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Abstract	The percutaneous inter option for patients at h heart structures to be c interpretation of the av clinical experience and	The percutaneous interventions in the treatment of structural heart diseases represent nowadays a viable option for patients at high risk for surgery. However, unlike during the traditional open heart surgery, the heart structures to be corrected are not directly visualized by the physician during the interventions. The interpretation of the available medical images is often a demanding task and needs specific skills i.e. clinical experience and complex radiological and echocardiographic analysis. The new trend for		

	cardiovascular diagnosis, surgical planning and intervention is, today, mutually connected with most recent
	developments in the field of 3D acquisition, interactive modelling and rapid prototyping techniques. This is
	particularly true when dealing with complex heart diseases since 3D-based techniques can really help in
	providing an accurate planning of the intervention and to support surgical intervention. To help the
	research community in confronting with this new trend in medical science, the present work provides an
	overview on most recent approaches and methodologies for creating physical prototypes of patient-specific
	cardiac structures, with particular reference to most critical phases such as: 3D image acquisition,
	interactive image segmentation and restoration, interactive 3D model reconstruction, physical prototyping
	through additive manufacturing. To this purpose, first, recent techniques for image enhancement to
	highlight anatomical structures of interest are presented together with the current state of the art of
	interactive image segmentation. Finally, most suitable techniques for prototyping the retrieved 3D model
	are investigated so as to derive a number of criteria for manufacturing prototypes useful for planning the
	medical intervention.
Keywords (separated by '-')	Rapid prototyping - 3D modelling - Medical imagery - Heart - Cardiovascular diseases - Surgical planning
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ORIGINAL PAPER



3D printing of cardiac structures from medical images: an overview of methods and interactive tools

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Abstract The percutaneous interventions in the treatment of structural heart diseases represent nowadays a viable 2 option for patients at high risk for surgery. However, unlike 3 during the traditional open heart surgery, the heart structures Δ to be corrected are not directly visualized by the physician 5 during the interventions. The interpretation of the available medical images is often a demanding task and needs specific skills i.e. clinical experience and complex radiological and 8 echocardiographic analysis. The new trend for cardiovascu-9 lar diagnosis, surgical planning and intervention is, today, 10 mutually connected with most recent developments in the 11 field of 3D acquisition, interactive modelling and rapid pro-12 totyping techniques. This is particularly true when dealing 13 with complex heart diseases since 3D-based techniques can 14 really help in providing an accurate planning of the interven-15 tion and to support surgical intervention. To help the research 16 community in confronting with this new trend in medical 17 science, the present work provides an overview on most 18 recent approaches and methodologies for creating physical 19 prototypes of patient-specific cardiac structures, with par-20 ticular reference to most critical phases such as: 3D image 21 acquisition, interactive image segmentation and restoration, 22 interactive 3D model reconstruction, physical prototyping 23 through additive manufacturing. To this purpose, first, recent 24 techniques for image enhancement to highlight anatomical 25 structures of interest are presented together with the current 26 state of the art of interactive image segmentation. Finally, 27 most suitable techniques for prototyping the retrieved 3D 28

Monica Carfagni monica.carfagni@unifi.it model are investigated so as to derive a number of criteria29for manufacturing prototypes useful for planning the medical30intervention.31

KeywordsRapid prototyping · 3D modelling · Medical32imagery · Heart · Cardiovascular diseases · Surgical planning33

1 Introduction

Thanks to the outstanding advances in both paediatric cardi-35 ology and cardiac surgery, the number of children affected by 36 congenital heart disease (CHD) surviving into adulthood con-37 tinues to increase [1]. Therefore, adults with CHD represents 38 an ever-growing population, of around 3 million patients in 39 Europe [2], with continues to increase by 5% per year. It 40 is therefore not surprising that the care, and management, 41 of adult patients with congenital or acquired structural heart 42 disease represents one of the most relevant areas of research 43 in cardiology, documenting a grow of studies related to this 44 vital area [3–5]. 45

Given the widely-ranged complexity of possible struc-46 tural heart defects, imaging is paramount in their treatments. 47 Although two-dimensional (2D) imaging modalities such as 48 echocardiography, computed tomography (CT), and mag-49 netic resonance imaging (MRI) are undeniably valuable in 50 the evaluation of adult patients with structural heart disease, 51 these methods are still constrained by their overall lack of 52 realism and inability to be "physically manipulated"; thereby, 53 such techniques remain limited in their ability to effec-54 tively represent the complex three-dimensional (3D) shape 55 of the heart and its peripheral structures. Quite the oppo-56 site, 3D medical data representation, obtained from medical 57 imagery, has the potential of providing information concern-58 ing heart structure, giving at the same time the opportunity 59

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Fig. 1 Patient-specific 3D modelling and printing workflow

of further investigations when combined with additive man ufacturing methods and technology. Recently, a number of
 most advanced hospitals [6–8] introduced in their praxis the
 use of physical 3D models of the patient cardiac structure to
 be used for surgical planning or for training.

These few cutting-edge experiences suggested that the 65 progress of 3D printing technology sector can provide a num-66 ber of advanced tools helping the medical staff to confront 67 with care and management congenital or acquired structural 68 heart diseases. In fact, the possibility of creating a patient-69 specific tangible 3D fabricated model provides medical staff 70 with information that goes beyond a simple 3D-shaded visu-71 alization on a flat screen. The "zero lead time" between 72 design and final production of accurate models together with 73 the possibility of creating specific models resembling the 74 actual structure of the patient heart accelerate the recent 75 medical trend towards "personalized" or "patient-specific" 76 treatment. 77

According to recent literature [9,10], the most effective way for creating 3D models starting from 2D medical imaging is based on a virtuous process cycle that usually starts from 2D and 3D medical image acquisition and ends with 3D printing of a model of the patient heart (see Fig. 1).

In detail, such a process starts from medical imagery with 83 particular reference to (but not exclusively) CT, multi-slice 84 CT (MCT) and MRI. All these methods basically [11-85 14] provide 3D information in the form of cross-sectional 86 "slices" of the entire body or parts of interest. Such slices 87 consist of digital pictures stacked through the measurement 88 volume, where each pixel represents the spatial position of 89 the imaged element and its colour (usually in grey values) 90 the reaction of the tissue to radiation or magnetic field (in 91

case of CT and MRI respectively). Acquired images are 92 then processed, usually by using semi-automatic and inter-93 active methods, in order to segment regions of interest, e.g. 94 heart chambers, valves, aorta and coronary vessels. These 95 segmented areas are, then, converted into 3D models, using 96 tools like volume rendering or surface reconstruction proce-97 dures. This process is very useful in clinical practice, since 98 it allows for interactive and easy visualization of differ-99 ent tissues and anatomical structures. Surface reconstruction 100 techniques provide 3D polygonal mesh model of the exter-101 nal surface of the part. The obtained 3D surface models 102 are useful for performing dimensional verification (volume 103 computation, thickness and centreline of blood vessels calcu-104 lation, etc.), structural analysis (e.g. by using finite elements 105 FE) and computational fluid dynamics (CFD) studies. More-106 over, the availability of topologically correct and optimized 107 surface models allows the manufacturing of a prototype of 108 cardiac structure, using for instance additive manufacturing 109 techniques. 110

Due to the increasing number of methods to comply 111 with the above mentioned process, and since such a virtu-112 ous process has been only recently introduced in clinical 113 practice, the main aim of the present work is to provide 114 an overview of methodologies dealing with patient-specific 115 3D modelling of cardiac structures. First, main imaging sys-116 tems medical imagery for acquiring 2D and 3D data inferred 117 to heart structure are introduced (Sect. 2). Then, the most 118 adopted algorithms for image enhancement and restoration 119 are explored (Sect. 3) and an overview of interactive seg-120 mentation and classification algorithms is described (Sect. 4). 121 Section 5 is devoted to briefly overview most promising tech-122 niques for 3D heart model reconstruction process. Finally, in 123

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126 2 Medical imaging

Since the discovery of X-rays by Wilhelm Conrad Rontgen 127 in 1895, medical images have become a major component of 128 diagnostics, treatment planning and procedures, and follow-129 up studies. Moreover, medical images are increasingly used 130 for education, documentation, and research since they pro-131 vide description of morphology as well as physical and 132 biological functions in 1D, 2D, 3D, and even 4D image data 133 (e.g., cardiac MRI, where up to eight volumes are acquired 134 during a single heart cycle) [15]. Today, a large variety of 135 imaging modalities have been established; these are mainly 136 based on transmission, reflection or refraction of light, radi-137 ation, temperature, sound, or spin. Most imaging techniques 138 are non-invasive, allowing to seek for internal structures of 139 the patient's body hidden by skin and bones. The output of 140 the imaging acquisition process and input of the rapid proto-141 typing following appropriate processing is a DICOM image 142 (Digital Imaging and Communications in Medicine), which 143 is the, traditional, virtual outcome of all medical professions 144 utilizing images [16]. 145

Over the years, a number of different medical imaging
modalities has been developed, using different technologies
and creating several kinds of images. Common types of medical imaging include: (i) X-ray (radiography, CT, etc.), (ii)
MRI, (iii) ultrasound, (iv) nuclear medicine (e.g., positron
emission tomography—PET).

152 2.1 Magnetic resonance imaging (MRI)

MRI is an imaging technique based on detecting different 153 tissue characteristics by varying the number and sequence 154 of pulsed radio frequency fields, taking advantage of the 155 magnetic relaxation properties of different tissues [17]. MR 156 imaging has the crucial advantage of not emitting X-ray radi-157 ations. Instead, the MR scanner provides a strong magnetic 158 field, which causes protons to align parallel or anti-parallel 159 to it. MR measures the density of a specific nucleus, nor-160 mally hydrogen, which is magnetic and largely present in 161 the human body, including heart [18], except for bone struc-162 tures. The speed at which protons lose their magnetic energy 163 164 varies in different tissues allowing detailed representation of the region of interest. This measurement system is volumet-165 ric, producing isometric 3D images (i.e. the same resolution 166 in all directions). 167

168 2.2 Computerized tomography (CT)

As widely known, CT combines many X-ray images, taken
 from different angles, to produce cross-sectional (i.e. tomo-

graphic) images of patient areas. Especially in the recent 171 advances, CT can provide detailed anatomical information of 172 chambers, vessels, coronary arteries, and coronary calcium 173 scoring. In particular, two imaging techniques are typically 174 carried out in cardiac tomography: (1) coronary calcium scor-175 ing with non-contrast CT and (2) non-invasive imaging of 176 coronary arteries with contrast-enhanced CT. Usually, non-177 *contrast CT* imaging exploits the natural density of tissues. 178 As a result, various densities using different attenuation val-179 ues such as air, calcium, fat, and soft tissues can be easily 180 distinguished. Non-contrast CT imaging is a low-radiation 181 exposure method within a single breath hold, determining 182 the presence of coronary artery calcium. 183

In comparison, contrast-enhanced CT (see Fig. 2) is used 184 for imaging of coronary arteries with contrast material such 185 as a bolus or continuous infusion of a high concentration 186 of iodinated contrast material. Furthermore, coronary CT 187 angiography has been shown to be highly effective in detect-188 ing coronary stenosis. Moreover, coronary CT angiography 189 can visualize not only the vessel lumen but also the vessel 190 wall, allowing non-invasive assessment of the presence and 191 the size of the non-calcified coronary plaque. 103

2.3 Ultrasound

In contrast to CT and MRI, ultrasound is a medical imaging 194 modality that is based on reflection of sound waves. Depend-195 ing on the transducer, 1D to 4D data is obtained. In the 1D 196 case (signal), a longitudinal sound wave is traveling through 197 the tissue of the human body. At transitions between differ-198 ent matter (e.g., muscle and fat), the sound wave is partly 199 reflected and transmitted (refracted if the surface is not hit 200 perpendicular). 201

Therefore, the echo runtime indicates the distance between 202 transducer and tissue border while the echo strength is 203 related to material properties. For cardiac usage, ultrasound is 204 applied by means of an echocardiogram able to provide infor-205 mation on the four chambers of the heart, the heart valves and 206 the walls of the heart, the blood vessels entering and leaving 207 the heart and the pericardium. Echocardiography was, for a 208 long time, limited to a 2-D examination. As 3D ultrasound 209 is being more widely studied, new applications for imaging, 210 visualization and quantification of the heart are emerging. As 211 3D ultrasound images (see for instance Fig. 3) are typically 212 acquired in smaller sectors, to maintain adequate spatial and 213 temporal resolution, registration can be used to fuse multiple 214 3D sectors together; this extends the field of view, allow-215 ing the quantification of larger structures while preserving 216 resolution [19]. 217

2.4 PET

Positron emission tomography (PET) is a non-invasive medical imaging modality that produces functional rather than 220

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Fig. 2 CT scan with contrast in diastolic phase: a axial view; b sagittal view; c coronal view; d rough 3D reconstruction from originally acquired images (i.e. without post processing phases), demonstrating the necessity of enhancing and restoring the input image prior to proceed with 3D model reconstruction



Fig. 3 Ultrasound image of a patient heart

anatomical images, thereby providing greater insight to the 221 patient's condition. Functional imaging means that the body's 222 metabolism is being shown, as opposed to its structure. In a 223 cardiac PET scan, the patient is administered with a drug that 224 creates an effect in the body that is similar to exercise. Images 225 of the myocardial blood flow are taken both before and after 226 the induced stress and the images are compared to examine 227 blood flow through the heart to determine if narrowing of the 228 coronary arteries is restricting maximum blood flow [20]. 229 PET can also identify dead tissue and injured tissue that's 230 still living and functioning. 231

2.5 Remarks

CT. PET. MRI provide, as said before, 3D information in the 233 form of thin cross-sectional "slices" of the entire body or parts 234 of interest, and are, therefore, called tomography imagery. 235 The slices are provided in the form of digital pictures stacked 236 through the measurement volume, where each pixel repre-237 sents the spatial position of the imaged element and its colour 238 (grey or RGB values) the reaction of the tissue to radiation 239 (X-rays or gamma-rays) or magnetic field, in case of CT, PET 240 and MRI respectively. Each of the above-mentioned imaging 241 methods has its own advantages and disadvantages, both for 242 potential patient's health hazard (some of them require expo-243 sure to radiation) and information gathered and displayed. 244 For example, CT and MRI are sensitive to different tissues 245 properties, so that soft tissues are poorly visible in CT images, 246 while they are distinctly evident in MRI. 247

3 Image restoration and enhancement

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and metallic implants cause imaging artefacts in MRI [21]). 254 A huge number of methods for enhancing the quality and for 255 restoring acquired data have been proposed in literature. To 256 enhance the visual quality of medical images, the two main 257 procedures are image restoration and image enhancement. 258

Image restoration algorithms act to denoise the images 259 also by reducing the blur effect due to the data acquisition 260 pipeline [22]. Denoising methods are based on the estima-261 tion and modelling of the blur and noise affecting the image. 262 The main factors that influence the models include capturing 263 sensors system, transmission media, image digitization, etc. 264

Enhancement belongs to the branch of digital image pro-265 cessing techniques that manipulate a digital image [23-25] to 266 enhance the contrast, in order to extract, or accentuate, certain 267 image features to improve the understanding of information 268 content and obtain an image ready for automated image pro-269 cessing (e.g. for highlighting structures such as tissues and 270 organs). 271

As mentioned above, restoration and enhancement meth-272 ods for medical images depends on the imaging system. 273 Therefore, an overview of most relevant methods is provided below.

3.1 Enhancing and restoration of MRI images 276

The main issue for cardiac images acquired using MRI is 277 related to intensity non-uniformity, i.e. a smooth intensity 278 variation across the image, resulting in biased intensities for 279 the same tissue according to its location. Barely noticeable to 280 a human observer, this distortion influences the subsequent 281 steps of segmentation and classification. In fact, most of the 282 automated, or interactive, quantitative methods for segmen-283 tation rely on the assumption that a given tissue has a similar 284 voxel intensities throughout the data, so these methods are 285 highly sensitive to variations on image intensities [26-28]. 286

Numerous methods were developed to solve the intensity 287 non uniformity problem; roughly these methods can be clas-288 sified into three main groups: 289

1) Filtering methods: assumes that the non-uniformity is 290 a low-frequency artefact that can be separated from 291 the high-frequency signal of the imaged object by a 292 low-pass filtering. Two main approaches can be fol-293 lowed to filter the original image: homomorphic filtering 294 [29] and homomorphic unsharp masking (HUM) [30]. 295 These methods are affected by other image features, such 296 as edge effects, most present in high contrast images. 207 Accordingly, some methods have been proposed to min-298 imize these effects [26]. 299

2) Surface fitting models. Since intensity non-uniformity is 300 slowly varying, it is reasonable to approximate it by a 301 parametric smooth function. As a consequence, the image 302 correction can be addressed by dividing (voxel-by-voxel) 303

the original image by the computed surfaces. The differ-304 ent algorithms using surface fitting models vary in the 305 way the fitting is performed; moreover, these methods 306 are linked to image segmentation, which leads to frame-307 works that simultaneously correct the non-uniformity and 308 perform the segmentation [31]. 309

3) Statistical methods. The statistical methods assume that 310 the intensity non-uniformity follows a statistical distribu-311 tion. In fact, Edelstein et al. [32] showed that pure noise in 312 magnitude images is governed by the Rayleigh distribu-313 tion. As a consequence, it is possible to apply well-known 314 look-up table correction schemes like the ones proposed 315 in [33,34]. These methods are among the most used when 316 dealing with cardiac MRI images. 317

3.2 Enhancing and restoration CT images

The low contrast of CT images due to the large dynamic 319 range acquisition (usually 16 bit) often does not allows to 320 properly discriminate acquired regions of interest, especially 321 when dealing with cardiac structures. The most common 322 method for enhancing the quality of the image, and at the 323 same time to better discriminate possible presence of patholo-324 gies, is to perform a histogram equalization (HE) followed 325 by smoothing filters as demonstrated, for instance, in Fig. 4 326 [35]. The main feature in HE is to find the mapping function 327 generated from the global histogram of an image, to obtain 328 a uniform distribution in the corresponding output image. 329 Unfortunately, traditional HE technique may introduce unde-330 sired effects such as level saturation or over-enhancement. To 331 solve this issue, a range of methods are proposed in litera-332 ture. Most relevant methods are adaptive HE method [36] 333 and Bi-histogram equalization [37]. Fuzzy Logic and other 334 IA-based methods have been also explored to remove image 335 artefacts [38]. 336

3.3 Enhancing and restoration of ultrasound images

Ultrasound images have, in general, low signal to noise ratio 338 (SNR) mainly due to pulsed signals that have short dura-339 tion in time and therefore a broad spectrum in frequency 340 domain [39]. Moreover, the coherent nature of the signals 341 gives rise to speckle noise; finally, sound waves are highly 342 distorted when traveling through the tissues. Consequently, 343 the acquired images are speckled and present multiple defects 344 such as, for instance, reflections represented by parallel bright 345 lines or by the so called "comet tail". Most errors are strongly 346 avoidable by pre-processing techniques in ultrasound imag-347 ing. However, the acquired image is still to be enhanced. 348 To this purpose a large number of methods have been pro-349 posed in literature, roughly classified in filtering methods and 350 deconvolution methods. Since noise in ultrasound has been 351 modelled as the combination of an additive error (e.g. elec-352

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Fig. 4 a Original image; b restored image obtained by applying a histogram equalization followed by a Gaussian filter with a sigma value (SD) equal to 0.6 mm

tronic or thermal noise) with a multiplicative error (speckle),
filtering algorithms are required to be adaptable to the noise
properties across the image. Consequently, ultrasound filtering techniques include adaptive filters based on local
statistics [40], anisotropic diffusion [41] and wavelets [42].
Differently from filtering approaches, deconvolution is a

technique used to improve resolution in ultrasound images by 359 counteracting the effect of the Point spread function (PSF) of 360 the imaging device i.e. the response of the system to an ideal 361 point target caused by diffraction spreading of the ultrasound 362 signals [43]. If the PSF is known, its effects on the image can 363 be eliminated by dividing the spectrum of the image by the 364 spectrum of the PSF. In case the PSF is not known, it can be 365 estimated based on the image itself [44] (this is called "blind 366 deconvolution"). 367

368 3.4 Enhancing and restoration of PET images

PET image artefacts are primarily due to metallic implants, 369 use of contrast media and respiratory motion. In particu-370 lar, metallic implants (e.g. dental implants or metallic clips) 371 are visualized by CT images as areas of high density [45]. 372 Analogously, the presence of a contrast media can intro-373 duce errors in PET images similar to the ones produced 374 by metallic implants. Both these errors, however, can be 375 addressed directly by the medical staff by contouring the 376 contrast regions and by excluding them from the image anal-377 ysis [46]. 378

Respiratory motion, instead, can induce an erroneous 379 attenuation correction: because of respiratory motion the 380 density of a particular organ could be attributed to an area 381 whose density is different. This issue is far to be solved today 382 since the best way to correct for respiratory motion would be 383 to acquire gated images to discriminate different intervals 384 of a breath cycle. Therefore, for examining PET images it is still crucial the help from medical staff and only a few 386 approaches for interactive or automatic image enhancement 387 are nowadays available [47]. 388

4 Segmentation and classification

Segmentation is the process of dividing an image into a 390 set of semantically meaningful, homogeneous, and non-391 overlapping regions with similar properties such as grey 392 level, colour, texture, brightness, and contrast [48]. In med-393 ical imagery, the segmentation consists in identifying and 394 subdividing different anatomical structures or regions of 395 interest (ROI) in the images resulting in a partition of the 396 image pixels in non-overlapping regions, belonging to the 397 same tissue class. 398

Automatic segmentation of medical images is a valuable 399 tool to perform a tedious task with the aim of making it faster 400 and, ideally, more robust than manual procedures. However, 401 it is a difficult task as medical images are complex in nature 402 and often affect by intrinsic issues such as mixing up of tissue 403 types, biased intensity, presence of artefacts, and closeness, 404 in grey values for different tissues. As demonstrated in [49] 405 a number of different approaches have been developed for 406 automatic image segmentation that is still a current and active 407 area of research. 408

Unfortunately, anatomical variability and intrinsic image 409 issues limit the reliability of fully automatic approaches. 410 Moreover, there is a substantial mistrust both from patients 411 and doctors towards fully automatic algorithms. Accord-412 ingly, there has been a recent drive towards interactive 413 segmentation [50]. Interactive approaches use a data-driven 414 automatic algorithm to process a majority of the volume. 415 As the automatic segmentation runs and displays the current 416 state, a human user can influence the algorithm's behaviour 417 to more closely align with an expected result [51]. 418

Interactive medical image segmentation employs software tools such as, for instance, Seg3D [52] and 3D-Slicer [53] for applying algorithms and, at the same time, visualizing the results (see Fig. 5). Another class of interactive methods makes use of active contour models, contour interpolation algorithms or their combination [54]. Graph Cuts [55] and CO3 approaches [56] are considered, further-

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Fig. 5 Example of threshold-based interactive segmentation of a single slice for the aorta and left ventricle: a axial view; b sagittal view; c coronal view; d 3D reconstruction from restored images

⁴²⁶ more, golden standard techniques for performing interactive⁴²⁷ segmentation.

(c)

Referring specifically to cardiac imaging, several segmentation techniques have been further proposed [57]. Most of the more recent studies focused on the segmentation of the diastolic and systolic left ventricle, since the related volume functional parameters are important predictors of heart diseases [58].

Schneider et al. [59] proposed a method to segment the 434 Mitral annulus using graph-cuts. Zheng et al. [60] presented a segmentation method based on the marginal space learning 436 by searching for the optimal smooth surface. Jolly [61] pro-437 posed a method for extracting the myocardium in 4-D cardiac 438 MR and CT images using the graph-cuts as well as EM-based 439 segmentation. Model-based techniques were also adopted for 440 cardiac MRI or CT image segmentation using active shape 441 models with PCA [62], atlas-based segmentation [63,64], 442 methods for region growing [65,66], thresholding [67,68] 443 and machine learning approach using decision forests [69] 444 were also used. 445

5 3D heart model reconstruction for rapid prototyping

As shown in [70], the generation of a 3D cardiac geometry represented by a 3D surface mesh can be considered the very first step of the construction process of a 3D cardiac model. It 450 has to be considered that the reconstructed geometry includes 451 one or several cardiac chambers (LV, bi-ventricular, atrial or 452 whole-heart models) and also other details such as the great 453 cardiac vessels including "outflow and/or inflow tracts [71-454 73], the fibrous annulus of atrioventricular valves [74,75], 455 part of the coronary tree, or some endocardial details such 456 as papillary muscles and trabeculae carneae for ventricles 457 or crista terminalis, pectinate muscles and fossa ovalis for 458 atria [76–78]". Moreover, it is important to highlight that 459 the anatomical accuracy required by a 3D cardiac model 460 depend on its final application. For instance, in [79] it was 461 concluded that structurally simplified models (without endo-462 cardial details or vessels) are well suited for a large range of 463 3D cardiac modelling applications aimed at EP simulation. 464 Consequently, depending on the particular kind of applica-465 tion of the 3D model, the main aim of 3D reconstruction is to 466 provide an accurate mesh after 3D image segmentation has 467 been carried out according to one of the methods described 468 in Sect. 4. 469

(d)

A surface model could be generated by using, for instance, a marching cube method [80] or other 3D contour extraction algorithms [81]. The resultant surface can be used as the starting point for either generation of higher order representation, such as non-uniform rational B-splines NURBS-based surfaces, or for meshing improvement using, for example, mesh

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methods [87,88]. 478 However, such an automatically retrieved 3D model is 479 not suitable, as it stand, for 3D printing; this is due to 480 a number of reasons such as, for instance, the presence of many separate mesh units and/or incomplete topologi-482 cal structure. Therefore, topological correction, decimation, 483 Laplacian smoothing, and local smoothing [89,90] are usu-484 ally needed to create a 3D model ready for 3D printing. In 485 general, the accuracy of the 3D printing object depends on 486 the combination of the accuracy of the medical image, which 487 should be as thin as possible, the appropriate imaging pro-488 cess for 3D modelling, and the 3D printing accuracy of the 489 system. 490

One major challenge faced in creating physical models 491 lies in disconnection between the digital 3D surface models 492 and the original 2D image. Currently available industry spe-493 cific image-processing software applications remains limited 494 in its ability to generate digital 3D models that are directly 495 applicable to rapid prototyping. As a result, true integration 496 of the raw 2D image data into the generated digital 3D sur-497 face models is lost. The post 3D processing (i.e., correction 498 of errant points and elimination of various artefacts within 499 the digital 3D surface model) therefore relies heavily on the 500 expert clinical and anatomic knowledge of the graphic editor, 501 especially because a wide array of structural heart anomalies 502 that significantly deviate from conventional cardiovascular 503 anatomy may be present. Once the surface model of the car-504 diac structure is interactively built, it is possible to manually 505 add to the 3D model some important features such as, for 506 instance, myocardial structure, cardiac conduction system 507 and other patient-specific features (e.g. pathologies). 508

It is worth noting that the 3D reconstruction needed for the 509 subsequent 3D printing process requires a continuous inter-510 action between the users (engineers and medical staff) and 511 the CAD software package used to perform the reconstruc-512 tion. Without this interaction it is quite unfeasible to extract 513 significant information from 3D images to create a complex 514 3D model of the anatomical part to be investigated. On a 515 "higher" level, the proposed framework of Fig. 1 involves 516 methods to virtually explore different solution spaces. In fact, 517 once the images are acquired using one of the methods pro-518 posed in Sect. 2, different data can be reconstructed on the 519 basis of which anatomy has to be modeled for medical pur-520 poses. Accordingly, users are required to virtually explore a 521 number of possible CAD solutions by interacting with the 522 acquired and pre-processed 3D data. 523

Another option to add myocardial structure is to esti-524 mate the fibre orientation associated to each element of the 525 volumetric mesh of a model from pre-established patterns 526 [91–94], most of them derived from Streeter's findings [95]. 527 The final result of the 3D reconstruction consists of a 3D 528

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model representing the cardiac structure to be printed. As a 520 consequence, despite several file formats exist, the model is 530 almost universally stored in standard tessellation language 531 (STL) format. 532

6 Additive technologies and 3D printing for cardiac structures

The most common additive technologies that can be used 535 for prototyping cardiac structures are selective laser sin-536 tering, fused deposition modelling, multijet modelling/3D 537 printing, and stereo-lithography. Selective laser sintering (3-538 D Systems Inc., Rock Hill, SC) uses a high-power laser to 539 fuse small particles of plastic, metal, or ceramic powders 540 into a 3D object [96]. Selective laser sintering has the abil-541 ity to utilize a variety of thermoplastic powders and has a 542 high geometric accuracy but is generally higher in cost than 543 other additive methods. In fused deposition modelling (e.g. 544 from Stratasys Inc.), a plastic filament (typically acrylonitrile 545 butadiene styrene polymer) is forced through a heated extru-546 sion nozzle that melts the filament and deposits a layer of 547 material that hardens immediately on extrusion. A separate 548 water-soluble material is used for making temporary support 549 structures while the manufacturing is in progress. The pro-550 cess is repeated layer by layer until the model is complete. 551 Multijet modelling or 3D printing (Z Corporation, Burling-552 ton, Mass) essentially works like a normal ink-jet printer but 553 in 3D space. In this process, layers of fine powder (either 554 plaster or resins) are selectively bonded by printing a water-555 based adhesive from the ink-jet print head in the shape of each 556 cross section as determined by the computer-aided design 557 file. Each layer quickly hardens, and the process is repeated 558 until the model is complete [97]. 559

In stereolithography, models are built through layer-by 560 layer polymerization of a photosensitive resin. A computer-561 controlled laser generates an ultraviolet beam that draws on 562 the surface of a pool of resin stimulating the instantaneous 563 local polymerization of the liquid resin in the outlined pattern. 564 A movable platform lowers the newly formed layer, thereby 565 exposing a new layer of photosensitive resin, and the process 566 is repeated until the model is complete. 567

Depending on their intended application (i.e. education, 568 catheter navigation, device sizing and testing, and so on), 569 physical models may be printed in multiple materials using 570 a variety of 3D printing technologies, each with its own col-571 lection of benefits and shortcomings. For example, multijet 572 modelling (see Fig. 5) technology can be used to gener-573 ate full-colour models to highlight anomalous structures or 574 specific regions of interest. Printing times are fast (approxi-575 mately 6–7 h per model) and cost-effective (Fig. 6). 576

Although flexible models may be prototyped by multi-577 jet modelling technology, the properties of the material often 578 2

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Fig. 6 Two examples of 3d printed models: on the *left* it is shown the model prototyped using multijet printer; on the *right*, the model realized using polyjet matrix technology is depicted

fail to accurately mimic true tissue properties. PolyJet Matrix 570 printing technology (see Fig. 5) offer the ability to print 580 physical models in materials that more closely resemble the 581 properties of native tissue, thus representing the new direc-582 tion in rapid prototyping technology with its ability to print 583 in different materials simultaneously. This unique technol-584 ogy will allow most physical models to be printed in durable materials (e.g., plastic), whereas specified segments (e.g., 586 interatrial septum, septal defects, vascular structures, and so 587 on) are printed in less durable, but more lifelike, materials 588 (e.g. rubber polymers) for more realistic manipulation. 589

Given the broad spectrum of anatomic variations and 590 pathologies, 3D printing could potentially be a game-changer 591 in cardiology, particularly in challenging anatomies and rare 592 pathologies, facilitating procedural planning, optimal sizing 593 and simulation. The knowledge of the exact dimensions of the 594 defect/communication/orifice that needs to be sealed and the 595 relationship with adjacent structures is of paramount impor-596 tance and often not readily available with conventional 2D 597 imaging. 3D printing allows for direct visualization and sim-598 ulation (trial and error) in order to identify the optimal device 599 and angulation required to obtain the best possible result. 600 The value of 3D printing in surgical pre-procedural planning 601 lies primarily in the precise delineation of the underlying 602 anatomy and is of paramount importance in complex congen-603 ital heart disease pre-procedural planning. Patient-specific 604 implants and custom-made devices can be designed, pro-605 duced, and tested, thus opening new horizons in personalized 606 patient care and cardiovascular research. 60

A taxonomy of possible applications of 3D printing for cardiac structures can be found in [9] and mostly referred to paediatric and adult heart diseases [98,99] and to aorta and great vessels pathologies [100, 101]. With reference to struc-611 tural heart diseases, most relevant applications of 3D printing 612 are related to the creation of patient-specific devices for clos-613 ing the left atrial appendage [98] or for determining the ideal 614 resection lines of the aneurysmectomy for patients affect-615 ing by cardiac aneurysms [102]. Other studies report the use 616 of 3D models for myectomies [103], atrial and ventricular 617 septal defects [104]. Referring to aorta and vessel patholo-618 gies, a number of studies have been carried out to create 3D 619 printed models of systemic vasculature; as stated in [9], in 620 fact, "Rigid and flexible 3D-printed models can aid endovas-621 cular interventions by enabling the assessment of optimal 622 stent dimensions and positioning in cases of transverse aortic 623 arch hypoplasia". Finally, 3D modelling of heart can be really 624 helpful for education, training and decision-making [9, 105]. 625 Training on 3D printed models can be performed virtually 626 anywhere, avoiding the cost and complexity of operating in 627 the controlled environments required for animals and human 628 cadavers. 629

7 Discussion and conclusions

With the development of inexpensive 3D printers, 3D print-631 able multi-materials, and 3D medical imaging modalities, 3D 632 printing medical applications for hearth diseases among oth-633 ers, have come into the spotlight. Due to the availability of 634 transparent, full-coloured, and flexible multi-materials, 3D 635 printing objects can be more realistic, miming the properties 636 of the real body; i.e., not only hard tissue alone but also hard 637 and soft tissue together. 638

The study and experimentation conducted show that inter-639 active approaches for image enhancement, segmentation, 3D 640 modelling and printing have reached a promising point for 641 medical applications. Moreover, the design framework intro-642 duced in Fig. 1 can be extended to other problems related to 643 patient specific modelling since almost all medical interven-644 tions share techniques based on medical image acquisition 645 and 3D reconstruction for diagnosis, education or inter-646 vention planning. It is therefore straightforward to add 3D 647 printing methods to this traditional workflow, when the com-648 plexity of the anatomy to be treated requires physical models 649 to be managed by medical staff. Some relevant examples can 650 be found in hip and knees prosthesis design as well as in max-651 illofacial sector and in personalized hand-wrist-arm orthosis 652 design. More in general, several different application areas 653 can be interested in wide spreading 3D printing of cardiac 654 structures. First, 3D printing can be really useful for aiding 655 intervention planning, giving the medical staff the opportu-656 nity of handle a realistic and patient-specific model prior to 657 proceed with the actual intervention. Secondly, the creation 658 of 3D models is crucial for training; in fact, this can be per-659 formed virtually anywhere, avoiding the cost and complexity 660

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of operating in the controlled environments required for ani-661 mals and human cadavers. Finally, 3D modelling of heart 662 can be really helpful for decision-making [9,105] giving the 663 medical staff the opportunity to test several different options 664 specifically addressed to the patient own anatomy. 665

While the opportunity of using 3D printing for simulating medical surgery has been widely accepted in the medical 667 field, for training purposes only a few studies are related to 668 the impact on learning, all recalled in a review work [106] 669 where authors demonstrate that despite there is no solid evi-670 dence that the use of 3D models is superior to traditional 671 teaching, more studies are still needed to examine the short-672 and long-term impacts of 3D models on learning using valid 673 and appropriate tools. 674

Development and optimization of the entire procedure, 675 from image acquisition to 3D printing fabrication, are 676 required for personalized treatment, even in emergency situ-677 ations. In addition, to produce an effective 3D printing object, 678 multidisciplinary knowledge of the entire 3D printing process 679 chain is needed; namely, image acquisition using a proto-680 col suitable for 3D modeling, post-processing of the medical 681 images to generate a 3D reconstructed model, 3D printing 682 manufacturing with an appropriate 3D printing technique, 683 and post-processing of the 3D printing object to adapt it for medical use. 685

On the other hand, several major limitations, such as 686 those associated with the technology and the time and cost 687 of manufacturing 3D phantoms, remain to be overcome. In 688 fact, during the virtual exploration of solutions for the 3D 680 reconstruction the presence of artifacts could lead to erro-690 neous medical interpretations. Therefore, the final model 691 could provide wrong indications on geometry or, worst, on 692 pathologies. To avoid these errors, that could strongly impact 693 on the intervention planning or even on the actual med-694 ical surgery, the intervention of experts, interacting with 695 CAD systems and/or with image processing-based algo-696 rithms is still needed to assess the accuracy of the final 697 products and correct for errors that may occur during the 698 automated phases of the procedure. Finally, an expanded heart team including physicians and engineers is needed to 700 assess the correctness of the products and correct inaccura-701 cies of automatic procedures, when translating the medical 702 imaging into 3D printing. This opens, for the next future, 703 the opportunity of creating new professionalism across med-704 ical science and engineering to confront with the incredible 705 boost of methods and techniques related to 3D printing in 706 biomedicine. 707

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