Identifying the optimal regional predictor of right ventricular global function: a high-resolution three-dimensional cardiac magnetic resonance study*

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Summary
Right ventricular (RV) function has prognostic value in acute, chronic and peri-operative disease, although the complex RV contractile pattern makes rapid assessment difficult. Several two-dimensional (2D) regional measures estimate RV function, however the optimal measure is not known. High-resolution three-dimensional (3D) cardiac magnetic resonance cine imaging was acquired in 300 healthy volunteers and a computational model of RV motion created. Points where regional function was significantly associated with global function were identified and a 2D, optimised single-point marker (SPM-O) of global function developed. This marker was prospectively compared with tricuspid annular plane systolic excursion (TAPSE), septum-freewall displacement (SFD) and their fractional change (TAPSE-F, SFD-F) in a test cohort of 300 patients in the prediction of RV ejection fraction. RV ejection fraction was significantly associated with systolic function in a contiguous 7.3 cm² patch of the basal RV freewall combining transverse (38%), longitudinal (35%) and circumferential (27%) contraction and coinciding with the four-chamber view. In the test cohort, all single-point surrogates correlated with RV ejection fraction (p < 0.010), but correlation (R) was higher for SPM-O (R = 0.44, p < 0.001) than TAPSE (R = 0.24, p < 0.001) and SFD (R = 0.22, p < 0.001), and non-significantly higher than TAPSE-F (R = 0.40, p < 0.001) and SFD-F (R = 0.43, p < 0.001). SPM-O explained more of the observed variance in RV ejection fraction (19%) and predicted it more accurately than any other 2D marker (median error 2.8 ml vs 3.6 ml, p < 0.001). We conclude that systolic motion of the basal RV freewall predicts global function more accurately than other 2D estimators. However, no markers summarise 3D contractile patterns, limiting their predictive accuracy.

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Introduction

Right ventricular (RV) function has prognostic value in acute [1, 2] and chronic [3–6] cardiorespiratory disease and in the peri-operative period [7–9], although rapid and accurate assessment is challenging due to the right ventricle’s complex geometry and motion [10, 11]. Several regional surrogate measures of global RV function have been investigated, most commonly targeting the contributions of longitudinal and transverse motion, because these are thought to predominate [12]. In normal subjects, longitudinal shortening may account for the majority of RV function [13] and can be measured by tricuspid annular plane systolic excursion (TAPSE). However, TAPSE may be unreliable when assessed by less experienced operators [14] and in mild and moderate RV dysfunction [15]. Prospective evaluations suggest TAPSE’s prognostic value is limited [16, 17]. Identifying the optimal two-dimensional (2D) index of systolic function is difficult as some regions are easy to identify on imaging, some are influential on ventricular function and some are affected by dysfunction. These three areas do not necessarily coincide [18] and may change because of disease processes [19]. Determining the individual contribution made to global function by the excursion of each point of the right ventricle requires a computational model of cardiac motion. Cardiac magnetic resonance (CMR) is considered the reference standard for RV volume and, recently, high-resolution three-dimensional cardiac magnetic resonance (3D-CMR) has been used for accurate segmentation of whole-heart anatomy [20–22]. The variation in regional motion within a population can then be explored using atlas-based segmentation techniques, where each subject’s images are co-registered [23, 24]. We decided to apply these techniques in order to determine which area of the right ventricle best reflects global function in the general population, and tested its predictive performance against conventional indices such as TAPSE and septum-freewall displacement (SFD).

Methods

This study was conducted at a single centre and was approved by the Hospital’s research ethics committee. Participants were recruited by advertisement and all gave written, informed consent. We did not study patients with cardiovascular disease or pregnancy, or those taking medication for hypertension, diabetes or hypercholesterolaemia. Standard published safety contraindications to MR imaging were applied [25]. The study was divided into atlas development, model development, and model validation stages (Fig. 1). Images from 40 subjects were chosen at random: 20 were used to create a cardiac atlas (10 women, age range 24–59 years), 20 were imaged twice to assess reproducibility (9 women, age range 18–54 years). Images from the remaining 600 volunteers were divided equally into a model development group and a model validation group.

The CMR examinations were performed on a 1.5 T Philips Achieva system (Philips Medical Systems International, Best, the Netherlands). In order to capture the whole-heart phenotype, a high-spatial resolution, 3D balanced, steady-state, free procession cine sequence was used that assesses the left and right ventricles in their entirety in a single breath-hold (voxel size 2 × 2 × 4 mm reconstructed to 2 × 2 × 2 mm, 60 sections, flip angle 50°, bandwidth 1250 Hz.pixel⁻¹, TE 1.5 ms, TR 3.0 ms, 20 cardiac phases, sensitivity encoding (SENSE) factor 2.0) [22]. Conventional 2D cine imaging was also performed. Images were curated on an open-source image database (MRIdb, Hammersmith Hospital Campus, Imperial College London, UK) [26].

Automated segmentation of the 3D-CMR images was used to extract the myocardial shape and quantify phenotypic parameters. Segmentation was implemented using Matlab (MATLAB and Statistics Toolbox Release 2013b, The MathWorks Inc., Natick, MA, USA) and performed on a high-performance server (Dell Technologies, Round Rock, TX, USA). The method uses prior knowledge by comparing each new set of images with a series of cardiac atlases where the anatomy has been manually defined [27]. Each subject’s dataset was co-registered to a 3D spatial template created from an average cardiac shape to ensure that every point in the statistical model had the same anatomical correspondence throughout the population. Non-rigid registration between cardiac phases was used to model RV motion [27]. Mean unsupervised automated analysis time was 46 min per 3D sequence.

Segmentation accuracy was confirmed by visual inspection of the segmented images, by voxel concordancy with manual segmentation of the same subjects and by comparison of RV ejection fraction (RVEF) measured by two readers with 2 years’ cardiac MR experience using CMRtools (Cardiovascular Imaging Solutions, London, UK) with RVEF derived from the model. Segmentation reproducibility was confirmed using repeat imaging of the same subjects.

In model development, local function was assessed as the distance travelled in 3D space from end-diastole to end-systole by each of the 11,262 points representing the RV...
wall (‘excursion’). Global function was assessed by RVEF using a method previously described. In model validation, ventricular function was assessed by four single-point measures in clinical use: TAPSE, fractional tricuspid annular plane systolic excursion (TAPSE-F), SFD and fractional septum-freewall displacement (SFD-F) and their measurement followed published methods [14, 18]. A fifth, novel measure was also taken, derived from the model development process described below. Data were analysed using R version 3.0.1 (R Development Core Team, Foundation for Statistical Computing, Vienna, Austria) [28].

Test–retest reproducibility was assessed using bias (mean and standard deviation) and intra-class correlation coefficient (ICC) with a two-way random model for absolute agreement [29]. Regional excursion was assessed for association with global systolic function using pointwise bootstrapped linear regression adjusted for age, sex, race and body surface area (BSA). Data were centred and scaled before analysis. Significance was calculated using permutation testing (10,000 permutations) with correction for multiple testing by false-discovery rate. Adjacent points on the RV wall are likely to have highly correlated function (multicolinearity), increasing the risk of unreliable predictions in new datasets. Principal component analysis (PCA) was used to overcome this problem. Principal components are orthogonal modes of variation seen in the data, from which the original dataset can be described with maximal economy. Being orthogonal, principal components are linearly uncorrelated with each other, and are therefore suitable variables for regression analysis and predictive models. The first, largest, principal component (PC1) was retained for subsequent regression. Only one component was retained to ensure estimations of the predictive strength of the 3D model were
Table 1 Baseline characteristics of healthy volunteers (n = 600) included in the discovery and validation cohorts. Values are number (proportion) or median (IQR [range]).

| Characteristic          | Value                  |
|-------------------------|------------------------|
| Race                    |                        |
| African                 | 18 (3%)                |
| Afro-Caribbean          | 30 (5%)                |
| Asian                   | 72 (12%)               |
| Caucasian               | 450 (75%)              |
| Chinese                 | 12 (2%)                |
| Other                   | 18 (3%)                |
| Sex; female             | 330 (55%)              |
| Body surface area; m²   | 1.80 (1.69–1.95 [1.34–2.57]) |
| Age; years              | 39.0 (29.5–47.6 [19.0–72.0]) |
| Right ventricular       |                        |
| ejection fraction; %    | 57 (52–62 [40–73])     |

conservative, and to allow a direct comparison with other ‘single-variable’ markers, such as TAPSE.

In model validation, each of the five 2D measures was tested by correlation, linear regression and accuracy of RVEF prediction. Correlation with RVEF was assessed using the Pearson product-moment coefficient. Difference in correlation coefficients was assessed by a method described by Steiger [30]. The association of each 2D measure with RVEF was assessed by linear regression with these variables, data were centred, scaled and coefficients bootstrapped (10,000 replications) with RVEF as the dependent variable and age, sex, race, BSA and the 2D measure of interest as independent variables (see also Supporting Information Table S1). Difference in regression models was assessed by treating the model type as an additional covariate. Right VEF prediction was tested using a leave-one-out analysis where RVEF for a single subject was predicted from the remaining 299 subjects and the process was repeated for all 300 validation subjects. Median absolute prediction errors were compared by Kruskal–Wallis rank sum test ($\chi^2$ statistic). A p value of < 0.05 was considered statistically significant.

Results

Six hundred and forty adult volunteers successfully completed the imaging protocol. All datasets were used for analysis and there were no failures of the segmentation algorithm. Baseline characteristics of participants included in the study are shown in Table 1.

No significant bias was detected between 3D automated, 3D manual and 2D manual volumetry of the right ventricle (bias $\pm$ 8 ml.m$^{-2}$ limits of agreement: $-6$ to $+22$ ml.m$^{-2}$). Intraclass correlation coefficients for test–retest reliability of the model were 0.99 for 3D automated RV volume index and 0.93 for pointwise systolic excursion. Mean ± SD point distance from automated to manual segmentation was 1.46 ± 1.46 mm (95%CI: 1.46–1.47).

Bootstrapped linear regression identified a distinct, contiguous 7.3 cm$^2$ patch in the basal free-wall where excursion was significantly associated with RVEF (Fig. 2). The patch was located between 4.1 cm and 6.9 cm from the apex and represented 9% of the total RV surface area. We

Figure 2 Regional function and principal component analysis. (a) The area where a significant association exists between regional excursion and right ventricular ejection fraction (RVEF) is shown by the coloured patch. The scale indicates the strength of correlation between patch function and patch function described by the first principal component. Overall direction of patch motion is indicated by the white arrow and comprises longitudinal, transverse and circumferential components. The intersection with the four-chamber plane (dotted line) was used to identify the location of an Optimised Single-Point Marker (SPM-O) for subsequent validation. PCA of excursion within the patch in subjects with (b) low RVEF (42%) and (c) high RVEF (74%) reveals how patch function varies with global function. Increasing the number of principal components increases the accuracy of the model, but at the expense of including the influence of noise. The vertical dashed line indicates the cumulative variance accounted for by PC1 (43%).
used PCA to create independent modes of variation within the excursion of this patch. The PC1 accounted for 43% of the observed variance in excursion and was significantly correlated with RVEF ($R = 0.34$, $p < 0.001$) and associated with RVEF in regression ($F = 40.4$, $p < 0.001$), Table 2.

Freewall function increased from apex to base and this was mostly due to an increase in longitudinal function (see also Supporting Information Figure S1). Patch function consisted of longitudinal contraction (35% of patch function, median (IQR [range]) excursion 8.9 mm (7.2–10.9 [5.7–13.2]), transverse contraction (38%, 9.7 mm, 8.5–11.1 [7.6–12.3]) and circumferential contraction (27%, 6.8 mm, 6.7–7.0 [6.6–7.0]).

To translate the automated computational findings to standard 2D analysis planes, a manual optimised single-point marker (SPM-O) measurement was derived by identifying a sub-group of voxels within the patch coinciding with conventional scanning planes. Excursion values in the patch were reconstructed using PC1 and correlated with the original excursion value. The highest correlating voxel ($R = 0.9$) coinciding with a scanning plane was identified at 43% of the distance from base to apex, in the basal freewall of the four-chamber plane. The displacement of this point from end-diastole to end-systole in millimetres was recorded manually from the four-chamber plane (Fig. 3).

Right VEF, SPM-O, and four validated markers of RV function (TAPSE, TAPSE-F, SFD, SFD-F) were evaluated prospectively in 300 subjects. All markers were significantly correlated with RVEF (Table 3). The SPM-O had the highest correlation with RVEF ($R = 0.44$), which was significantly higher than that of TAPSE ($p = 0.002$) and SFD ($p < 0.001$), and non-significantly higher than TAPSE-F ($p = 0.500$) and SFD-F ($p = 0.860$).

In regression, SPM-O accounted for the highest proportion of variance seen (19%) although differences between median (IQR [range]) were non-significant (see also Supporting Information Table S2, Fig. S2). Leave-one-out analysis found that the median (IQR [range]) absolute error in the prediction of RVEF was significantly smaller for SPM-O 2.8 (0.89–5.35 [0.00–18.75]) ml than other markers ($p = 0.003$, see also Supporting Information Table S3 and Figure S2).

**Discussion**

Several parameters have been proposed to estimate global RV function, but this is the first report of a systematic computational approach to search for the optimal measurement. Using 3D-CMR, we found that systolic displacement of the basal RV freewall is the strongest predictor of global function, outperforming TAPSE and other related indices in predicting RVEF. However, the 3D contractile pattern of the right ventricle limits the accuracy of estimators derived from 2D imaging planes. Clinicians

| Comparison | Test statistic | p value |
|------------|---------------|---------|
| Correlation RVEF ~ PC1 | $R = 0.34$ | $< 0.001$ |
| Regression RVEF ~ PC1 | $F = 40.4$ | $< 0.001$ |
| + Age + Sex + Race + BSA |       |         |
| Age        | $F = 17.9$   | $< 0.001$ |
| Sex        | $F = 8.8$   | 0.003   |
| Race       | $F = 0.6$   | 0.450   |
| BSA        | $F = 0.04$  | 0.850   |

Table 2: The association of motion in the identified patch with right ventricular ejection fraction (RVEF) as assessed by correlation and linear regression. In linear regression, variables associated with altered right ventricular ejection fraction are controlled for.

PC1, first principal component; BSA, body surface area.
managing patients at risk of RV dysfunction should be aware of these limitations in their assessments.

Right ventricular dysfunction is increasingly recognised as a significant predictor of morbidity and mortality in patients in the acute \([1, 2]\), chronic \([3–6]\) and peri-operative \([7–9]\) settings. The complexity of RV function has prompted efforts to summarise function by markers of longitudinal \([14, 15, 18]\) and transverse \([31]\) function, though these markers have weaknesses \([17, 32]\). Measures which combine both, such as fractional area change, appear to perform best but rely on accurate endocardial delineation and require suitable analysis software. These indices of RV function are derived from standard 2D imaging planes despite evidence that RV myocardial architecture is oblique to these planes, incorporates 3D contraction patterns and shows regional variation \([33, 34]\). Furthermore, global function is affected by age, sex, race, BSA and interactions between these factors \([35–37]\).

Our data show that the complex 3D contractile pattern of the right ventricle can be accurately assessed using CMR coupled with segmentation techniques, allowing consistent comparisons to be made between anatomical points across a population. This method allows modelling of the 3D contractile pattern of the right ventricle, which originates from its layered architecture with deep, longitudinal myofibrils \([11]\) contributing to long-axis excursion \([38, 39]\) and superficial myofibrils contributing to transverse and circumferential contraction \([11]\). Applying these approaches to a cohort of healthy subjects allows the contributions of these components to be assessed in the normal ventricle \([18, 19]\). We suggest that although these contributions may vary in disease processes, and particularly in response to elevated RV pressure \([19, 40–42]\), they are still useful for clinicians involved in the assessment of the right ventricle.

Principal component analysis was used to summarise wall motion as a single linear component, allowing translation to a conventional image plane and comparison with other single-point excision measures. This process retained the optimised balance of transverse and longitudinal contraction, reflected by SPM-O’s superior

| Measure | SFD | TAPSE | SFD-F | TAPSE-F | SPM-O |
|---------|-----|-------|-------|---------|-------|
| SFD     | 100 | 90    | 80    | 70      | 60    |
| TAPSE   | 80  | 70    | 60    | 50      | 40    |
| SFD-F   | 70  | 60    | 50    | 40      | 30    |
| TAPSE-F | 60  | 50    | 40    | 30      | 20    |
| SPM-O   | 50  | 40    | 30    | 20      | 10    |

Figure 4 Proportion of variance in observed right ventricular ejection fraction accounted for by regression covariates. SFD(-F), septum-freewall displacement (-fractional); TAPSE(-F), tricuspid annular plane systolic excursion (-fractional); SPM-O, statistically optimised point marker.
performance in comparison with other point markers. However, quantifying excursion in the long-axis plane fails to account for circumferential contraction, which may explain why all 2D single-point assessment methods have only a modest relationship with RVEF in the healthy population, even when optimised. The clinical value of this study is the finding that global RV function is most strongly associated with function in the basal RV freewall, but markers of function confined to two dimensions, although convenient to use and easily translated across imaging modalities, may fail to identify dysfunction orthogonal to the imaging plane.

Our study has several limitations. Work was confined to healthy volunteers with a presumed low incidence of regional wall motion abnormalities which are known to impair the performance of regional markers of global function [43]. Regional markers based on excursion take no account of abnormal diastolic function or ventricular coordination, which may exist without systolic dysfunction [44]. We evaluated this marker’s performance using CMR in order to optimise global assessment so its reproducibility and performance in other imaging modalities and patient groups requires further investigation. Lastly, we have assumed that a linear relationship exists between clinical factors (age, race, sex, BSA), local function (as indicated by single-point markers) and global function (as indicated by RVEF). Although this allows easy comparison of markers, we acknowledge that non-linear relationships may exist between single-point markers and clinical factors [45].

In conclusion, using 3D-CMR with computational analysis allows a comprehensive understanding of how each component of RV motion contributes to pump function. Systolic excursion of the basal RV freewall is a better predictor of RVEF than other established parameters, including TAPSE, as it assesses both longitudinal and transverse contraction. This finding has implications for how global function is estimated using conventional 2D imaging.

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**Supporting Information**

Additional supporting information may be found online in the Supporting Information section at the end of the article.

**Table S1.** Overview of model analysis methods in the discovery and validation cohorts.

**Table S2.** Regression parameters in the assessment of right ventricular function by two-dimensional single-point markers and by global, three-dimensional ejection fraction.

**Table S3.** Leave-one-out analysis of SPM-O and other two-dimensional markers. N = 300.
**Figure S1.** Motion characteristics of the RV freewall. A plot to show the relative contribution of circumferential, longitudinal and transverse excursion within the right ventricle from apex to base (in a plane corresponding to the four-chamber view). The dashed lines mark the zone in the freewall where global function (right ventricular ejection fraction) and local function (excursion) are significantly associated (4.1–6.9 cm, \( P < 0.05 \)).

**Figure S2.** Prediction of RVEF in model validation cohort. RVEF, right ventricular ejection fraction; SFD(-F), septum-freewall displacement (-fractional); TAPSE(-F), tricuspid annular plane systolic excursion (-fractional); SPM-O, optimised single-point marker.