colour space to hue–saturation–value (HSV) colour space using OpenCV (version 3.4.4). In HSV, pixel colour is composed of components hue (H), determining the colour hue in degrees on a colour wheel: saturation (S), on a scale between grey and neutral; and brightness, called value (V), on a scale between black and bright colours. Every pixel was classified automatically as myocardium, fibrosis, or background, based on thresholds in HSV colour space, which were determined from the colour
Histograms of the whole image (Figure 1). Histograms were median-filtered (window of 3 bins) to reduce quantization noise.

Background intensities appear as a prominent sharp peak at the lower end of the S-histogram. The minimum between this peak and the next (lower and wider) peak was used as a threshold: pixels with an S-value below this threshold were considered background (Figure 1B).

The remaining pixels were segmented as either myocardium or collagen, based on the H-histogram, which shows the two populations as separate peaks (Figure 1C). Since the H-values are based on a colour wheel (i.e., they are cyclical), two thresholds were defined at the central minima between the two peaks, separating the pixels into two populations. By determining the thresholds in HSV individually, based on the histograms of each image, rather than applying a fixed threshold for all images, the fibrosis quantification results are not influenced by potentially arbitrary threshold selection.

Next, to exclude adjacent non-myocardial, collagen-containing structures, especially thickened endocardium (Figure 1D, blue arrow), images were divided uniformly into non-overlapping squares of 50 pixels × 50 pixels (superpixels) with an area of 4389 μm² each. For each of these superpixels, the relative amount of myocardium contained was determined by dividing the number of myocardium pixels by the number of all non-background pixels (if there were only background pixels in a square, this ratio was set to 0). A breadth-first-search was then employed to exclude certain superpixels from further analysis: starting from all corners of the image, superpixels with relative myocardium content below a threshold of 0.33 were excluded from further analysis, if they could be reached via other superpixels with a myocardium content below the same threshold. Fibrosis content for the section was then determined for the remaining superpixels, as the percentage of pixels that were segmented as fibrosis relative to the overall number of non-background pixels (Figure 1D). Severe reduction or increase in superpixel size or myocardium content threshold (e.g., to < 5 pixels × 5 pixels) or > (95 pixels × 95 pixels); thresholds of <0.1 or >0.6) leads to erroneous exclusion of intramyocardial fibrosis or inclusion of parts of thickened endocardium, and thus to under- or overestimation of fibrosis. However, superpixel sizes between 25 pixels × 25 pixels and 75 pixels × 75 pixels, and myocardium content thresholds between 0.1 and 0.6 yielded relatively constant values for percent fibrosis (Supplementary material online, Table S1 and Figures S1 and S2).

Manual fibrosis quantification in histological sections

One section per tissue block was randomly selected for manual fibrosis quantification and subsequent method validation. In each raw image, an experienced human observer performed manual segmentation of collagen structures. A mask was created manually to exclude adjacent confounding collagen-containing myocardial structures (e.g., endocardial
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thickening), as well as the background. Compared to the automated method, which also identifies small clefts in the tissue as background, such small clefts remained within the mask. Percentage of interstitial fibrosis was then determined by calculating the proportion of the area identified as containing collagen relative to the overall area of non-background tissue in the mask.

Local eccentricity analysis in histological sections

In order to check for a possible correlation between detected percentage of fibrosis by automated quantification and the alignment of tissue cutting plane with locally prevailing cardiomyocyte orientation, we developed a surrogate parameter that can be computed easily for each of the superpixels that we analysed. This parameter was designed to describe the eccentricity of structures (such as muscle cells) within each square. It is computed from the 2D discrete Fourier transform of each image square. A Gaussian window of the form

$$w(x, y) = \exp\left(-\frac{1}{2} \left(\frac{x - Q}{s} \right)^2 - \frac{1}{2} \left(\frac{y - Q}{s} \right)^2\right)$$

with Q the width and height of the superpixel (here, Q = 50 level-2 pixels), x, y < Q the co-ordinates local to the image square, and s = |W/H| a width parameter, was applied to the blue channel image square. The blue channel was chosen for the strong contrast that cardiomyocyte membranes exhibited there. The discrete Fourier transform was then calculated using the fast Fourier transform (FFT) algorithm as implemented in the numpy python library.

With the FFT quadrants ordered such that the offset intensity (FFT co-ordinate [0, 0]) is located in the centre pixel, the FFT shows a point-symmetric image of the contribution of spatial frequencies to the original square. Following the intensities (i.e. absolute values) of the FFT along a line through the centre point, one can deduce regularity of structures at that angle in the original image. At angles where the FFT features multiple local maxima of intensities, the original image will have many periodic repetitions of an intensity pattern (e.g. membranes) closely together. At angles where the FFT shows no maxima other than the centre, no periodic repetitions occur in the image (e.g. along a cardiomyocyte’s longitudinal direction). Thus, if the original square shows cardiomyocytes cut perpendicularly to their long axis (visible membranes forming circles), the FFT will exhibit a round pattern of multiple concentric waves. The closer the tissue sectioning plane is to the cardiomyocyte orientation (cells cut longitudinally), the more elliptic the concentric FFT pattern will be.

To quantify the ellipticity of the FFT pattern, we used image moments. First, we calculated the FFT’s second-order moments matrix

$$M = \begin{bmatrix} \mu_{00} & \mu_{01} & \mu_{02} \\ \mu_{10} & \mu_{11} & \mu_{12} \\ \mu_{20} & \mu_{21} & \mu_{22} \end{bmatrix}$$

with the image central moments $\mu_{ij}$ of the image of absolute FFT intensities. OpenCV was used to calculate image moments. With $\lambda_1 > \lambda_2$ the eigenvalues of M describing the variance along the major axes of the FFT, ellipticity was calculated as

$$\varepsilon = \sqrt{1 + \frac{\lambda_2}{\lambda_1}}.$$

We tested positive correlation between our surrogate parameter $\varepsilon$ and the eccentricity of ellipses in randomly generated images, also in the presence of random noise (see Supplementary material online, Figure S3). Plausibility of the resulting ratio as a surrogate for cutting direction was verified visually for random samples and samples with extreme values. Only squares with more than 50% cardiomyocyte content were considered for local eccentricity analysis, as the method depends on the presence locally of representative cardiomyocyte outlines.

Second harmonic generation imaging

Live RVOT tissue samples from five patients were cut into 500 μm thick tissue slices with a vibratome (7000sz, Campden Instruments, Loughborough, UK) and mounted on glass slides in phosphate buffered saline. Three randomly chosen regions were immediately imaged with SHGI for each tissue slice. SHGI was performed using an upright multi-photon microscope (TCS SP8 DIVE, Leica Microsystems, Wetzlar, Germany). The SHGI signal was generated using 920 nm excitation light from a pulsed laser (InSight X3 Dual, Spectra-Physics, Santa Clara, USA) and detected with a hybrid detector using a 25 x water-immersion objective (IRAO L 25x/1.00 W, Leica Microsystems). Autofluorescence was also recorded using a second detection channel. 3D volumes were acquired with a size of 442.86 μm × 442.86 μm in x-y-direction, and 90–110 μm in z-direction (x-y-z voxel size: 0.099 μm × 0.099 μm × 0.51 μm). Excitation compensation was applied linearly along the z-axis to compensate for intensity loss at increasing tissue depths. After live tissue imaging, slices were fixed for 12h in 4% paraformaldehyde. Tissue landmarks were used to identify regions that had been imaged previously in live tissue, and SHGI of those areas was repeated. Due to technical issues, one of the five tissue slices was imaged only before fixation.

Fibrosis quantification in second harmonic generation imaging volumes

SHGI volumes were down-sampled by a factor of 2 in x- and y-directions to reduce computational costs for volumetric operations on the high-resolution image volumes. We investigated the effect of downsampling on our analysis, which showed them to be of little to no effect (see Supplementary material online, Figure S4). Volumes were then median-filtered (window: 5 pixels in x- and y-directions, and 3 pixels in axial direction) to reduce pixel noise. As SHGI is a nonlinear process, collagen signal was segmented by forming a threshold-based binary image, similar to previously described algorithms for 2D SHGI images. With $W$ the mode of the histogram and $\sigma$ the standard deviation of intensities, all intensities greater than $m + \sigma$ were considered collagen, and all other voxels were considered background. The collagen percentage of the SHGI stack was thus the percentage of voxels segmented as collagen relative to the total number of voxels of the tissue volume. It must be noted, that despite adapting the threshold for each individual SHGI volume by this method, fibrosis quantification still remains sensitive to the threshold applied (Supplementary material online, Figure S5, e.g. the increase or decrease of percent fibrosis measured depends on multiplication of the factor $\sigma$), and therefore percent fibrosis in SHGI volumes should only be compared if thresholds are defined in a consistent manner across samples. Areas of non-tissue did not need to be specifically excluded, as all imaging was performed exclusively within the tissue, without identifiable breaks or gaps.

To enable the comparison of collagen content in SHGI volumes of live and fixed tissue, we determined fixation-induced tissue shrinkage. We measured the lateral distance between unconnected collagen fibres as an indicator of myocardial tissue shrinkage, as well as the distance between identifiable points along the same collagen fibres to determine para-fibrillar shrinkage, in ≥50 regions in each of the non-fixed and corresponding fixed SHGI stacks (see Supplementary material online, Figure S6). Tissue volume shrinkage was assumed to correspond to the third power of mean differences observed in linear measurements. This was used to determine percent shrinkage by which to correct the collagen content quantification in fixed SHGI volumes.
Collagen fibre orientation analysis in second harmonic generation imaging volumes

Exemplary 3D SHGI volumes from two patients with a visually different degree of collagen fibre anisotropy were analysed to quantify collagen fibre orientation. Both volumes were down-sampled from 4472 pixels \( \times \) 4472 pixels to 1118 pixels \( \times \) 1118 pixels in \( xy \)-directions, and collagen fibre skeletons were determined semi-automatically using an ImageJ-based open-source software developed for tracing neurons and other tubular structures.26 A principal component analysis of the determined fibre skeleton was used to compute fibre directionality: the eigenvector with the largest eigenvalue was defined as the orientation of the traced fibre. In one volume, cardiomyocyte autofluorescence proved clear enough to determine cardiomyocyte longitudinal axes, so that strands of cardiomyocytes were traced and analysed by the same method. In the other volume, cardiomyocyte autofluorescence was too weak to identify single cardiomyocyte strands.

Statistical analysis

Paired Student’s \( t \)-tests were used for the comparison of manual and automated fibrosis counts, and for comparison of SHGI data obtained from the same tissue before and after chemical fixation. Unpaired Student’s \( t \)-tests were used for all other analyses. Analyses were performed with OriginPro 2017 (OriginLab Corporation, Northampton, NA, USA). \( P < 0.05 \) was considered indicative of a significant difference between means. Values are presented as mean ± standard deviation.

Results

Patient and sample data

RVOT tissue samples from nine unrepaired TOF patients (age 5–39 months, median 8 months), one unrepaired patient with ventricular septal defect and pulmonary stenosis (age 20 months), and one repaired TOF patient (age 27 years) were collected. Tissue sample thickness ranged from 0.5 to 8 mm (median 4 mm). Before histological processing, five tissue samples from unrepaired patients (age 5–20 months, median 10 months) were imaged by live tissue SHGI. All samples were thereafter fixed, sectioned and stained for histological analysis. Sections with artefacts (such as tissue folds) or inconsistent staining were excluded from the analysis. The number of sections analysed per tissue sample ranged from 30 to 150 sections (median 50 sections per sample; 748 sections in total with 7032 mm\(^2\) of total tissue area). On qualitative inspection of sections, we were able to identify three previously described forms of fibrosis: compact/focal, perivascular, and interstitial diffuse fibrosis. Two or three fibrosis forms could occur in the same sample or even section. Samples demonstrated varying degrees of endocardial thickening.

Collagen fibre orientation analysis in second harmonic generation imaging volumes

Figure 2 Comparison of automated and manual fibrosis quantification in histology sections. (A) Pairwise values of automated vs. manual histological fibrosis quantification in one section per patient sample, overall resulting in a higher mean for automated fibrosis quantification (red) compared to manual (black). (B) Exemplary eccentricity analysis of the myocardium within all superpixels of all sections of one patient (black) and of one section (red); no correlation between cardiomyocyte orientation and percent fibrosis is seen for this sample (Pearson’s \( r = -0.016 \)) and across all samples (Pearson’s \( r = -0.017 \)).
We further assessed whether the direction of sectioning relative to locally prevailing cardiomyocyte orientation had an influence on fibrosis quantification, using eccentricity analysis of cell cross section shapes. While this showed a weak positive or negative correlation in 4 out of 11 samples, there was no correlation between eccentricity and percent fibrosis across grouped data from all samples (Pearson’s $r = -0.017$, Figure 2B).

**Inter- and intra-individual variability of percent fibrosis in histological samples**

Across all patient samples, mean percent fibrosis was 14.94 ± 2.7%, as measured by automated quantification. Mean percent fibrosis ranged from 4.59% to 29.98% between individual samples (Figure 3A). The mean and maximum deviation of the mean of one patient sample from the overall group mean were $5.20 \pm 4.58\%$ and $15.04\%$, respectively.

Within the individual patient samples, the fibrosis distribution pattern varied (Figure 3B), demonstrating generally similar fibrosis percentage in closely neighbouring sections. The mean of the maximum difference between two nearest sections (10 $\mu$m gap in-between) was $4.53 \pm 1.58\%$. The mean and maximum range of percent fibrosis within individual samples established on the basis of a single section were $11.07 \pm 5.35\%$ and $20.01\%$, respectively (Figure 3A, and Patient 5 in Figure 3B). In relative terms, this corresponds to an up to 1.4-fold over-estimate or an up to 1.8-fold under-estimation of actual RVOT tissue fibrosis levels if analyses were based on a single tissue section only. The mean and maximum deviation of one section from the overall sample mean (in absolute %) were $6.59 \pm 3.25\%$ and $11.79\%$, respectively.
Mean section area was 9.75 mm² (range 3.11–20.20) across all patients, with a minimum of 2.35 mm² (range 0.06–6.80) and a maximum of 14.17 mm² (range 5.19–31.41) across all patients. We assessed the minimum number of sections within one sample that had to be analysed in order to obtain a mean fibrosis percentage level that would be within the range of mean ± 1.96 SEM of the sample (corresponding to the 95% confidence interval, assuming fibrosis percentage per section is normally distributed). With a probability of 95%, this ranged from 2% [1 out of 50 sections, Figure 3, Patient 4 (mean section area for this patient: 11.38 mm²)] to 55% of all analysed sections [52 out of 94 sections, Figure 3, Patient 5 (mean section area for this patient: 6.22 mm²)]. When converting this to minimum required tissue area, analysis of an average of 67.79 mm² [ranging from between 8.89 mm² for highly homogeneous (Figure 3, Patient 8) to 318.64 mm² for highly non-homogeneous tissue (Figure 3, Patient 5)] of 10 μm thick fixed, dehydrated, wax-embedded tissue was needed to arrive at a mean fibrosis percentage that would, with 95% probability, lie within the 95% confidence interval of the sample mean.

Assessment of fixation-induced tissue shrinkage by second harmonic generation imaging

The assessment of fixation-induced tissue volume changes by SHGI showed a significantly higher reduction for linear measurements across myocardial (non-collagenous) structures than for linear measurements along collagen fibres (reduction of 4.69 ± 5.90%, n = 52 vs. 0.23 ± 4.72%, n = 50, P < 0.001). This suggests that shrinkage occurs predominantly in the cell-rich tissue areas, whereas the length of collagen fibres remains relatively constant. The associated volume effect corresponds to tissue shrinkage of 13.42% after fixing, but before dehydration and embedding.

Collagen percentage and collagen fibre orientation measured by second harmonic generation imaging

Collagen content in SHGI volumes from live tissue samples was compared to the corresponding regions after fixation. Fixed SHGI data were corrected for fixation-induced shrinkage. In total, 13 regions in tissue samples from five patients were analysed. There was no significant difference of collagen percentage in the pairwise comparison of SHGI volumes in live and shrinkage-corrected fixed tissue (12.21 ± 10.55% vs. 11.33 ± 10.20%; n = 9 volumes from n = 5 patients, P = 0.839).

We analysed collagen fibre orientation in exemplary SHGI volumes from two patients, both aged 5 months (Figure 4). Assessment of the azimuthal angle φ represents the in-plane orientation (x–y), and the polar angle θ represents the out-of-plane orientation (with 0° referring to an orientation perpendicular to the x–y plane) of each traced collagen fibre. Volume A demonstrated a homogenous distribution of collagen fibres, i.e. a relatively anisotropic network (Figure 4A). In Volume B, the azimuthal distribution favoured 70–180°, whereas polar orientation was restricted to 45–135°, i.e. we saw an isotropic organization of collagen with few cross-linking
collagen fibres, both in-plane and out-of-plane (Figure 4B). Due to clear autofluorescence of cardiomyocytes in Volume B, we were able to relate collagen fibre orientation to cardiomyocyte orientation (see Supplementary material online, Figure S7), which appeared similar to collagen fibre orientation in this volume. Autofluorescence in Volume A was not clear enough to enable cardiomyocyte orientation analysis. This confirms the principal suitability of SHGI for assessment of 3D fibrosis distribution.

Discussion

Our findings emphasize that: (i) RVOT fibrosis in TOF is high (mean 14.94 ± 2.7% in histology data) with pronounced inter-individual variability (mean deviation of a single patient sample from the group mean 5.20 ± 4.58%; maximum deviation 15%); (ii) intra-individual variability of single section fibrosis levels shows variability similar to the level of inter-individual differences (mean deviation of one section from the patient sample mean 6.59 ± 3.25%; maximum deviation 12%); (iii) analysis of an average tissue area of 67.79 mm² (of 10 μm thick tissue sections) per patient is necessary for reliable fibrosis assessment representative for the given TOF patient; and (iv) SHGI data acquisition from a subgroup of samples yielded fibrosis levels of 12.21 ± 10.55%, and it may have potential for in vivo assessment of myocardial tissue structure, offering an exciting direction for further research.

We present an automated analysis method for fibrosis quantification in 2D histology sections of cardiac tissue samples. A comparison of automated and manual fibrosis quantification demonstrated good separation of adjacent collagen-containing structures, such as thickened endocardium or endocardial fibroelastosis, from myocardium with both approaches. Manual histological fibrosis quantification generally reported a lower collagen percentage than automated analysis (Figure 2A), which we attribute to more precise identification of small collagen structures by automated segmentation. Therefore, the automated analysis approach will be advantageous in particular for samples containing prominent diffuse or punctate collagen deposits. Additionally, automated fibrosis quantification was more than 1000 times faster than manual fibrosis assessment, making it a time-saving and effective approach.

The here reported mean percent fibrosis of 14.94% in automatically analysed histology sections is slightly higher than previous reports from TOF RVOT, where 3–11% were observed by semi-automated, colour-based quantification in histological studies, but similar to fibrosis found in other RV areas (13–18%) (see also Supplementary material online, Introduction section).

Inter-individual variability, with collagen percent ranging between 4.59% and 29.98% across samples, was similar to one prior report (6.0–27.5%) and, somewhat higher than two other RV fibrosis studies (9.7–18.3% and 3.0–11.5%).

More importantly, there was pronounced intra-individual heterogeneity of fibrosis in TOF RVOT tissue, which was of similar magnitude to the extent of inter-individual variability (Figure 3). This intra-individual variability necessitates analysis, on average, of 67.8 mm² of 10 μm thick fixed, dehydrated and embedded tissue, for reliable fibrosis quantification in RVOT tissue. This corresponded to 5–10 individual histology sections from our intraoperatively resected myectomies. For endomyocardial biopsies, for which a volume of 2–3 mm³ is considered optimal, one would need to study approximately 35 histology sections of such a biopsy, if it was cube-shaped and optimally cut. Therefore, our data highlight, that the utility of single section-based fibrosis analysis must be regarded as severely limited and significantly error-prone when quantifying fibrosis in TOF RVOT myectomies or biopsies.

Histology-based fibrosis assessment requires myectomies, and is time consuming (even with automated image analysis), as samples have to be dehydrated, embedded, sectioned, stained, coverslipped, and imaged. Also, high-quality sections must be obtained (no tears or folds, with consistent staining) for robust analysis. In contrast, SHGI allows for collagen imaging without the need for extensive tissue preparation prior to imaging, including the option of immediate data acquisition from live tissue. To assess its utility, to probe the effect of chemical fixation on apparent fibrosis content, and to explore the possibility of 3D collagen analysis, we imaged five patient samples with 3D SHGI in their native, live (non-fixed) state, and re-assessed them after chemical fixation. Chemical fixation led to more than 13% shrinkage of the myocardial tissue volume in our study, while collagen structures retained their dimensions. This is similar to previous reports of chemical fixation-induced linear shrinkage of 2.7% (corresponding to a volume effect of ∼8%) in smooth muscle, while collagen structure is preserved. After corresponding correction, SHGI-determined collagen (volume)-percent did not differ between live and fixed samples. This demonstrates that SHGI may be used with similar results both in native, non-fixed, and unstained tissue, and in paraformaldehyde-fixed tissue, for example to examine cardiac structures at later time points. This increases the utility of SHGI-based assessment of fibrosis, especially for rare samples, such as excised human ventricular myocardium.

As demonstrated in our SHGI measurements, chemical fixation induces tissue shrinkage of around 13% in our samples. Previous findings of Burton et al. demonstrated that on further histological processing, involving not only chemical fixation but also dehydration and wax embedding, overall tissue shrinkage may reach 49%. Therefore, the amount of tissue required for representative fibrosis quantification in TOF RVOT is likely to exceed 1 mm³ of fresh tissue. As tissue shrinkage will depend on tissue composition, especially the extent of collagen present (which is less prone to shrinkage, as demonstrated by our findings and others), we did not attempt to correct our histology data with a single value from the literature.

Methods for fibrosis quantification by SHGI can be extended to include information concerning individual collagen fibres, such as length, width, volume, and morphology. As collagen fibre orientation affects myocardial mechanics both in the intact myocardium and in myocardial scars, we investigated the possibility of characterizing spatial aspects of the 3D collagen network in two SHGI volumes with visually distinctly different degrees of collagen anisotropy. Volume A demonstrated a relatively homogenous distribution of collagen orientations, while Volume B showed a more isotropic pattern (Figure 4). We were able to identify and quantify these differences by SHGI.

Our data confirm that SHGI may offer a tissue-preserving option for collagen quantification in living myocardium, as reported previously in non-cardiac tissue. The possibility of 3D characterization of the collagen fibre network, and the much faster data acquisition
Compared to histology, make SHGI a promising tool for investigating of cardiac collagen structures. Recent technical developments suggest that it may even be possible to conduct intravital two-photon and/or SHGI microscopy, using equipment that is miniaturized to the size range of endoscopes.\(^{(37,38)}\) Liang et al.\(^{(39)}\) developed a fibre-optic two-photon/SHGI endomicroscopy platform, which allows one to conduct in vivo imaging at histological resolution without signs of photo-induced tissue damage. Of course, further optimization is required for precise endoscopic 3D characterization, in particular, of a ‘moving organ’ such as the heart, but in principle a minimally invasive in vivo fibrosis assessment by SHGI may become feasible.

**Limitations and perspectives**

The amount of RVOT myocardium analysed in our study exceeded by far the amount of tissue analysed in prior fibrosis assessment reports in TOF (see Refs \(^{(13–18)}\) and Supplementary material online). As we concentrated on comparing and intra- and inter-individual fibrosis levels, the number of patients studied was small, and, for clinical interpretation regarding different TOF phenotypes, larger patient numbers would be required.

Furthermore, the exact location and orientation of tissue explants within the RVOT is not known, and, while we aimed to cut histological sections in parallel to cardiomyocyte long axes, we are unable to confirm the interrelation of section slicing plane relative to the cardiac wall. While we show that our analyses appear unaffected by the interrelation of cutting plane and local cardiomyocyte orientation, we cannot draw any conclusions about transmural or apico-basal fibrosis distribution in the RVOT.

Tissue fixation, and processing for histology, give rise to pronounced myocardial shrinkage.\(^{(30)}\) This means that our histology-based fibrosis data are likely to over-estimate fibrosis in terms of ‘volume-%’. We assessed fixation-induced shrinkage (which amounted to 13.4\% by same-region SHGI, but not the further tissue volume reduction caused by subsequent dehydration, as histology data does not provide reliable access to same-region tissue volume information.

The in-plane tissue dimensions analysed by SHGI in this study were limited, due to the nature of samples available. Extending observations into the z-direction is difficult, as with increasing z-depth the signal decreases while noise increases. Even though z-compensation allowed for slightly deeper imaging, we were not generally able to analyse volumes much beyond 100 \(\mu\)m z-depth. The here established minimum histological tissue area of 10 \(\mu\)m thick myocardium that needs to be analysed for reliable fibrosis quantification should be regarded as corresponding to a live histological sample volume of at least 1 mm\(^3\). Such volume would be accessible if one could image an x-y-area of about 3.16 mm\(^2\) with SHGI, assuming a z-depth of 100 \(\mu\)m. If one imaged tissue from two opposite surfaces (e.g. of an explanted sample), this would be reduced to about 2.23 mm\(^2\) surface area.

**Conclusions**

Compared to manual histological fibrosis quantification in Picrosiris red-stained myocardium, our automated quantification method is advantageous especially when investigating diffuse fibrosis. Our tissue samples demonstrated high levels of fibrosis in TOF RVOT, with significant and quantitatively comparable inter- and intra-individual variability. As analysis of an average tissue area of 67.8mm\(^2\) of 10 \(\mu\)m thick, histologically processed tissue is necessary for reliable assessment, fibrosis quantification should not be restricted to single histology sections. SHGI enables fast, high-resolution, and 3D collagen structure analysis in native tissue, providing additional information to complement findings from conventional histology. Optimization of SHGI protocols and larger tissue samples may even allow for comprehensive fibrosis quantification by SHGI only.

**Supplementary material**

Supplementary material is available at Europace online.

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**Conflict of interest:** The authors have no conflicts of interest.

**Data availability**

The data underlying this article will be shared on request to the corresponding author.

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