# Looking Forward with

minimally invasive ultrasound Jovana Janjic

# Looking Forward with Minimally Invasive Ultrasound

Jovana Janjic

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# Looking Forward with Minimally Invasive Ultrasound

Vooruitkijken met minimaal-invasief ultrageluid

Thesis

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In memory of my grandfather Drago Janjic who thought me that learning is a lifestyle

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# Chapter 1

Introduction

#### **1.1 MINIMALLY INVASIVE PROCEDURES**

Surgical procedures have evolved over the centuries with the first successful open heart surgery using a heart-lung machine in 1953 by John H. Gibbon [1]. Since then surgery has become safer and more effective thanks to the continuous development of medical instrumentation and improved medical knowledge. A great step towards safer interventions has been achieved with the introduction of minimally invasive procedures, where small incisions are used to access the inside of the human body. Narrow surgical instruments such as trocars, needles and catheters are then advanced to examine and treat internal tissues. In the first half of the twentieth century biopsies of cancerous tissues in different organs were accomplished using fine needles [2]. In the same period minimally invasive procedures for cataracts, spine, knee, hip and brain surgery were developed and laparoscopic surgeries on the gall bladder were successfully accomplished [3,4]. Towards the end of the twentieth century, angioplasty was implemented to treat atherosclerotic lesions in the human arteries [5]. Angioplasty employs a tiny catheter that is commonly inserted through the femoral or radial artery and advanced towards the stenosis. The lesion is then opened inflating a balloon and stents are deployed to maintain the artery open. Angioplasty started with Charles Dotter, who inadvertently advanced an angiography catheter through an iliac plaque in 1963 and with Andreas Grüntzig, who performed the first balloon angioplasty in 1974 and the first coronary catheterization in 1977 [6].



**Figure 1.1** Artist's impression of open heart surgery (left) and minimally invasive techniques (right) for cardiac interventions requiring small incisions and less trauma. The image has been reproduced from: http://www.policymed.com

Beside the treatment of atherosclerosis, endovascular approaches have also been developed for other applications, such as cardiac valve replacement, treatment of congenital defects in the heart, catheter ablation to treat atrial fibrillation and the creation of shunts between the hepatic vein and the portal vein in the liver to treat portal hypertension (Transjugular intrahepatic portosystemic shunt or TIPS).

The wide spread of minimally invasive interventions in different medical fields demonstrates the success of this approach. With shorter recovery time and reduced patient trauma, minimally invasive procedures are the preferred treatment choice in many surgeries. Other advantages of minimally invasive procedures over open surgery are reduced risk of infection, shorter hospital stay, fewer complications, and less blood loss [7]. However, some disadvantages are also present, such as the need of specialized medical equipment and specialized training for the surgeons. The success of a minimally invasive procedure is highly dependent on the skills and experience of the surgeon in manipulating and guiding the devices towards the desired target. Certain minimally invasive procedures can take longer than the corresponding traditional open surgery. Therefore, there is a constant need to further improve minimally invasive techniques and develop new technologies to achieve better outcomes and shorten the procedure time.

An appropriate visual feedback regarding the location of the instrument in the body and the working field during minimally invasive procedures is fundamental for a successful intervention. Moreover, with appropriate image guidance, information on the effects of a specific interventions can be gained. In laparoscopic procedures, imaging of the inside of the abdomen is achieved with a video camera that is inserted through small incisions [4]. A related technique, named angioscopy, has been developed to investigate the inside of the coronary arteries [8], but clinical use remains limited due to practical limitations and the unclear added value. Angioscopy requires clearance of blood from the lumen using saline, employs large catheters that cannot reach distal vessels and cannot pass narrow areas. Moreover, angioscopy provides only surface imaging [9].

An imaging modality that is clinically employed in minimally invasive vascular procedures is angiography, an X-ray imaging technique that generates 2D projection images of the arterial lumen. Angiography shows vessel diameter, allowing an assessment of stenosis, but does not provide images of the vessel wall. Angiography is used as image guidance in percutaneous interventions of the coronary and peripheral arteries and to provide guidance during TIPS procedures. Overall, since imaging of the arteries is achieved from outside, angiography provides limited information on the vessel, the local working field, and often has limitations in visualizing interventional devices such as stents or balloon catheters which have limited radioopacity.

A successful optics-based imaging technique that has been applied in intravascular applications is optical coherence tomography (OCT). OCT uses laser light to provides cross-sectional images of the arteries with a resolution of approx. 15  $\mu$ m [10]. The high resolution and the possibility of integration into intravascular catheters, make OCT a very attractive technique. A limitation of OCT is the penetration depth, which is limited to 1–2 mm and does not allow to visualize the total thickness of the artery wall.

Ultrasound-based imaging modalities achieve better penetration depth than OCT, still retaining acceptable resolution. Ultrasound imaging is employed in many minimally invasive procedures and continuous research and investigations are performed to improve minimally invasive ultrasound in terms of device miniaturization, improved field of view and increased resolution and penetration depth.

#### **1.2 ULTRASOUND FOR MINIMALLY INVASIVE PROCEDURES**

Conventional ultrasound imaging employs a piezoelectric (PZT) transducer array and is based on transmitting high frequency sound waves (above 20 kHz). The small PZT elements of the array are the active part of the transducer. An electrical signal causes vibration of the elements, which then generate pressure waves that propagate through the medium and are reflected back from acoustic heterogeneities in the tissue. The reflected pressure waves that are received by the elements are then converted back in electrical signals and processed to generate an image based on the time of flight and the signal strength for different tissues. Most of the clinically available ultrasound scanners employ a linear transducer array and generate B-mode images of the underlying tissue. The image is formed by electronic switching, sweeping the ultrasound beam over the array elements [11]. The higher the frequency and the larger the bandwidth used in ultrasound imaging, the higher the resolution. However, due to the increased attenuation with increased frequency, there is always a compromise between the resolution and the penetration depth required for each specific application.



**Figure 1.2** (a) TEE probe (Oldelft Ultrasound, Netherlands, 14 mm width), figure reproduced from http://www.oldelft.nl; (b) ICE catheter (Siemens AcuNav, 8-1-French), figure reproduced from http://www. kpiultrasound.com; (c) IVUS catheter (Philips Eagle Eye, 3 French), figure reproduced from https://www.usa. philips.com; (d) 2D TEE image (courtesy of Dr. C. Ren); (e) ICE image reproduced from earlier publication [12]; (f) IVUS image of a coronary artery (courtesy of E.M.J. Hartman). LA, left atrium; LV, left ventricle; AV, aortic valve; RA, right atrium.

Conventional non-invasive transthoracic echocardiography employs a transducer that is placed on the thorax and the ultrasound waves have to pass in between the ribs and the lungs before reaching the heart. Due to the quite long ultrasound travel distance, low frequencies (1–5 MHz) are usually the preferred choice. 2D Imaging of the heart from a closer distance can be achieved using transesophageal probes (TEE) that are placed in the esophagus, very close to the upper chambers of the heart. TEE has been successfully used during catheter ablation for atrial fibrillation [13] and treatment of atrial septal defects (ASD) [14].

Further developments led to the introduction of intracardiac echocardiography (ICE) [15,16], which is a catheter based technique that allows to advance the ultrasound transducer through the vasculature in the mid-right atrium. Conventional ICE catheters are one-dimensional arrays with a side-looking aperture, rendering imaging and interpretation of cardiac structures a task for experts. Moreover, the frequency used in ICE is approximately 7 MHz, close to the frequency used in TEE, hence limiting the imaging resolution [17].

A minimally invasive ultrasonic imaging technique that employs higher frequencies and is used to image the inside of the arteries, is intravascular ultrasound (IVUS). Coronary IVUS catheters (2.6–3.2 French) use frequencies in the range of 20–60 MHz and consist of either a single element mechanically rotated around the catheter axis or a circular array with the ultrasound beam electronically steered [15,18]. Both designs provide cross-sectional images of the arterial lumen, giving insights into atherosclerotic plaque morphology and aiding the treatment choice during coronary [19] and peripheral vasculature interventions. Since conventional IVUS provides side-looking imaging, it can only be used for lesions which are not highly stenotic and have a free lumen dimension compatible with IVUS catheters.

#### **1.3 MINIATURIZED FORWARD-LOOKING 3D ULTRASOUND**

Ultrasound imaging during minimally invasive procedures is mostly based on linear transducers that provide 2D images of 3D structures. Attempts have been made in the development of 2D matrix transducers that could be mounted on cardiac catheters to achieve volumetric imaging, but most of them are limited to side-looking approaches [20-22]. These design choices are primarily dictated by the limited dimensions of minimally invasive devices and the difficulties in integrating 2D matrix transducers at their tip.

To improve image guidance, miniaturized forward-looking (FL) transducers capable of generating 3D images of the tissue structures ahead of the device are required. This could be beneficial for applications such as monitoring during ASD corrections, valve replacement, TIPS, placement of stents, biopsy procedures and guidance during crossing of intravascular chronic total occlusions (CTO), which are atherosclerotic lesions that completely fill the vessel lumen and have a highly heterogeneous tissue composition, with variable mechanical properties [23].

First attempts in the development of forward-looking ultrasound imaging devices for minimally invasive procedures consisted of single-element transducers mounted at the tip of the catheter and mechanically scanned using wobbling and rotating mechanisms [24-28] making the whole assembly bulky and unstable. To avoid rotating mechanisms and achieve volumetric imaging, multi-element arrays have been investigated with each individual element directly connected to the external acquisition system [29-33]. Efforts towards reduction of coaxial cables have been taken by implementing integrated circuits with multiplexers to interface multiple PZT elements [34,35]. Moreover, a technology based on capacitive micromachined ultrasonic transducers (CMUT) has been explored, which allows fabrication of transducers with arbitrary geometry using standard micro-fabrication processes [36,37]. Employing this technology several forward-looking transducers providing 2D and 3D images have been investigated [38-44].

## **1.4 CHALLENGES IN MINIMALLY INVASIVE 3D FORWARD-LOOKING ULTRASOUND**

Since intensive effort is ongoing in the development of 3D forward-looking ultrasound imaging for minimally invasive procedures, it is important to understand the associated challenges. These challenges can be separated in those related to hardware and technical implementations and those related to image formation.

#### 1.4.1 Hardware challenges

The small dimensions of minimally invasive devices such as needles and catheters impose a major constraint on the effective aperture of the ultrasound transducer that can be mounted at the tip, limiting the total ultrasound dimensions to a few millimeters. Moreover, since minimally invasive devices can be placed very close to the imaging target, higher frequencies (>10 MHz) are preferred, which correspond to a very thin PZT layer (<150  $\mu$ m). Beside the thickness, the materials and the overall geometry of the transducer array will affect the acoustic behavior. Fabrication of transducers that fulfill all the geometrical requirements and have small dimensions is not trivial and requires careful acoustic stack design to optimize the transducer performance. The same observation holds for CMUT transducers, which have an operating frequency inversely related to the radius of the capacitor.

Next to the acoustic optimization and the transducer stack fabrication, electrical interconnections pose additional challenges for the successful integration of miniaturized ultrasound transducers in catheter and needles. To achieve volumetric imaging ahead of the minimally invasive device without using bulky mechanisms, a 2D matrix transducer array should be considered. However, for these multi-element transducers, an electrical connection is required between each element and the external imaging system. which poses the problem of accommodating multiple cables in the device [29-33]. Many minimally invasive instruments are designed with a hollow lumen to allow insertion of additional instruments such as guidewires, balloons and stents and sometimes they incorporate mechanical steering mechanisms to improve the device dexterity. This further reduces the available space that can be allocated for electrical cables. Integrated circuits have been developed in the effort of reducing the cable count [38-44], but careful circuit design and optimization is needed to interface all the elements.

An alternative approach could employ only a few elements, or even a single element, which sequentially scans the volume ahead of the catheter. Computational approaches are then needed to reconstruct a volumetric image from the acquired data. This method is similar to the circumferential scan in IVUS pullbacks, where serially acquired A-lines are spatially arranged to map out a volume. For forward-looking instruments, a transducer position tracking mechanism is required to make this work. The direct advantage of this concept is device simplicity, but image reconstruction is more complicated than with array sensors.

#### **1.4.2** Imaging challenges

The constraints on the aperture size of the transducers that can be mounted at the tip of minimally invasive devices greatly limits the imaging capability. Optimally, to provide sufficient information over the working field, forward-looking ultrasound transducers should provide volumetric imaging with sufficient spatial coverage over the region of interest and with high resolution in all the dimensions. It is known that lateral image resolution depends on the transducer aperture [45]. A way to overcome the aperture limitation is to consider synthetic aperture techniques, where multiple acquisitions at different spatial locations are combined in an effectively wider aperture. Simple hand manipulation of minimally invasive devices could provide a mean to move the transducer over multiple spatial location. This could be achieved by mounting the transducer at the tip of steerable devices. However, to combine the different acquisitions, real time 3D spatial coordinates of the device tip should be recorded with an appropriate motion tracking system. Video and magnetic motion tracking systems have been investigated for 3D freehand ultrasound imaging using conventional linear arrays [46,47]. These systems are rather bulky and not easily scalable to the dimensions needed for minimally invasive approaches. A tracking technique that is showing great potential in the intravascular field is optical shape sensing (OSS) [48,49], which employs a thin optical fiber and a laser system to reconstruct the 3D coordinates of the device in which the fiber is inserted (see grey text box). This technique has been combined with preoperative CT imaging to provide a roadmap during peripheral vasculature interventions.

#### **1.5 THESIS OUTLINE**

In the effort of overcoming the previously described challenges, this thesis presents solutions for miniaturized 3D forward-looking ultrasound imaging suitable for minimally invasive interventions. The goal was to develop compact and smart systems that could provide

volumetric ultrasound imaging with high resolution and that could be integrated at the tip of interventional catheters.

This thesis focuses on mainly two approaches to achieve minimally invasive 3D forward-looking ultrasound imaging. In the effort of addressing the hardware challenges, the first approach consists in developing a 2D matrix piezoelectric transducer array with an optimized acoustic design and an appropriate interconnection scheme requiring a limited number of electrical cables (**Chapter 2** and **3**). The second approach consists in developing less complex ultrasound devices having only a single element transducer that is integrated in the device together with steering mechanisms and position tracking systems that are exploited to provide volumetric imaging ahead of the tip (**Chapters 4**, **5**, **6** and **8**).

Transducer geometry optimization is the focus of **Chapter 2** with a detailed analysis on the influence of subdicing on the radiation impedance, on the time and frequency response and on the directivity of high frequency linear array elements.

A miniaturized 2D matrix transducer with a front-end ASIC with reduced cable count, suitable for FL-IVUS imaging is presented in **Chapter 3**. Transducer design, fabrication and acoustic characterization are described and FL volumetric imaging is demonstrated.

FL imagining approaches based on single element transducers are investigated starting from **Chapter 4**. Here a novel tip design for minimally invasive steerable devices that accommodates a 14 MHz single element transducer and an OSS fiber is described. A prototype of the integrated device has been manufactured and, by combining ultrasound A-lines and OSS position information acquired while steering the tip device, feasibility of 3D FL reconstruction has been demonstrated on a phantom. To achieve FL volumetric imaging, a method to process the sparse information obtained by manually steering the catheter tip with the transducer element needs to be implemented. This has been achieved in **Chapter 5**, where a clinically available steerable intracardiac catheter has been modified to accommodate a 25 MHz single element transducer and the OSS fiber. An adaptive normalized convolution method has been implemented and successful sparse image reconstruction of the surface of a tissue-mimicking phantom and of a human *ex-vivo* carotid plaque have been achieved. These successful results have been extended to 3D image reconstructions using an innovative intracardiac steerable catheter in **Chapter 6**.

To improve the lateral image resolution achieved using a single element transducer, the spatial information of the ultrasound beam is exploited. A well-known phenomenon resulting from spatially structured ultrasound beam with phase variance is speckle. A theoretical overview of speckle in ultrasound is presented in **Chapter 7**, whereas **Chapter 8** introduces Structured Ultrasound Microscopy (SUM), a novel imaging technique that achieves wide-field, depth-independent microscopic resolution using a single element transducer and a coding mask

that scrambles the phase uniformity and allows volumetric image reconstruction by solving an inverse scattering problem.

**Chapter 9** provides a discussion over the solutions presented for minimally invasive FL imaging together with a conclusion and future directions.

#### **Optical Shape Sensing (OSS)**

The optical shape sensing system that has been used in part of this thesis is a research prototype developed at Philips Research, Eindhoven, The Netherlands and it is based on the technology described in [48,49]. This system provides distributed sensing over a fiber and it could be beneficial for medical devices that are highly flexible, like catheters and guidewires.

The system consists of two major components: a multi-core optical fiber with an outer diameter of 200  $\mu$ m and a scanning laser system. The fiber has one core running along the longitudinal axis and three cores located axially 120° apart from each other and arranged in a helical fashion around the central one. Fiber Bragg Gratings (FBG) are patterned into the cores along the entire length of the fiber (1.8 m) with a spacing of 50  $\mu$ m. FBGs consist in periodic changes in refractive index that reflect a band of light with a wavelength that is related to the periodicity of the index variation. When the fiber is bent, stretching and compression induce local changes in the period of the FBG, hence varying the wavelength of the reflected light. It is through this mechanism that distributed strain measurements are obtained.

The Philips scanning laser system has a central wavelength set to 1545 nm and is swept over 17 nm. The laser system is used to interrogate the sensors along the fiber using an optical frequency domain reflectometry technique. The specific location of the FBG will result in a specific frequency modulation of the carrier signal received, hence enabling distributed measurements of the strain applied to each of the cores. By integrating along the fiber starting from the first sensing point (that is fixed in the reference frame) and considering that each sensing point consists of a triplet of values, the strain measurements can be processed to obtain the 3D shape of the fiber. Since the position of each sensing point is computed based on the previous one, the accuracy of the reconstruction is inversely proportional to the length considered.



**Figure 1.3** The picture shows an x-ray aortogram during a pre-clinical experiment, taken before navigating with shape sensed devices. In the aortagram contrast fluid is injected into the vessels to make the location of the vessels visible on x-ray. The contrast injection catheter is also clearly visible. This 2D roadmap of the vessels is registered with Philips shape sensed devices, and can be used to navigate the devices to a specific location without further use of x-ray. Yellow shows a guidewire, and blue shows a catheter. The devices are used for navigation inside the aorta, in this case to cannulate the left renal artery. Picture courtesy of Philips Research.

# Chapter 2

### Improving the Performance of a 1D Ultrasound Transducer Array by Subdicing

Based on:

**Jovana Janjic**, Maysam Shabanimotlagh, Martin D. Verweij, Gijs van Soest, Antonius F.W. van der Steen, Nico de Jong, Quantifying the effect of subdicing on element vibration in ultrasound transducers, *Proc. IEEE International Ultrasonics Symposium (IUS), Oct. 2015, pp 1-4.* 

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#### ABSTRACT

In medical ultrasound transducer design, the geometry of the individual elements is crucial since it affects the vibration mode of each element and its radiation impedance. For a fixed frequency, optimal vibration (i.e. uniform surface motion) can be achieved by designing elements with very small width-to-thickness ratios. However, for optimal radiation impedance (i.e. highest radiated power), the width should be as large as possible. This leads to a contradiction that can be solved by subdicing wide elements. To systematically examine the effect of subdicing on the performance of a 1D ultrasound transducer array, we applied finite-element simulations. We investigated the influence of subdicing on the radiation impedance, on the time and frequency response, and on the directivity of linear arrays with variable element widths. We also studied the effect of varying the depth of the subdicing cut. The results show that, for elements having a width greater than 0.6 times the wavelength, subdicing improves the performance compared to that of non-subdiced elements: the emitted pressure may be increased up to a factor of three, the ringing time may be reduced by up to 50%, the bandwidth increased by up to 77%, and the side lobes reduced by up to 13 dB. Moreover, this simulation study shows that all these improvements can already be achieved by subdicing the elements to a depth of 70% of the total element thickness. Thus subdicing can improve important transducer parameters and, therefore, help in achieving images with improved signal-to-noise ratio and improved resolution.

#### **2.1 INTRODUCTION**

Most of the clinically available ultrasound imaging probes are one-dimensional (1D) linear transducers. They consist of small piezoelectric (PZT) elements, which are the active components of the probe. An electrical excitation is used to induce vibration of the individual elements, which in turn generate pressure waves that propagate into the medium. The pressure waves reflected from different scatterers are subsequently received by the elements and converted back into electrical signals, which are processed in the ultrasound machines to yield echo images. To create images with high resolution and high signal-to-noise ratio, the geometry of the elements plays a crucial role in transducer design. This is because the geometry affects both the element vibration and the element radiation impedance.

Regarding the element vibration, it is known that if the width-to-thickness ratio is smaller than 0.7 [50], the element will mainly vibrate along the thickness direction. Knowing that the wavelength in PZT is almost twice the wavelength in water ( $\lambda$ ), we can also say that thickness vibration is obtained when the width of the element is well below 0.7 $\lambda$ . Elements with a thickness vibration have a piston-like behavior; this behavior is considered optimal because all the surface points of the element, when excited, will have the same velocity amplitude and phase, which is favorable for the transmission efficiency.

The geometry also affects the radiation impedance, which describes the acoustic coupling of the vibrating element to the medium, and is usually defined for piston motion [51]. To obtain the highest radiated power, the radiation impedance should be real and equal to the acoustic impedance of the medium. For a wavelength  $\lambda$  and a circular element with diameter a, the complex radiation impedance is mainly imaginary for  $a < 0.5\lambda$  [51,52]. This means that little or no energy is transmitted into the medium. Similar behavior is found for square elements with sides a and infinitely long strip elements with width a: the radiation resistance drops quickly when a assumes values below  $0.7\lambda$  [53,54]. The strip geometry is representative of 1D ultrasound transducer arrays, where each element is many times longer than it is wide.

From what was stated in the previous paragraphs, it is clear that to design 1D transducer arrays with piston-like behavior, elements with very small widths are preferred. However, for optimal radiation impedance, the element should be as large as possible. This leads to a contradiction when trying to achieve both optimal vibration and optimal radiation impedance. The current approach is to consider elements having a width between  $0.5\lambda$  and  $\lambda$ . This, however, might not be the optimal solution.

The geometrical requirements for optimal vibration and optimal radiation impedance are even more crucial for high frequency transducers. To better explain this, we can consider a linear array with center frequency of 15 MHz and wavelength  $\lambda$ =100 µm in water. If we assume a PZT speed of sound of approximately 3000 m/s, the element thickness has to be 100 µm. Therefore, for optimal vibration and optimal radiation efficiency, the width of the element has to be about 70  $\mu$ m. Such small dimensions, however, are difficult to achieve when manufacturing transducers, especially if we consider the difficulty of making electrical connections to these individual elements. A possible solution to this problem is to consider elements with a width greater than 0.7 $\lambda$  and perform subdicing.

Subdicing means cutting each transducer element in two or more sub-pillars while keeping the electrical connection between the elements intact [55]. Subdicing decreases the width-to-thickness ratio of the vibrating pillars and separates the lateral vibration from the thickness resonance [50,55-58]. The depth of the subdicing cut is also important, mainly because of fabrication issues; by partially subdicing the elements, damage of the electrodes can be prevented and mechanical stability can be ensured. To date, it seems that no systematic studies have been performed that quantitatively show the effect of subdicing in itself and the influence of the subdicing depth on the transducer performances. In our opinion, such systematic studies are important to further improve the efficiency of the transducer design process.

In this paper, we systematically present the effect of subdicing by showing Finite Element Analysis (FEA) simulations obtained with PZFlex software (Weidlinger Associates, Los Altos, CA). We focus on the performance of linear array elements with an operational frequency of approximately 14 MHz. We analyze the influence of subdicing on the radiation impedance, on the time and frequency response and on the directivity pattern. We also investigate the effect of the depth of the cut to understand if it is possible to partially subdice the elements and still achieve good transmission performance.

#### 2.2 DEFINITION OF SIMULATION MODEL AND ROLE OF SUBDICING

In 1D arrays the length of the elements is much larger than the width and the thickness [59]. Therefore, in this paper the elements will be considered infinitely long, which reduces our simulations to two-dimensional (2D) problems.

The 2D simulation study is performed using the FEA software PZFlex, and the considered model is depicted in Figure 2.1. In this study, the thicknesses of all the layers are fixed, while the width of each of the elements is varied from 10  $\mu$ m to 250  $\mu$ m in steps of 10  $\mu$ m. Only a portion of a realistic linear array is modeled, with a central active element and three neighboring passive elements on both sides. The space between the elements and in the subdicing cuts is approximately 8  $\mu$ m wide and is taken to be void [60], while water is used as the medium in which the ultrasound is propagating. After subdicing, the width of the elements and the subdicing cuts may differ from the ideal values by at most ± 2  $\mu$ m, i.e. one grid step, due to numerical rounding. To reduce the simulation time, we assume symmetry in the plane *x*=0. With this condition, only half of the model needs to be simulated. The domain is surrounded

by absorbing boundary conditions to avoid reflections from its edges. The thicknesses and material properties of each layer of the transducer are given in Table 2.1.



**Figure 2.1** The 2D simulation model. Seven transducer elements are modeled and only the central element is electrically excited. The width of each of the elements is *a*, the width of the sub-pillars is *w* and the depth of the cuts is *d*.

| Table 2.1 Thicknesses and Material Properties Used in the Simulations |  |
|---|--|
|   |  |

| Layer    | Material                    | Thickness<br>[µm] | Material<br>Properties  |
|----------|-----------------------------|-------------------|---|
| PZT      | Ceramic<br>(PZT-5A)         | 100               | $\begin{array}{c} \rho {=}7700 \text{ kg/m}^3 \\ c_{11}{=}147, c_{12}{=}105, c_{13}{=}93.7, \\ c_{33}{=}113, c_{44}{=}23, \\ c_{66}{=}21.2 \text{ MPa} \\ \epsilon_{11}{=}10, \epsilon_{33}{=}8.09 \times 10^{-9} \text{ C/Vm} \\ e_{15}{=}11.64, e_{31}{=}{-}3.09, \\ e_{33}{=}16 \text{ C/m}^2 \end{array}$ |
| Matching | Conductive<br>Epoxy         | 30                | ρ=3350 kg/m³,<br>Z=6.7 MRayl<br>c <sub>shear</sub> =1050 m/s  |
| Backing  | Tungsten<br>loaded<br>epoxy | 500               | ρ=2975 kg/m³,<br>Z=5.8 MRayl<br>c <sub>shear</sub> =1960 m/s  |

To validate the PZFlex model, we have compared the simulation results for the radiation impedance and for the directivity to the analytical curves. Moreover, a numerical test has been performed with a twice denser numerical grid. The obtained results showed no significant differences with the results obtained for our standard grid. This ensures the numerical accuracy of our simulations.

The electrical impedance plot is the primary tool that we will employ to see whether an element has vibration modes with a non-uniform surface motion. To explain this, we show in Figure 2.2 and Figure 2.3 the electrical impedance versus frequency for elements in different situations. Figure 2.2 demonstrates how the electrical impedance plot for a narrow (a= 40 µm =0.4 $\lambda$ ),

non-subdiced element with piston-only motion is formed. The magnitude of the electrical impedance for a bare PZT element with a piston-like behavior is shown in Figure 2.2(a). In the considered frequency band, a single minimum value of the electrical impedance occurs in this case at a frequency of 12.6 MHz. For this so called resonance frequency, the corresponding mode shape is shown, confirming that the element vibrates only along the thickness direction. When a matching layer is added on top of the PZT, the electrical impedance changes, as shown in Figure 2.2(b). It now has two minima and two resonance frequencies due to the presence of two different layers. For both resonance frequencies the indicated mode shapes are still showing the typical piston-like motion.

Figure 2.3 illustrates what happens to the electrical impedance for a wide element before and after subdicing. Figure 2.3(a) shows the electrical impedance of a wide element ( $a=120 \mu m=1.2\lambda$ ) before subdicing. Multiple local minima are present, which correspond to different resonance modes. Two of the three resonance mode shapes are clearly showing an undesired non-uniform surface motion. The occurrence of the typical local minima in Figure 2.3(a) is an indication that subdicing will be opportune. Figure 2.3(b) shows the electrical impedance when the same element is provided with two subdicing cuts that extend to the bottom of the PZT layer. Doing so yields an electrical impedance plot with two minima and corresponding mode shapes with a piston-like behavior. For the wide element considered, subdicing has restored the desired situation of Figure 2.2(b).



**Figure 2.2** Plot of the electrical impedance per unit length for an element with a small width (a=40 µm = 0.4 $\lambda$ ) consisting of bare PZT (a) and PZT with a matching layer (b). The gray insets show the mode shapes for the resonance frequencies at each minimum (light gray=PZT; dark gray=matching layer).



**Figure 2.3** Plot of the electrical impedance per unit length for an element with a large width  $(a=120 \ \mu m =1.2\lambda)$  before subding (a) and after subdicing with 2 cuts that run through the entire PZT and matching layer (b). The gray insets show the mode shapes for the resonance frequencies at each minimum (light gray=PZT; dark gray=matching layer).

#### 2.3 OPTIMAL NUMBER OF SUBDICING CUTS

We determined the optimal number of subdicing cuts for elements with widths *a* between 10 µm and 250 µm. We have defined the optimal number of cuts as the lowest *n* that yields  $a/(n + 1) \le \lambda/2 = 50$  µm. We have numerically verified that this is the number of equidistant subdicing cuts that will exclude modes with non-uniform surface motion, i.e. for obtaining the situation in Figure 2.3(b). For the simulations, the central element was excited with a chirp pulse having a duration of 1 µs and a frequency sweep between 5 and 25 MHz. The chirp excitation was designed to have an even frequency content across the range of interest. The center frequency  $f_c$  of the subdiced element was obtained as the center of the -3 dB bandwidth of the average surface pressure.

In Table 2.2 the optimal number of subdicing cuts *n* and the corresponding center frequency  $f_c$  are shown for elements with different widths *a*. The table also shows the individual sub-pillar width *w* after subdicing as well as the products and  $k_c a$  and  $k_c w$ , where:

$$k_c = \frac{2\pi f_c}{c} = \frac{2\pi}{\lambda_c} , \qquad 2.1$$

with *c* the speed of sound and  $\lambda_c$  the wavelength at  $f_{c'}$  both in water. The number of optimal subdicing cuts *n* varies from 0 to 4, while the sub-pillar width is ranging from 26 to almost 47 µm. After subdicing, the product  $k_c$  is always close to 2 and the frequency  $f_c$  is about 14 MHz for all the widths *a*. This confirms that after subdicing the dominant resonance frequency depends

mainly on the thickness of the element, which is kept constant. In all the remaining simulations, the resonant frequency  $f_c$  is used as the excitation frequency for each specific width a.

**Table 2.2** The Optimal Number of Subdicing cuts , the Center Frequency  $f_{c'}$  the Sub-pillar Width w, and the Products  $k_c a$  and  $k_c w$ , for Each Element Width a.

| а   | n | $f_c$<br>[MHz] | <i>w</i><br>[µm] | k <sub>c</sub> a | k <sub>c</sub> w |  |
|-----|---|----------------|------------------|------------------|------------------|--|
| 10  | 0 | 17.9           | 10               | 0.75             | 0.75             |  |
| 20  | 0 | 15             | 20               | 1.26             | 1.26             |  |
| 30  | 0 | 14.5           | 30               | 1.83             | 1.83             |  |
| 40  | 0 | 13.5           | 40               | 2.27             | 2.27             |  |
| 50  | 0 | 12.8           | 50               | 2.69             | 2.69             |  |
| 60  | 1 | 14.6           | 26               | 3.68             | 1.59             |  |
| 70  | 1 | 14.3           | 32               | 4.20             | 1.92             |  |
| 80  | 1 | 14.2           | 36               | 4.77             | 2.15             |  |
| 90  | 1 | 13.4           | 42               | 5.07             | 2.36             |  |
| 100 | 1 | 13.1           | 46               | 5.50             | 2.53             |  |
| 110 | 2 | 14.7           | 33.3             | 6.79             | 2.06             |  |
| 120 | 2 | 14.5           | 36               | 7.31             | 2.19             |  |
| 130 | 2 | 14.3           | 40 7.81 2.4      |                  | 2.40             |  |

| а   | п | $f_c$<br>[MHz]  | <i>w</i><br>[μm] | k <sub>c</sub> a | k <sub>c</sub> w |
|-----|---|-----------------|------------------|------------------|------------------|
| 140 | 2 | 13.9            | 42.7             | 8.17             | 2.49             |
| 150 | 2 | 13.5            | 46.7             | 8.50             | 2.65             |
| 160 | 3 | 14.4            | 35               | 9.68             | 2.12             |
| 170 | 3 | 14.3            | 38               | 10.21            | 2.28             |
| 180 | 3 | 14.1            | 40               | 10.66            | 2.37             |
| 190 | 3 | 13.6            | 38               | 10.85            | 2.17             |
| 200 | 3 | 13.5            | 45               | 11.34            | 2.55             |
| 210 | 4 | 14.3            | 37.6             | 12.61            | 2.26             |
| 220 | 4 | 14.2            | 39.2             | 13.12            | 2.34             |
| 230 | 4 | 14.1            | 41.6             | 13.62            | 2.46             |
| 240 | 4 | 13.8            | 43.2             | 13.91            | 2.50             |
| 250 | 4 | 13.6 45.6 14.28 |                  | 14.28            | 2.60             |

#### **2.4 RADIATION IMPEDANCE**

For each width a and corresponding frequency  $f_{c'}$  two different types of excitations have been simulated: a 10-cycle sinusoid is used to simulate a quasi-continuous excitation, and a 2-cycle sinusoid is used to simulate an impulse like excitation, which is commonly employed in imaging systems. After driving the element with either type of excitation, the average pressure and the average velocity over its surface have been computed. In Figure 2.4(a) and Figure 2.4(b) the maximum of the envelope of the average pressure  $P_{max}$  and the maximum of the envelope of the average pressure  $P_{max}$  and the maximum of the envelope of non-subdiced and the optimally subdiced case.

The obtained values for  $P_{max}$  and  $v_{max}$  are also used to obtain the modulus of the radiation impedance Z. The latter quantity is defined for a piston moving in a perfectly rigid baffle as the average complex pressure amplitude divided by the real normal particle velocity amplitude [61].Therefore, |Z| may be found from Figure 2.4(a) and Figure 2.4(b) as:

$$|Z| = \frac{P_{max}}{v_{max}} .$$
 2.2

Figure 2.4(c) shows the relative radiation impedance  $|Z_r|$ , defined as:

$$|Z_r| = \frac{|Z|}{Z_{water}} , \qquad 2.3$$

where  $Z_{water} \cong 1.5$  MRayl. Moreover, in Figure 2.4(c) the analytical curve for the radiation impedance of an infinitely long 2D strip with piston-like behavior in a rigid baffle is shown for comparison. This curve is the result of the numerical computation of the following equation (see Appendix for detailed derivation):

$$Z_r(\omega) = \frac{1}{K} \int_{-K/2}^{K/2} \int_{-K/2}^{y} H_0^{(2)}(y - y') dy' dy , \qquad 2.4$$

where  $H_0^{(2)}$  is the Hankel function of the second kind and zero order, and  $K=k_ca$ .

For  $k_c a > 3.5$  (or  $a > 0.55 \lambda_c$ ), the pressure and velocity after subdicing are up to 3 times higher than those in the non-subdiced case for both types of excitations. This improvement in the transmit efficiency can be explained by two effects. The first effect of subdicing is the removal of modes with non-uniform surface motion, which causes an inefficient radiation into the acoustic medium. The second effect is a change in the electromechanical coupling factor of the PZT: when subdicing wide elements, the vibrational behavior of the PZT will change from plate mode to bar mode, which has a higher coupling factor [59,62,63].

In Figure 2.4(c), for  $k_c a < 2$  (or  $a < 0.32 \lambda_c$ ), the simulation results are in good agreement with the theoretical curve, confirming that for very small elements the radiation impedance drops quickly, impairing the energy radiation into the medium.

Without subdicing, the radiation impedance for the 10-cycle excitation shows an irregularity of 40% around  $k_c a=10$ , unlike the 2-cycle excitation. This difference in behavior may be explained by the bandwidth of the two excitation signals. For a quasi-continuous excitation with a narrow bandwidth, most of the energy is used to excite a specific frequency, which might correspond to a mode with non-uniform surface motion. The 2-cycle pulse simultaneously excites all occurring resonance frequencies, including the dominant thickness resonance. Therefore, simulations with narrow pulse excitation result in a radiation impedance that is in better agreement with the analytical derivation.

Disregarding the drop at  $k_c a=10$  that is seen for the 10-cycle excitation in the absence of subdicing, the radiation impedances obtained from the simulations assume in general values higher than those expected from the analytical derivation. This may be caused by the fact that the analytical curve is derived for a piston in a rigid baffle, while the simulated model is more representative of a transducer in a baffle with a finite compliance. When a compliant baffle boundary condition is assumed, the pressure at the surface of the element is higher than that

for rigid baffle boundary condition [64]. Therefore, with a compliant baffle, a higher radiation impedance is expected.

Overall, for  $k_c a > 5$  (or  $a < 0.48 \lambda_c$ ) and an optimal number of subdicing cuts, a dominant thickness vibration is obtained for both types of excitation, which results in a relative radiation impedance close to 1. This means that the radiation impedance is almost equal to the acoustic impedance of the medium and, therefore, almost all the power is radiated into the far field.



**Figure 2.4** Maximum average pressure (a), maximum average velocity (b), and the relative radiation impedance (c) for both types of excitations, plotted against  $k_c a$ . The width a of the elements is varying from 10 µm to 250 µm. For the subdiced curves the optimal number of subdicing cuts n is indicated.

#### **2.5 TIME AND FREQUENCY RESPONSE**

In this section we consider the effect of subdicing on the time and frequency responses of the elements in transmit and receive. We consider four different element widths that require a different optimal number of subdicing cuts:  $a=60 \ \mu m (k_c a \cong 3.7)$ ,  $a=120 \ \mu m (k_c a \cong 7.3)$ ,  $a=180 \ \mu m (k_c a \cong 10.7)$ , and  $a=240 \ \mu m (k_c a \cong 13.9)$ . For the transmit response, a 2-cycle sinusoidal voltage excitation with 1 V amplitude is defined at the electrodes of the element, and the resulting emitted average pressure at the surface of the element is computed. For the receive response, a uniform 2-cycle sinusoidal pressure excitation with 1 Pa amplitude is defined close to the surface of the element, and the voltage at the electrodes of the element is computed.

For both the subdiced and the non-subdiced elements, the time trace of the transmitted pressures and the envelope of these signals are shown in Figure 2.5. For all but the smallest width, a relevant enhancement in the peak pressure amplitude is seen after subdicing, which is consistent with the results shown in Figure 2.4(a). For the non-subdiced case, Figures 2.5(b), (c) and (d) show multiple peaks in the envelopes and long ringing tails. Subdicing substantially reduces these adverse effects in the transmitted waveform.



**Figure 2.5** Average transmit pressure versus time for elements of size  $a=60 \ \mu m \ (k_c a \cong 3.7)$  (a),  $a=120 \ \mu m \ (k_c a \cong 7.3)$  (b),  $a=180 \ \mu m \ (k_c a \cong 10.7)$  (c), and  $a=240 \ \mu m \ (k_c a \cong 13.9)$  (d). Plots for both the non-subdiced (n=0) and the optimally subdiced  $(n\neq 0)$  case are given, with envelopes. Excitation occurs with a 2-cycle sinusoidal pulse with 1 V amplitude.

Figure 2.6 depicts the frequency responses of the subdiced and the non-subdiced elements, both for transmit and receive. The differences between subdiced and non-subdiced elements are also apparent in the frequency domain. For non-subdiced elements, Figure 2.6 shows strong spectral amplitude fluctuations in both transmit and receive spectra, which disappear after subdicing. In general, the frequency responses in transmit and receive have a similar shape after subdicing, although they are not exactly the same. The small differences are explained by the reciprocity equation [65], which relates the transmit function to the receive transfer function via the electrical impedance of the element. Another interesting observation from both the time responses and the frequency spectra is the similarity between the curves for all the different element widths after subdicing. This is because the final sub-pillar width is almost constant (see Table 2.2). This means that the time and frequency responses after subdicing have become dependent on the sub-pillar width rather than on the total element width. To further quantify the improvement of the time response of the transmitted pressure, we define the -6 dB and -20 dB ringing intervals  $\Delta t_{-6 dB}$  and  $\Delta t_{-20 dB}$  which are the time intervals over which

the amplitude of the envelope is greater than -6 dB and -20 dB of its corresponding maximum. In addition to that, we introduce the relative ringing amplitude, defined as the size of the second peak of the pressure envelope divided by the size of the first (main) peak. The frequency domain performance in transmit is further quantified by the relative -3 dB bandwidth ( $BW_{-3 dB}$ ), and the maximum dip within the band (in dB relative to the -3 dB level).



**Figure 2.6** Frequency responses in transmit and receive for elements of size  $a=60 \ \mu m \ (k_c a \cong 3.7)$  (a),  $a=120 \ \mu m \ (k_c a \cong 7.3)$  (b),  $a=180 \ \mu m \ (k_c a \cong 10.7)$  (c), and  $a=240 \ \mu m \ (k_c a \cong 13.9)$  (d). Plots for both the non-subdiced (n=0) and the optimally subdiced ( $n\neq 0$ ) case are given.

Table 2.3 lists these quantities for an element of size  $a=120 \ \mu m \ (k_c a \cong 7.3)$  and a number of subdicing cuts ranging from 0 to 4. For this particular element, two subdicing cuts remove the spurious modes from the electrical impedance (Figure 2.3). However, it is interesting to investigate if n=2 is also optimal in terms of time and frequency response. Table 2.3 shows that when is increased from 0 to 2 the ringing times are shortened and the relative ringing amplitude is decreased. For n=2, a -3 dB bandwidth of 43.2% without dips is obtained. However,

no further substantial improvement is obtained for n>2. Therefore, for optimal performance in both the time and the frequency domain, two subdicing cuts are sufficient.

The considerations for  $k_c a \cong 7.3$  can be extended to the other element widths. In Table 2.4 the characteristic time and frequency domain quantities are shown for the non-subdiced and the optimally subdiced cases. The results demonstrate that the ringing time may be shortened by up to 50% ( $\Delta t_{-20 \ dB}$  for  $k_c a \cong 13.9$ ) and the relative ringing amplitude may be decreased by up to 80% ( $k_c a \cong 7.3$ ). Moreover, the maximum increase in bandwidth is 77% ( $k_c a \cong 3.7$ ), while the maximum dip is reduced by up to around 24 dB ( $k_c a \cong 7.3$ ).

**Table 2.3** Characteristic Time Domain and Frequency Domain quantities for an element of size  $a=120 \ \mu m$  ( $k_c a \cong 7.3$ ) and the number of subdicing cuts n ranging from 0 to 4.

|                                  | n | ∆t <sub>-6 dB</sub><br>[ms] | Δt <sub>-20 dB</sub><br>[ms] | Ringing<br>Amplitude<br>(%) | BW <sub>-3 dB</sub><br>(%) | Max.<br>dip<br>[dB] |
|----------------------------------|---|-----------------------------|------------------------------|-----------------------------|----------------------------|---------------------|
| a=120 μm<br>k <sub>c</sub> a=7.3 | 0 | 0.25                        | 0.42                         | 69.0                        | 60.2                       | 23.7                |
|                                  | 1 | 0.17                        | 0.58                         | 13.2                        | 29.8                       | 8.4                 |
|                                  | 2 | 0.14                        | 0.32                         | 13.5                        | 43.2                       | 0                   |
|                                  | 3 | 0.14                        | 0.33                         | 15.4                        | 44                         | 0                   |
|                                  | 4 | 0.14                        | 0.33                         | 15.6                        | 45                         | 0                   |

**Table 2.4** Characteristic time domain and frequency domain quantities for elements of size  $a=60 \ \mu m$  ( $k_c a \cong 3.7$ ),  $a=120 \ \mu m$  ( $k_c a \cong 7.3$ ),  $a=180 \ \mu m$  ( $k_c a \cong 10.7$ ), and  $a=240 \ \mu m$  ( $k_c a \cong 13.9$ ). Data for both the non-subdiced (n=0) and the optimally subdiced ( $n\neq 0$ ) case are given.

|                                      | n | $\Delta t_{-6 dB}$<br>[ms] | ∆ <i>t<sub>-20 dB</sub></i><br>[ms] | Ringing<br>Amplitude<br>(%) | <i>BW</i> <sub>-3 <i>dB</i></sub> (%) | Max.<br>dip<br>[dB] |
|--------------------------------------|---|----------------------------|-------------------------------------|-----------------------------|---------------------------------------|---------------------|
| <i>a</i> =60 µm                      | 0 | 0.18                       | 0.64                                | 23                          | 25.5                                  | 0                   |
| <i>k<sub>c</sub>a</i> =3.7           | 1 | 0.14                       | 0.34                                | 15.4                        | 45.1                                  | 0                   |
| a=120 μm                             | 0 | 0.25                       | 0.42                                | 69                          | 60.2                                  | 23.7                |
| $k_{c}a=7.3$                         | 2 | 0.14                       | 0.32                                | 13.5                        | 43.2                                  | 0                   |
| a=180 μm                             | 0 | 0.14                       | 0.32                                | 41.8                        | 37.4                                  | 1.2                 |
| <i>k<sub>c</sub>a</i> =10.7          | 3 | 0.14                       | 0.29                                | 13.3                        | 45.2                                  | 0                   |
| a=240 μm                             | 0 | 0.15                       | 0.65                                | 19.1                        | 27.2                                  | 1.3                 |
| <i>k</i> <sub>c</sub> <i>a</i> =13.9 | 4 | 0.14                       | 0.32                                | 12.7                        | 45.1                                  | 0                   |

#### 2.6 DIRECTIVITY PATTERN

Directivity can be calculated in PZFlex using the extrapolation toolkit, which enables the computation of the magnitude of the transmitted pressure in the far field, assuming linear propagation. The directivity patterns for both the non-subdiced and the optimally subdiced cases are shown in Figure 2.7. Additionally, in the same figure we show the analytical curve of the directivity  $D(\Theta)$ , which follows from:

$$D(\theta) = \operatorname{sinc}\left(\frac{a\sin(\theta)}{\lambda}\right),$$
 2.5

with  $\Theta$  the observation angle, *a* the transducer width and  $\lambda$  the wavelength [59]. This expression holds for an ideal piston in a rigid baffle.



**Figure 2.7** Directivity plots for elements of size  $a=60 \ \mu m \ (k_c a \cong 3.7)$ , (a)  $a=120 \ \mu m \ (k_c a \cong 7.3)$ , (b),  $a=180 \ \mu m \ (k_c a \cong 10.7)$  (c), and  $a=240 \ \mu m \ (k_c a \cong 13.9)$ . (d) Plots for both the non-subdiced (n=0) and the optimally subdiced ( $n\neq 0$ ) case are given, together with the analytical curve.
For the smallest element size  $a=60 \ \mu m \ (k_c a \cong 3.7)$  no side lobes are present and no substantial difference is seen between the non-subdiced and the subdiced cases. The other element sizes give rise to side lobes, which are reduced by subdicing. Moreover, the directivity patterns for the subdiced elements are in good agreement with the analytical curves. This agreement further demonstrates the benefits of subdicing in achieving the ideal piston-like behavior.

It is important to mention that the directivity and, hence the choice of the element width, will depend on the specific application [64].

# 2.7 SUBDICING DEPTH

Because the cutting depth is an important aspect in the construction of a transducer, in this section we consider the influence of using subdicing cuts that do not extend to the bottom of the PZT layer. The depth of the subdicing cut is varied from 0 to 100% of the total element thickness, i.e. the thickness of the PZT and the matching layer, in steps of 10%. Figure 2.8 shows the maximum of the average surface pressure, relative ringing amplitude, bandwidth, and maximum dip within the frequency band, as function of the depth of the subdicing cut.

For  $a=60 \ \mu m \ (k_c a \cong 3.7)$ , both the maximum average surface pressure (Figure 2.8(a)) and the relative ringing amplitude (Figure 2.8(b)) are almost constant for increasing depth of cut because the element is so small that piston motion is predominant even in the non-subdiced case. This is in agreement with the results of the previous sections. For wider elements the increase in pressure and the decrease in relative ringing amplitude are appreciable when subdicing up to 70% of the total element thickness. The -3 dB bandwidth (Figure 2.8(c)) and the maximum dip within the band (Figure 2.8(d)) as function of the depth of the cut have a less clear trend for the different element widths. Nevertheless we can state that to have both a large bandwidth and avoid strong fluctuations within the band, the depth of the cut should be between 70 and 100%, irrespective of the width of the element. In this range also the emitted pressure and the relative ringing amplitude are about constant and assume the highest and lowest value, respectively.

Figure 2.9 shows the relative side lobe levels for single elements of size  $a=180 \ \mu m \ (k_c a \cong 10.7)$ and  $a=240 \ \mu m \ (k_c a \cong 13.9)$  as function of the depth of the cut. For increased subdicing depth, the side lobes are reduced and, similar to the other characteristic quantities, for a depth of cut between 70 and 100%, an almost constant minimum value is reached.



**Figure 2.8** Influence of the relative depth of the subdicing cut *d*. Maximum average surface pressure (a), relative ringing amplitude (b), relative –3 dB bandwidth in % (c), maximum dip in the frequency band (d). Data are shown for elements of size *a*=60 µm ( $k_c a \cong 3.7$ ), *a*=120 µm ( $k_c a \cong 7.3$ ), *a*=180 µm ( $k_c a \cong 10.7$ ), and *a*=240 µm ( $k_c a \cong 13.9$ ). In each case, the optimal number of subdicing cuts is applied. In the shaded area, almost no further improvements are obtained.



**Figure 2.9** Relative side lobe level versus the relative depth of the subdicing cut. Data are shown for elements of size  $a=180 \text{ }\mu\text{m}$  ( $k_c a \cong 10.7$ ) and  $a=240 \text{ }\mu\text{m}$  ( $k_c a \cong 13.9$ ). In the shaded area, almost no further improvements are obtained. The dashed line is the analytical side lobe level.

# 2.8 CONCLUSION

In this study we have shown, through FEM simulations, that subdicing improves the performance of transducer elements having a width greater than 0.6 times the resonance wavelength in water. More precisely, subdicing improves the emitted pressure, the radiation impedance, the time response, the frequency response, and the directivity pattern.

The increase in the emitted pressure and in the radiation impedance after subdicing leads to an increase in the power radiated into the medium. High power is beneficial for improved signal-to-noise ratio during imaging. Regarding the time response, we showed that subdicing reduces the ringing time and the ringing amplitude. Ringing is an unwanted effect, which occurs when the radiation impedance of an element including a matching layer deviates from the impedance of the medium. This is the case when there is a non-uniform motion of the element surface. Subdicing restores the uniform motion at the surface and thus ringing will be avoided. A time response without ringing has a shorter pulse length, which in turn improves the axial imaging resolution. The frequency response also benefits from subdicing by obtaining a higher bandwidth and avoiding strong spectral dips.

Regarding the directivity pattern, low levels of side lobes are necessary to avoid imaging artifacts originating from off-axis ultrasound beams. We have shown that subdicing can also reduce the side lobe levels of wide elements. The depth of the subdicing cut may be crucial during the fabrication process. Cutting all the way through the element can lead to mechanical instability and cause damage to the electrodes. We have demonstrated by our simulations that it is possible to achieve the most significant improvements in transducer performances by cutting to only a depth of 70% of the total element thickness.

We expect that the presented quantitative knowledge about the effect of subdicing on the transducer performances may help to improve the design of high quality imaging transducers.

### **APPENDIX: RADIATION IMPEDANCE FOR A 2D STRIP**

In order to derive the radiation impedance for a 2D strip of width *a* and uniform velocity  $v_0 \exp(j\omega t)$  over the surface, we assume a rigid baffle boundary condition and we consider the two-dimensional frequency-domain Green's function [66]:

$$G(x, z, \omega) = -\frac{j}{4} H_0^{(2)}(kr) \text{ with } r = \sqrt{x^2 + z^2}$$
 2.6

where x is the coordinate in the direction of the width of the strip, z is the coordinate in the direction perpendicular to the strip surface,  $H_o^{(2)}$  is the Hankel function of the second kind and order zero, and k is the wave number. The Rayleigh integral for the pressure  $p(x,z,\omega)$  is then:

$$p(x, z, \omega) = j\omega\rho_0 v_0 \int_{-a/2}^{a/2} -\frac{j}{2} H_0^{(2)}(kr) dx'$$
with  $r = \sqrt{(x - x')^2 + z^2}$ 
2.7

and at the surface, where z=0, we have:

$$p(x,\omega) = \frac{\omega \rho_0 v_0}{2} \int_{-a/2}^{a/2} H_0^{(2)} (k|x - x'|) dx'$$
 2.8

The average pressure over the strip surface is then:

$$p^{A}(\omega) = \frac{\omega \rho_{0} v_{0}}{2a} \int_{-\alpha/2}^{\alpha/2} \int_{-\alpha/2}^{\alpha/2} H_{0}^{(2)}(k|x-x'|) dx' dx$$
 2.9

Because of the symmetry in the integration domain we may replace the integral in (2.9) by twice the integral over a triangle. Thus:

$$p^{A}(\omega) = \frac{\omega \rho_{0} v_{0}}{a} \int_{-a_{2}}^{a_{2}} \int_{-a_{2}}^{x} H_{0}^{(2)}[k(x-x')] dx' dx \qquad 2.10$$

Therefore, the acoustic radiation impedance can be expressed as:

$$Z(\omega)^{A} = Z_{0} \frac{1}{K} \int_{-K_{2}}^{K/2} \int_{-K_{2}}^{y} H_{0}^{(2)}(y - y') \, dy' dy$$
 2.11

where  $Z_0$  is the acoustic impedance of the medium, K=ka, y=kx, and y'=kx'. This form can easily be evaluated by numerical means.

# Chapter 4

# Minimally Invasive Forward-Looking Ultrasound Imaging using a Novel Miniature Catheter Tip Design

Based on:

Jovana Janjic, Merel D. Leistikow, Aimee Sakes, Frits Mastik, Nico de Jong, Johan G. Bosch, Antonius F.W. van der Steen, Gijs van Soest, "3D Imaging with a single-element forward-looking steerable IVUS catheter: initial testing," 2016 IEEE International Ultrasonics Symposium (IUS), Tours, 2016, pp. 1-4.

©2016, IEEE

Aimee Sakes, Awaz Ali, **Jovana Janjic**, Paul Breedveld, Novel Miniature Tip Design for Enhancing Dexterity in Minimally Invasive Surgery, *accepted for publication in Journal of Medical Devices*.

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# ABSTRACT

In the field of vascular interventions, forward-looking intravascular ultrasound transducers (FL-IVUS) are needed for better visualization of complex lesions, such as chronic total occlusions. In this work, we propose a strategy for 3D imaging using a single-element transducer and an optical shape sensing fiber (OSS) in a novel multisteerable catheter tip that has an outer diameter of 2 mm (Ø1 mm lumen) and it consists of 4 steerable segments with 2 degrees of freedom (DOF) each. We evaluate the performance of the integrated device by imaging a sixwire phantom submerged in water. While steering the catheter tip across the wires, ultrasound and OSS data are acquired continuously. We combine the distance information obtained from the ultrasound data with the tip position and direction obtained from the OSS data to reconstruct the wires in 3D space. We quantify the accuracy of the imaging technique by the distance between the wires, and find a mean relative error of 36%. We discuss how this estimate can be further improved by modifications of the probe. This proof-of-principle test demonstrates the feasibility of FL-IVUS imaging using a single-element transducer integrated in a steerable catheter together with an OSS fiber.

## **4.1 INTRODUCTION**

Chronic total occlusions (CTOs) are complex intravascular lesions that occur throughout the arterial system and completely block the blood flow in the artery [23]. These lesions can be found in the relatively small coronary arteries, but also in the larger peripheral arteries. CTOs are challenging to treat percutaneously and the major reason for the low procedural success rate (70–75%) is the inability to cross the lesion with a guidewire. This is caused by the lack of accurate catheter steering and the lack of proper visualization of the occluded site [89]. In fact, conventional angiographic imaging and commercially available side-looking intravascular ultrasound transducers fail to provide adequate information about the best entry point or composition of the CTO.

To improve the CTO treatment and aid the crossing procedure, forward-looking intravascular ultrasound (FL-IVUS) catheters can provide useful information for intervention guidance. Intravascular imaging catheters need to be small, with practical upper limits on the diameter of 1 mm in coronary vessels and 2 mm for the larger and less tortuous peripheral arteries. Several FL transducers have been proposed and investigated in the past decades. Past designs have explored intravascular catheters with 2D beam scanning mechanisms, either by rotating a single element transducer [24,25], or by incorporation of a FL linear phased array [29,73]. However, the complex rotating mechanism makes the whole catheter bulky and unstable, and interconnection of multi-element transducers in the limited space of an imaging catheter is challenging to fabricate. Capacitive micromachined ultrasonic transducers (CMUT) have been investigated for FL-IVUS as well, offering flexibility in element shapes and layout, such as ring arrays [42,90,91]. This allows scanning of a different (more 3D-like) volume, but the devices are faced with much the same hurdles as linear arrays.

In this work, we consider an alternative strategy for 3D imaging in a FL approach. We developed a support catheter which integrates three components: a single-element ultrasound transducer mounted at the tip of the catheter, an optical shape sensing (OSS) fiber that reconstructs the 3D shape of the catheter, and a novel steering mechanism. While steering the catheter, ultrasound A-lines can be acquired and mapped to a specific position in 3D space using the information from the OSS data. To test the functionality of the integrated device, we performed a preliminary imaging test using a wire phantom submerged in water.

In the next sections, we first present the design of the multisteerable device followed by the description of the integration of the ultrasound transducer and the OSS fiber. We then present the imaging experiment and the results obtained.

# **4.2 THE MULTISTEERABLE TIP**

### 4.2.1 Tip Design

The multisteerable instrument tip, named *Accura*, has been developed at Delft University of Technology based on the patented cable-ring mechanism [92,93] and an early mutlisteerable device named *Multiflex* [94]. Briefly, the cable-ring mechanism employs two concentric coil springs and a ring of cables in between to achieve tip deflection. This mechanism is suitable for steering multiple, stacked segments as demonstrated in the *Multiflex* device, where five segments with an outer diameter of 5 mm and an inner lumen of 1 mm enable 10-DOF motion. Moreover, in the *Multiflex* device, the spring coils are replaced with rigid incompressible components increasing the tip stiffness.



**Figure 4.1** Schematic Illustration of the Multisteerable Tip of *Accura*. The tip of *Accura* has and outer diameter of Ø2 mm and consists of four 2-DOF steerable segments; giving a total of 8 DOF, allowing for forming complex shapes and single radius curves with a total curve angle of up to 90°. Each steerable segment consists of two rounded slotted cylinders (blue) and two rings (green) in which the rounded elements rotate; forming ball-and-socket type joints. The slots in the blue element allow for the attachment of four Ø0.2 mm stainless steel cables (yellow) using glue and provide cable guidance.

The novel multisteerable tip *Accura* has a smaller diameter (Ø2 mm, see Figures 4.1-4.2), reduced number of components and improved manufacturability compared to the *Multiflex* device. *Accura* contains four 2-DOF steerable segments, resulting in an 8-DOF multisteerable tip. Each steering element allows for 22.5° deflection, allowing for motions along complex curvatures and single radius curves with a total curve angle of 90°, which is comparable to currently available steerable instruments. Each steerable element (see Figure 4.1) consists of two rings (Ø2 mm, green in Figure 4.1) and two rounded slotted cylinders (Ø2mm, Ø<sub>lumen</sub> =1 mm, blue in Figure 4.1). Every other slotted cylinder is connected to four Ø0.2 mm stainless steel cables (yellow in Figure 4.1), giving a total of  $4 \times 4 = 16$  steering cables. In Figure 4.1, *Accura* is illustrated forming a 90° single-radius curve, a complex S-curve in one plane, and a complex curve in 3D.

In comparison to *Multiflex*, the segmented structure in *Accura* has been simplified in three ways: 1) the pill-like structure, in the old design consisting out of three elements, is now made out of one single grooved part combining cable alignment with fixation, 2) the holes for cable alignment and the spherical recesses present in the *Multiflex* design are removed, and 3) the outer spring is removed as the relative stiffness of the used stainless steel cables, as compared to their free length, is in this miniaturized design so high that they will stay within the grooves even at the maximum deflection, requiring no need to cover them and limit their outward motion (see for example Figure 4.2). These simplifications allow for easier manufacturability at an extremely small size and a substantially higher aspect ratio between lumen and outer diameter: from 5:1 in *Multiflex* to 2:1 in *Accura*. As a result, *Accura* is to our knowledge world's thinnest multisteerable 8-DOF MIS instrument [95,96].



**Figure 4.2** Multisteerable Tip of *Accura* illustrating different curvatures. Top: Complex single-plane curve side view. Middle: Single-plane 90° single-radius curve top view. Bottom: Complex 3D-curve side view. The match is illustrated for scale purposes.

## 4.2.2 Controlling the Multisteerable Tip

Enhancing the steerability of a multisteerable instrument increases its complexity, especially when multiple DOF need to be controlled simultaneously [97]. In order to explore the possibilities of the *Accura* mechanism, the multisteerable tip was combined with a rigid hollow shaft ( $\emptyset$ 2 mm, *L*=200 mm) and an experimental steering unit (see Figure 4.3). The steering unit allows for precise and independent control of all the steering segments in the tip using four individual control elements (one for each element) and can be easily fastened to a standard metric breadboard or optical table. In each of the control elements four steering cables are connected using a dedicated clamping mechanism (red and green in Figure 4.4). Each control element independently controls the sideways and upwards/downwards motion of one segment using two "dials" (green and blue in Figure 4.4). The dials also incorporate a locking mechanism (purple in Figure 4.4) to lock the tip position in place.



**Figure 4.3** Total *Accura* Prototype. The *Accura* steering unit is designed to be easily attachable to a breadboard or optical table using two M6 screws. Each segment is controlled individually using 4 control elements. Each element can be locked individually to fixate the tip at any given shape or allow for a scanning motion. Amplification is minimized to allow for direct and precise control of the tip

### 4.2.3 Bending Stiffness Determination

While performing medical procedures, high bending stiffness is required to prevent parasitic tip movement by external forces and unwanted tip deflection. In order to determine the bending stiffness range of *Accura*, a proof-of-principle experiment was conducted with the cable tension set to 0 and 0.5 N (see Figure 4.5). A lateral force (*F* [N]) was applied to the tip using a mass of 2, 5, 10, and 15 grams. Subsequently, the lateral deflection ( $\delta$  [mm]) was measured using a laser interferometer (*OptoNCDT 1402*, Micro-epsilon, maximum resolution 0.6 µm). Each measurement was performed 3 times. From this data, for each of the two cable pretensions (i.e., 0 N and 0.5 N), the overall bending stiffness was calculated by dividing the exerted force

*F* [N] by the deflection  $\delta$  [mm] of the steerable tip. At the 0 N cable pretension, a bending stiffness of approximately  $5.0 \cdot 10^{-3} \pm 0.6 \cdot 10^{-3}$  N/mm was measured. The bending stiffness could be increased to approximately  $0.10 \pm 0.01$  N/mm at a cable tension of 0.5 N. Note that these are approximate calculations, as we did not account for internal friction in the tip.



**Figure 4.4** Exploded View of the Control Elements of the Handle. A control element consists of an: outer cardan part (blue; with a dial (black) controlling sideways motion), inner cardan (green; with a dial (black) controlling upwards/downwards motion), conus part (red; that guides the cables), two axles (orange), 4 screws (not indicated), and a scale (transparent). The tip position can be locked in place by tightening the knobs on the dials (purple). The cables are clamped between the inner cardan (green) and conus part (red). To allow for smooth movement of the control elements, a bearing (yellow) is placed over the main axis of the outer cardan part.



**Figure 4.5** Bending Stiffness Measurement Facility. The measurement set up consisted of the multisteerable instrument, a guiding structure to prevent movement of the instrument shaft, a mass connected to the tip to exert a sideways external force on the multisteerable tip, and a laser interferometer *(OptoNCDT 1402, Micro-epsilon)* to measure the tip deflection.

# **4.3 ULTRASOUND AND OSS INTEGRATION**

In the inner lumen of the catheter, we integrated the OSS fiber and the coaxial cable of the single-element transducer. The ultrasound transducer has a square aperture (1.4 mm side) and is embedded in a cylindrical housing mounted at the tip of the catheter (Figure 4.6) with the coaxial cable (Ø160 µm) guided through the inner lumen of the catheter. The OSS fiber (Philips Research, In-Body Systems, High Tech Campus Eindhoven) is 200 µm in diameter. To protect the fiber a hollow flexible tube (500 um in diameter) is guided all the way through the inner lumen of the catheter and its extremity is fixed at the back of the transducer housing. The fiber is then inserted in the tube and advanced manually until it reaches the back of the transducer. The fiber has 4 inner cores, with one core running along the center line and three cores wound around it in a helical fashion. Fiber Bragg gratings (FBG) are patterned along the whole length of the fiber. The fiber used in this study has 6535 sensing points, spaced by 96 µm, for a total length of approximately 63 cm. When bending the fiber, the deformation of the cores induces a variation in the refractive index of the FBG. Using optical frequency-domain reflectometry, the local strain can be measured and the 3D shape of the fiber can be reconstructed [98]. To obtain the 3D shape, the position of each sensor point is computed based on the preceding point. This means that the error in precision increases with the length of the fiber. To minimize this error, in our work we reconstruct only the last 30 mm of the fiber, which correspond to the steerable length of the catheter.



Figure 4.6 Catheter steerable tip and single-element transducer compared to a match.

# **4.4 IMAGING EXPERIMENT**

The integrated device is tested by imaging a wire phantom submerged in a water tank. The wire phantom (Figure 4.7) consists of 6 parallel tungsten wires with a diameter of 40  $\mu$ m. The catheter tip with the transducer is submerged in water and positioned relative to the phantom as shown in Figure 4.8.

To reconstruct the tungsten wires in 3D space, we stream the OSS and ultrasound data to the acquisition systems, while steering the catheter tip twice across the wires. The ultrasound excitation used is a 2 cycle sinusoidal pulse at 14 MHz with a pulse repetition frequency of 300 Hz, while the OSS system acquires shapes at 60 Hz and stores the data on a local computer. The total acquisition time is 60 s. We process the data in Matlab to combine the distance information from the ultrasound data and the tip position and orientation from the OSS data. First, we filter the RF ultrasound data with a bandpass filter (5–25 MHz) and average the OSS data to reduce the noise. Further, in each trace we search for the signal from the wires by computing the maximum of the correlation between the trace and the standard wire response. The wire response has been extracted from the first 2000 A-lines acquired with the transducer beam pointing at the wire, without moving the catheter. We further consider only the A-lines with a maximum correlation value above 0.5 and the corresponding lag is used to estimate the distance. We then plot the intensity values of the wires in 3D space based on their distance from the tip and the position and direction obtained from the OSS data. This leads to two clouds of points per wire, one for each crossing. Finally, the point with maximum intensity is selected from each cloud and a line is fitted through each pair.

To measure the distance between the estimated wires we define the mean direction of all the fitted lines and we compute, for each pair of points, the coordinates of the middle point. We then consider six parallel lines with direction equal to the mean direction, each passing through one of the middle points. We compute the mean average distance between the lines as an estimation of the wire distance and compare it to the expected distance based on Figure 4.7.



Figure 4.7 Drawing of the wire phantom with the relative wire distances.

Figure 4.9(a) shows the envelope signal for each A-line. Besides the signal from the wires, also the signal from the back plate of the phantom is visible at the beginning and at the end of the ultrasound acquisition. Figure 4.9(b) shows few example shapes of the last 30 mm of the catheter and the corresponding tip direction, while Figure 4.9(c) shows the amplitude of the signal from the wires plotted in 3D space based on the OSS data.

The reconstruction of the six wires is shown in Figure 4.10 together with the catheter tip position and direction. The estimated average distance between two wires is 1.4 mm, which compared to the distance expected from the phantom design leads to a mean relative error of 36%.



**Figure 4.8** Imaging set up. The catheter tip is positioned such that the phantom is located in front of the single-element transducer.



**Figure 4.9** (a) Ultrasound A-line acquired while steering the catheter tip. The signal from the wires appears at depths between 10 and 20 mm, while the back plate of the phantom is visible at 30 mm at the beginning and at the end of the acquisition. (b) Examples of 3D shapes of the steerable tip (black) with direction (red arrows). (c) Intensity of the signals from the wires plotted in 3D as points together with the distance from the tip (light gray lines).



**Figure 4.10** 3D reconstruction of the six tungsten wires. The lines are fitted through the light green points identifying the location of the maximum intensity. The catheter tip direction, one every 20 shapes, is also shown (red arrow).

# 4.5 DISCUSSION

### 4.5.1 Intuitive Control of the Multisteerable Tip

For the clinical application of Accura, redesign of the steering unit is a necessity to allow for intuitive and single-handed control in a clinical setting. In a study of Chunman et al. [99], a number of control strategies are categorized into three main groups: parallel single-segment control, serial single-segment control, and integrated single-segment control (see Figure 4.11). In parallel single-segment control, each segment has its own controller, and each controller functions independently from the other controller. This type of control strategy was applied in the steering unit of Accura. In a serial single-segment controller, the motion of each controller depends closely on the motion of the adjacent ones, which is the case in *Multiflex's* handle. Finally, in the *integrated single-segment controller*, also called follow-the-leader controller, only the first (most distal) segment is steered, while the additional segments follow the "leader segment" passively. Even though integrated single-segment control may be preferred from an operator's perspective, it does not allow for actively controlling the full range of motion of the multisteerable tip. Therefore, in the future a combination of the parallel single-segment control and the integrated single-segment control strategies will be looked into, in which the surgeon only needs to steer the most distal segments while the rest of the segments remain in place, similar to the way in which a snake travels through a rocky terrain. This approach will give the surgeon the most freedom in steering the device, while at the same time not making it unnecessarily difficult by the need to control multiple segments simultaneously.

### 4.5.2 Multiple Functionalities in the Multisteerable Tip

As space is often limited, research should also be focused on the integration of multiple functionalities into one instrument. These functionalities should, in the authors' opinion, include: steerability, visualization, and the ability to guide a tip tool towards the operation site. First steps have been made in *Accura* to integrate multisteerability with a forward-looking ultrasound transducer and OSS fiber to visualize the operation area. The information from the forward-looking ultrasound transducer and OSS fiber can be used for navigational purposes but also for tissue characterization, which can be helpful to distinguish healthy from diseased tissues. Furthermore, since the 3D shape of the instrument's tip and shaft is reconstructed using OSS, the use of X-ray for visualization of difficult-to-reach operation areas or blood-filled spaces may be minimized, as the interventionists can obtain additional information from the combined forward-looking ultrasound and OSS data to image the surroundings and determine their position. Next to visualizing the operation area, the ability to guide a tip tool, such as a scalpel, grasper, or biopsy forceps, through the instrument, or fix one such a tool to the distal tip, will aid the surgeon in reaching the intended goal of the MIS, such as tumor removal or taking a biopsy, without the need for an additional instrument.



**Figure 4.11** Multisteerable Control Strategies. Top: *parallel single-segment control*, in which each segment is controller. This way, each segment can be controlled independently. Middle: 2, in which each segment is controlled by a separate controller, but the motion of the controller (and thus the tip) is dependent on the position of the adjacent controllers. Bottom: *integrated single-segment control*, in which only the most distal segment is steered, while the other segments follow the "leader segment" passively

### 4.5.3 Forward-looking Imaging Limitations

We successfully reconstructed six tungsten wires using the ultrasound information from a single-element transducer combined with the 3D position information of the catheter tip obtained from the OSS data. The catheter is steered across the wires twice enabling to identify two points for each wire. By fitting a line through the points it is possible to reconstruct position and orientation of the wires.

We reconstruct only the last 30 mm of the fiber to minimize the error in shape reconstruction. Moreover, by averaging the shapes, we reduce the jittering error. However, the reconstructed wires are not all parallel. One reason for this is that the tungsten wires may not be mounted perfectly straight on the plastic support. Secondly, and more importantly, we were able only to access a limited scan pattern in this experiment, caused by difficulties in controlling the steering mechanism. The specific pattern described by the steerable catheter allowed us to cross the wires only twice with the imaging beam, without much lateral displacement. This lead to the identification of two points for each wire which are very close to each other, affecting the accuracy of the line fitting. In addition, the ultrasound transducer might insonify the wires under an angle leading to errors in the distance estimation.

Moreover, the exact position of the tip of the fiber in the catheter is very difficult to estimate and it might change during the measurement, causing errors in the position estimation of the wires in 3D space. Adjustment and improvement to the catheter tip and handle design are ongoing.

### 4.6 CONCLUSION

In conclusion, we have demonstrated the feasibility of performing 3D imaging with a single-element forward-looking steerable IVUS catheter using OSS. Future work will include optimization of the steering device and imaging of more complex structures like tissue mimicking phantoms. In an actual application for intra-arterial scanning, there will not be any discrete target. Rather, we will investigate how to best visualize the 3D data of complex mixed tissue by mechanically steering the ultrasound transducer across the target and integrating the ultrasound information with the OSS data. Proper visualization and mapping of heterogeneous tissue structures, such as CTOs, will aid the physician during the crossing procedure and improve the treatment success rate.

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# Chapter 5

# Sparse Ultrasound Image Reconstruction from a Shape-Sensing Single-Element Forward-Looking Catheter

Jovana Janjic, Frits Mastik, Merel D. Leistikow, Johan G. Bosch, Geert Springeling, Antonius F. W. van der Steen, Gijs van Soest, Sparse Ultrasound Image Reconstruction from a Shape-Sensing Single-Element Forward-Looking Catheter, *IEEE Trans. On Biomedical Engineering, in press.* 

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# ABSTRACT

**Objective:** Minimally invasive procedures, such as intravascular and intracardiac interventions, may benefit from guidance with forward-looking (FL) ultrasound. In this work, we investigate FL ultrasound imaging using a single-element transducer integrated in a steerable catheter, together with an optical shape sensing (OSS) system.

**Methods:** We tested the feasibility of the proposed device by imaging the surface of a tissuemimicking (TM) phantom and an ex vivo human carotid plaque. While manually steering the catheter tip, ultrasound A-lines are acquired at 60 Hz together with the catheter shape from the OSS system, resulting in a 2D sparse and irregularly sampled data set. We implemented an adaptive Normalized Convolution (NC) algorithm to interpolate the sparse data set by applying an anisotropic Gaussian kernel that is rotated according to the local direction of the catheter scanning pattern. To choose the Gaussian widths tangential ( $\sigma_t$ ) and normal ( $\sigma_n$ ) to the scanning pattern, an exhaustive search was implemented based on RMSE computation on simulated data.

**Results:** Simulations showed that the sparse data set contains only 5% of the original information. The chosen widths,  $\sigma_n = 250 \,\mu\text{m}$  and  $\sigma_t = 100 \,\mu\text{m}$ , are used to successfully reconstruct the surface of the phantom with a contrast ratio of 0.9. The same kernel is applied successfully to the carotid plaque data.

**Conclusion:** The proposed approach enables FL imaging with a single ultrasound element, mounted on a steerable device.

**Significance:** This principle may find application in a variety of image-guided interventions, such as chronic total occlusion (CTO) recanalization.

# **5.1 INTRODUCTION**

Forward-looking (FL) ultrasound catheters can provide guidance during minimally invasive procedures, such as recanalization of arterial chronic total occlusion (CTO). In such cases, an atherosclerotic plaque has grown to completely block the vessel. Conventional intravascular ultrasound (IVUS) imaging in a side-looking geometry is of limited use. CTOs are complex lesions with heterogeneous tissue composition consisting of lipids, organized thrombus, calcium and collagen-rich fibrous tissue [23]. They can develop in the relatively small coronary arteries, but also in the larger peripheral arteries. In order to aid the CTO crossing procedure, FL ultrasound may provide useful information, such as the best entry point or local tissue composition [72, 100]. Forward-looking transducers may be beneficial for other applications as well, such as intracardiac echography for the diagnosis of cardiac arrhythmias [87,101], or guidance during transjugular intrahepatic portosystemic shunt (TIPS) procedures, where a shunt is created in the liver to decrease the high blood pressure in patients with portal hypertension [102,103] and the portal vein needs to be located under image guidance.

Catheter-based FL ultrasound technology is still in development and several concepts have been pursued in experimental settings. Early work explored single element transducers mounted on a rotating cam assembly [24,25]. This mechanism is rather complex and makes the whole catheter bulky and unstable. In other studies, linear transducer arrays have been investigated as a way to provide 2D imaging ahead of the catheter tip without mechanical scanning [29,73]. The major problems with this approach are the limited space available and the constraints in PZT fabrication that make the integration into catheters very challenging. Capacitive micromachined ultrasonic transducers (CMUT) offer a great advantage in terms of fabrication over PZT since very small dimensions and arbitrary geometries of the elements can be achieved using lithography techniques and micro-fabrication processes [37]. This technology has been used by several groups in the attempt of developing catheter-based FL ultrasound transducers [41,42,44,87,90,91]. However, integration and miniaturization of electronics remains the major problem.

In this work, we propose an alternative approach to achieve FL ultrasound imaging by integrating three components in one device: a single-element ultrasound transducer, an optical shape sensing (OSS) fiber and a catheter that can be steered along two directions in one plane. While steering the catheter, single ultrasound A-lines can be acquired at different locations together with the catheter 3D shape obtained from the OSS data. Having a single-element transducer reduces the complexity of integration, whereas the position and direction information from the OSS data allow to combine the individual A-lines in 3D space. This paper builds on an earlier initial work by our group to integrate a similar device with preliminary results on target reconstruction [104,105].

Since the steering is performed manually and is limited to specific patterns due to the catheter design, a sparse and irregularly sampled data set is obtained. To reconstruct FL images of the target from sparse data, an interpolation algorithm is required. In previous works different approaches to sparse image reconstruction have been proposed [106,107]. A method that has shown good performance is Normalized Convolution (NC) [108-110]. NC takes into account the certainty of the signal and applies a diffusion kernel to fill the empty spaces in the image. We implement an adaptive interpolation method based on NC, where the diffusion kernel changes orientation dynamically depending on the local scanning pattern of the catheter. A similar approach was used in [111] with the difference that there the diffusion kernel was adapting to the image gradient. We tested our adaptive NC method both on simulated and real data, focusing on the reconstruction of the front surface of a tissue mimicking (TM) phantom and of a carotid plaque obtained from endarterectomy.

### 5.2 METHODS

### 5.2.1 Experimental configuration

To perform FL ultrasound imaging with a single element transducer on a steerable catheter, we have developed a prototype device based on a commercially available bi-directional steerable guiding sheath (8.5 F, MobiCath, Biosense Webster). The dimensions of this sheath are suitable for peripheral vasculature applications, but exceed the allowable dimension for coronary artery access. By rotating the catheter handle (Figure 5.1(a)), the tip can be steered in two directions in one plane and by rotating the entire catheter, the steering plane can be chosen. At the tip of the catheter, which is free from steering cables, we place a single-element ultrasound transducer (Figure 5.1 (b)). The ultrasound transducer is a circular PZT element with a diameter of 1 mm and 25 MHz center frequency, mounted in a custom-made housing (Figure 5.1 (c)). The single element transducer can also be mounted on steerable catheters with smaller outer diameter. The transducer is electrically wired with a 160 µm coaxial cable that is guided through the inner lumen of the catheter. The OSS system (a research prototype developed at Philips Research, Eindhoven, NL) consists of an optical fiber with a diameter of 200 µm, connected to a scanning laser system [94,104]. The fiber has four inner cores, one along the longitudinal axis and three wound around it in a helical fashion. Fiber Bragg Gratings (FBG), which consist of periodic variation in refractive index, are patterned along the entire length of the fiber (1.8 m). Bends in the fiber induce local stretching and compression of the period of the FBG, allowing measurements of the local strain based on optical reflectometry in the frequency domain [48]. The central wavelength of the scanning laser is set to 1545 nm, and is swept over 17 nm. The spatial resolution along the sensor is 50 micrometer. The 3D shape is obtained by performing a distributed measurement of the applied strain to each of the four cores. By integrating forward along the fiber from the first sensing point, which is fixed in the reference frame, and combining the information from all four cores, the geometry of the sensor can be obtained. Since the position of each sensor point is estimated based on the previous one, the precision of the estimation decreases with increasing length of reconstruction. Additional details about the working principle can be found in [48,49]. A helical hollow strand (HHS) tube (0.5 mm in diameter) protects the OSS fiber. The HHS tube is guided through the catheter lumen and one end is attached to the back of the transducer housing, whereas the other end is accessible to insert the fiber.



**Figurre 5.1** (a) Handle of the steerable catheter (MobiCath, Biosense Webster) (b), tip of the steerable catheter with the single element transducer (inner gray circle) (c), schematic drawing of the transducer housing with the HHS tube for the OSS fiber, and (d) the measurement set-up, with TM phantom and carotid plaque.

We simultaneously acquire OSS data and US data while steering the catheter. A 2-cycle sinusoidal pulse at 40 Vpp with a center frequency of 25 MHz is sent to the transducer using a waveform generator (Tektronix, 3102) and an RF power amplifier (ENI, 310L), whereas the received signal is amplified by 40 dB (Hewlett-Packard, 461A) and digitized with a sampling frequency of 350 MHz (Acqiris, DP310). The OSS data are acquired at 60 Hz via the user interface developed by Philips, providing the total shape and curvature of the fiber. A Matlab script provides a real time feedback of the covered area by plotting the catheter tip coordinates in 2D and the corresponding ultrasound A-lines. Further signal processing and image reconstruction are performed off-line in Matlab (R2016a), using a personal computer with an Intel i7-3770, 3.4 GHz, 4-core processor. Figure 5.1(d) shows a schematic of the complete set-up used in our measurements. The imaging target is submerged in a water tank into which the steerable catheter is inserted from the side of the tank, clamped using an O-ring and guided into a plastic tube that confines and limits its movements.

As imaging target we used a tissue mimicking (TM) phantom and an ex vivo human carotid plaque (Figure 5.1(d)). The agar-based phantom has a cylindrical geometry with a diameter of 16 mm and has hollow channels running through the total length and steel spheres embedded at the surface. The *ex vivo* human carotid plaque is obtained following endarterectomy procedures, under a research protocol sanctioned by the Medical Ethics Committee of the Erasmus Medical

Center MC (MEC-2008-174). The human specimen was snap frozen in liquid nitrogen within 10–20 min following endarterectomy and stored at 80°. The tubular specimen was then thawed, cut open along the longitudinal axis, mounted on a rubber support and submerged in saline solution with the inner plaque layer facing the ultrasound probe. A 3D ultrasound image of the plaque was obtained as ground truth using a high frequency pre-clinical ultrasound scanner operated at 21 MHz (Vevo 2100 with MS250, VisualSonics Inc., Toronto, ON, Canada). In addition, we scanned the plaque with the 25 MHz single element transducer clamped to a positioning stage and mechanically scanned along a regular grid pattern with a step size of 100  $\mu$ m in both x and y directions.

### 5.2.2 Ultrasound data processing

Each ultrasound A-line is filtered in Matlab using a Butterworth filter of order 5 with a bandwidth between 15 MHz and 45 MHz. After filtering, the envelope is computed using the Hilbert transform. We focus on the reconstruction of the surface of TM phantoms with high-intensity hard spots and low-intensity channels and on visualization of the surface of the carotid plaque. The surface is obtained from the ultrasound A-lines by intensity thresholding: for each A-line, if the signal is above the noise level, we consider the corresponding distance value as the surface location. The noise level was measured considering the average signal in the water region before the surface of the phantoms. For points where we do not find a distance value, like the phantom channels, we linearly interpolate between known neighbors. Next, we consider the envelope signal corresponding to a 400 µm thick slice of the imaged object starting from the surface and we project the envelope mean values on a 2D plane based on the tip position and direction obtained from the OSS system and based on the tip-to-target distance computed with ultrasound. At the same time, we also store the tip-to-target distance for each tip position. When computing the tip-to-target distance we apply an offset of 5 mm to take into account that the OSS fiber tip is not exactly at the same axial location as the transducer (see Figure 5.1(c)). The resulting 2D data sets form two sparse and irregularly sampled maps: one for the subsurface intensities and one for the tip-to-target distances.

### 5.2.3 Estimation of catheter tip position and direction

For each acquired data set, the 3D coordinates of the sensing points along the fiber and their curvature are provided by the OSS system. We use the curvature plot of each fiber shape to find the point that corresponds to the location where the catheter is clamped to the side of the water tank. At this location, a peak in the curvature is visible, due to the deformation applied by the clamping mechanism. Knowing that the clamping point is not moving, we can use it as a new reference point and reduce the reconstructed length of the fiber, thus decreasing the error in position estimation. Next, we compute the catheter tip position and direction relative to the new reference point. The obtained coordinates are further smoothed applying a spline fit in order to obtain a more regular scanning pattern and remove possible outliers.

### 5.2.4 Sparse data from simulated target

To simulate a sparsely sampled data test, we use a measured scanning pattern obtained from one of the OSS acquisitions and simulate an imaging target in a plane perpendicular to the mean direction of the catheter tip, at a distance of 4 mm. The simulated target consists of a regular grid of holes and strong reflectors (Figure 5.2(a)). To simulate a target that is sampled by an ultrasound transducer, we convolve the target image with the ultrasound transducer point spread function (PSF) at 4 mm (Figure 5.2(b)), which has a -3 dB diameter of approximately 500 µm. An image of the target after PSF application is shown in Figure 5.2(c). We sample the simulated target after PSF application by projecting the catheter tip position on the target plane.



**Figure 5.2** (a) Simulated imaging target consisting of a regular grid of holes and strong reflectors (b), ultrasound PSF at 4 mm (c), imaging target convolved with the ultrasound PSF at 4 mm.

### 5.2.5 Image reconstruction with adaptive Normalized Convolution (NC)

NC is a technique used for interpolation of sparse, irregularly sampled data based on their certainty by applying a predefined diffusion kernel [108-110]. More in detail, considering a sparse image S and its certainty map C, the normalized convolution result R is obtained as follows:

$$R = \frac{(S \cdot C) \otimes G}{C \otimes G}$$
 5.1

where G is the diffusion kernel and  $\otimes$  indicates convolution. The certainty map represents the certainty for the image samples, which is 1 for each sampled position and 0 everywhere else. The diffusion kernel is usually represented by the 2D Gaussian function

$$G(x,y) = \exp\left[-\left(\frac{x}{\sigma_x}\right)^2 - \left(\frac{y}{\sigma_y}\right)^2\right]$$
5.2

where the parameters  $\sigma_x$  and  $\sigma_y$  represent the width of the kernel in the x and y direction, respectively. In conventional NC the diffusion kernel is an isotropic Gaussian, with  $\sigma_x = \sigma_y$ .

In our study, the sparsely sampled data obtained from both the simulations and the real measurements are used to generate a sparse matrix image with a pixel size of 40 x 40  $\mu$ m. In the simulations the sparse image contains the intensity values sampled using a measured scanning patterns obtained from one of the OSS acquisitions. For the real measurements, beside the intensity sparse map, we also define the sparse map of the tip-to-target distances. Therefore, during the scan, we accumulate the amplitude value per pixel in  $S_i$  and the distance values per pixel in  $S_d$  while keeping track of the number of measurements in each pixel of C. As diffusion kernel we apply an anisotropic Gaussian function that is rotated according to the local scanning pattern of the steerable catheter. The idea is to diffuse the sparse intensity map, the sparse distance map and the certainty map taking into account that we have a very dense sampling along the scanning path is due to the high sampling rate (60 Hz) combined with the slow motion of the catheter (approximately 0.6 mm/s). More in detail, the adaptive diffusion kernel is defined as:

$$G(x_{\theta}, y_{\theta}) = \exp\left[-\left(\frac{x_{\theta}}{\sigma_{t}}\right)^{2} - \left(\frac{y_{\theta}}{\sigma_{n}}\right)^{2}\right] \text{ with } (x_{\theta}, y_{\theta}) = (x, y) \bullet \begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix}$$
 5.3

where  $\sigma_t$  represents the width of the kernel along the direction tangential to the path,  $\sigma_n$  represents the width of the kernel along the direction normal to the path and  $\Theta$  is the angle between the direction normal to the path and the axis of the image. A schematic drawing is shown in Figure 5.3. Both the sparse map (intensities and distances) and the certainty map are locally convolved with the adaptive kernel following 5.5.1 and the resulting diffused maps are divided to obtain the adaptive NC result for both the intensities and the distances.

To find the optimal parameters for  $\sigma_t$  and  $\sigma_n$  we perform an exhaustive search using the simulated sparse intensity data. Both  $\sigma_t$  and  $\sigma_n$  are varied between 50 and 500 µm and, for each possible combination of parameters, the adaptive NC result is computed. The obtained image  $I_{aNC}$  is compared to the simulated ground truth image  $I_{truth}$  (Figure 5.2(a)) by computing the root-mean-square error on the linear intensities as:

RMSE=
$$\sqrt{\frac{1}{N_{\text{pixel}}} \sum_{n=0}^{N_{\text{pixel}}^{-1}} (l_{aNC}(n) - I_{truth}(n))^2}$$
 5.4

The values of the parameters  $\sigma_t$  and  $\sigma_n$  for which we get a low RMSE, are then used in the adaptive NC to reconstruct the image of the simulated target, the TM phantom and the carotid plaque. For the TM phantom we apply adaptive NC with different combinations of parameters  $\sigma_t$  and  $\sigma_n$ , and we compute the contrast ratio CR, which is defined as [59]

$$CR = \frac{\mu_b - \mu_c}{\mu_b + \mu_c}$$
 5.5

where  $\mu_b$  is the average signal amplitude from the tissue-mimicking material and  $\mu_c$  is the average signal amplitude in the hollow channel region.



**Figure 5.3** Schematic drawing of a scanning pattern (dotted line) on the imaging target plane with a rotated anisotropic kernel (gray area). Dotted line represents the scanning path, gray ellipsoid represents the diffusion kernel;  $\sigma_t$ =width of the kernel along the direction tangential to the path ,  $\sigma_n$ =width of the kernel along the direction normal to the path,  $\Theta$ =angle between the direction normal to the path and the axis of the image.

### 5.3 RESULTS

Figure 5.4(a) shows a typical 3D reconstruction of the entire OSS fiber, including the portion integrated into the steerable catheter, which is clamped at the side of the water tank. The z-axis of the 3D shape reconstruction corresponds to the catheter long axis. Figure 5.4(b) shows the curvature of three different shapes acquired while manually scanning the catheter. The three curvature plots are representative of the initial, intermediate and final shape of the catheter during one acquisition. The curvature peaks around 0.9 m coincide with the catheter handle location, whereas the peak indicated by the red mark corresponds to the location at which the catheter has been clamped to the side of the water tank. Since only the tip of the catheter is steering, we use the catheter clamping point as a new reference point for the OSS reconstruction. Figure 5.4(c) shows the shapes of approximately the last 10 cm of the catheter during one acquisition, whereas Figure 5.4(d) shows the tip position and direction after spline filtering. The results presented in the following sections where achieved with a non-optimized algorithm that computes the adaptive kernel at a rate of 68 data points (tip positions with corresponding intensities) per second. Since each acquisition contains 15000 tip positions and relative ultrasound intensities, the final adaptive NC images were computed in 3 min and 40 sec.



**Figure 5.4** (a) A representative 3D shape of the fiber inserted into the steerable catheter (b), curvature plot for three different shapes: first, middle and last shape during one measurement (c), 3D shapes of the last 10 cm of the steerable catheter during one measurement (d), Catheter tip position and direction (black arrows) in 3D. The gray plane delineated by red lines indicates the imaging plane. The red asterisk indicates the location at which the catheter has been clamped to the side of the water tank.

# **5.3.1** Adaptive Normalized Convolution on simulated data: optimal parameter choice

The scanning pattern shown in Figure 5.4(d) is applied to the simulated target (Figure 5.2(c)) to obtain the sparse certainty and sparse intensity maps of Figure 5.5(a) and (b). In the resulting map, only 5% of the original image information is retained. Figure 5.5(d)-(f) show the conventional NC steps where an isotropic Gaussian kernel is applied with  $\sigma_x = \sigma_y = 400 \,\mu\text{m}$ . The original simulated imaging target is also shown as a reference (Figure 5.5(c)). The optimal parameters and for the adaptive NC are found utilizing an exhaustive search. The RMSE for the different combination of parameters is shown in Figure 5.5(g). The chosen parameter combination is indicated by the white star and corresponds to  $\sigma_n = 250 \,\mu\text{m}$  and  $\sigma_t = 100 \,\mu\text{m}$ . The anisotropic kernel with the optimal parameters is applied to the sparse data resulting in the adaptive NC reconstruction shown in Figure 5.5(i).



**Figure 5.5** Standard NC: (a) simulated certainty image and (b) sparse intensity image, both with maximum value clipped for better visibility; (c) original image; (d) diffused certainty image and (e) diffused intensity image using an isotropic kernel with  $\sigma_x = \sigma_y = 400 \ \mu\text{m}$ ; (f) NC result; (g) RMSE based on linear intensities resulting from the exhaustive search for the NC parameters  $\sigma_n$  and  $\sigma_t$ . The white star indicates the chosen combination of parameters; (h) sparse intensity image diffused with the anisotropic kernel; (i) adaptive NC reconstruction.

### 5.3.2 Adaptive Normalized Convolution on phantom data

An example of ultrasound data obtained from the TM phantom during one measurement is shown in Figure 5.6. The data are shown after filtering and after envelope detection. The red line indicates the phantom surface. From the ultrasound A-lines shown in Figure 5.6 it is possible to see the gaps in the signal corresponding to the hollow channels (indicated by the yellow solid arrows) and strong reflections corresponding to the steel spheres embedded at the surface (indicated by the yellow dashed arrows).



**Figure 5.6** Ultrasound A-lines obtained by scanning the transducer across the surface of one of the TM phantoms. The red line indicates the phantom surface, the yellow solid arrows indicate some hollow channels and the yellow dashed arrows indicate some strong steel reflectors.

The average amplitude for each A-line over a depth of 400 µm from the phantom surface is mapped to a 2D plane representing the phantom surface, knowing the tip-to-target distance based on ultrasound and the tip position and direction in 3D based on the OSS data. The resulting average intensity image is shown in Figure 5.7(b). Figure 5.7(c) shows the corresponding adaptive NC result images with the chosen kernel parameters  $\sigma_n$ =250 µm and  $\sigma_t$ =100 µm. Both average intensity image and adaptive NC images are shown with a dynamic range of 40 dB. Next, we apply the adaptive NC reconstruction on the tip-to-target distance (Figure 5.7(d)) and, by combining the distance and intensity reconstructions, we obtain the surface plot shown in Figure 5.7(e), with most of the intensity values located in one plane, in agreement with the fact that the front surface of the phantom is flat.

To confirm the improved contrast obtained by applying adaptive NC with the chosen parameters over standard NC, we compute the contrast ratio for images obtained applying adaptive NC with different combinations of parameters and  $\sigma_n$  and  $\sigma_t$ . The result is shown in Figure 5.7(f), demonstrating that our choice of  $\sigma_n$  and  $\sigma_t$  lies in the region of highest contrast ratio. In fact, we achieved a contrast ratio equal to 0.9 by using an adaptive Gaussian kernel with  $\sigma_n$ =250 µm and  $\sigma_t$ =100 µm.



**Figure 5.7** (a) Schematic drawing of the phantom surface; (b) average sparse intensity image; (c) adaptive NC reconstruction of the intensity; (d) adaptive NC result on the tip-to-target distances; (e) surface plot combining the intensity and distance information; (f) Image contrast ratio obtained using adaptive NC with different combinations of  $\sigma_n$  and  $\sigma_t$  for the TM phantom. The chosen combination of parameters is indicated with a black star.

### 5.3.3 Adaptive Normalized Convolution on carotid plaque data

Figure 5.8 (a) shows the area of the carotid plaque that was imaged and scanned. The yellow arrows indicate the location of calcified areas. The surface plot obtained with the Vevo 2100 preclinical ultrasound scanner is shown in Figure 5.8(b). Figure 5.8(c) shows the image obtained by scanning the 25 MHz single element transducer over a regular grid pattern mechanically translating the transducer. Figure 5.8(d) shows the sparse distance map, whereas Figure 5.8(e) shows the intensity sparse map. Both the sparse maps are obtained by manually steering the catheter with the 25 MHz transducer at the tip and acquiring OSS data, similarly to the TM phantom measurements. The result of applying the adaptive NC on both the tip-to-target distances and on the intensity values is shown in Figure 5.8(f).



**Figure 5.8** (a) picture of the imaged portion of the carotid plaque with yellow arrows indicating calcium; (b) surface reconstruction using the Vevo 2100 scanner with a 21 MHz linear array, (c) surface reconstruction obtained by mechanically scanning the 25 MHz single element transducer on a regular grid pattern; (d) sparse distance map and (e) sparse intensity map obtained by manually scanning the 25 MHz single element transducer mounted on the steerable catheter; (f) surface reconstruction of the after adaptive NC on both the sparse distance and the intensities values.

# 5.4 DISCUSSION

In this paper, we have introduced a novel approach to catheter-based FL ultrasound imaging using a single element transducer integrated into a clinically available steerable catheter together with an OSS system. We also have demonstrated that adaptive NC is successful in interpolating the sparse and irregularly sampled data obtained by manually scanning the catheter. A single element catheter is much simpler to build and use than a multi-element one, evidenced by [29,42,87,90,91]. The only additional component needed to combine the different A-lines is the OSS fiber, which fits easily in the catheter along the transducer coaxial cable, due to its small outer diameter.

The device used in this study has an outer diameter of 4 mm, which makes it suitable for peripheral vascular interventions, such as CTO recanalization. The localization of the channels relative to the calcified regions in the plaque is important during CTO revascularization, since it can point the physician towards the best entry point, shortening the procedure and reducing the risk of complications, such as vessel perforation. In principle, both the transducer and the OSS fiber can be integrated in smaller catheters, such as those used in coronary artery interventions. However, the realization of steerable coronary catheters remains a major challenge.

We combined lateral steering and rotation, resulting in a very high spatial sampling in the central region, but very sparse sampling elsewhere (Figure 5.4(d)). The catheter used in this study is a clinically available device. Since the proposed method should be readily applicable to real patient data, using a state of the art catheter ensures easier translation of this novel concept into the clinics. More agile steerable devices may produce sampling patterns with different characteristics. We applied our experimental scan pattern to the simulated image, retaining only 5% of the original image, which further proves the steering limitation of our device.

Despite the limited information available, NC with isotropic kernel reconstructs the major features of the image as shown in Figure 5.5(f). However, using an isotropic kernel does not take into account the large difference in sampling density along the path compared to the transverse direction, which results in muted image contrast. Using an adaptive anisotropic Gaussian kernel the major features of the simulated target and the TM phantom are reconstructed, preserving the edges and reducing the blurring along the scanning path, as shown in Figure 5.7(c). This is also confirmed by the CR map in Figure 5.7(f), showing that the chosen combination of parameters and lays in the region where the highest CR values are obtained. For the phantom and the carotid plaque specimen, we can reconstruct most of the features which are not recognizable in the sparse maps. A NC algorithm with adaptive kernel was also used in [111] where the anisotropic kernel was rotated according to the gradient angle of a randomly sampled image. However, in our case, due to the particular scanning pattern and the highly sparse nature of the data, the computation of the gradient is not a suitable choice.

In this work, we have focused on the reconstruction of the imaging target surface by combining the intensity signals and the tip-to-target distance information from ultrasound. For the phantom this results in flat surface reconstruction, whereas in the ex-vivo carotid plaque measurements the surface plot shows the differences in distances for the different structures. The same method can be extended to 3D to reconstruct volumetric images of the structures ahead of the catheter.

The FL ultrasound imaging approach presented in this paper has some limitations. If certain regions of the imaging target are not scanned at all, the reconstructed image will fail in providing information about those regions (see for instance the hole at 8 o'clock in the phantom image). This means that the ability to reconstruct FL images with the proposed device requires the scanning pattern to achieve a certain minimum coverage. Steerable catheters with more flexibility than the one used in this work, as well as random probe motion of which there usually is plenty in (cardiac) interventions, may help in optimizing the coverage of the region of interest. To reduce the reconstructed length of the fiber, hence decreasing the error in position estimation, we set the reference point to correspond to the clamping point on the side of the water tank. In a clinical setting this will result in moving the reference point within the human body. Since the OSS system provides strain and curvature information over time for each sensing point along the fiber, this information could be used to identify the stable clamping point within the body and set it as reference point.

# **5.5 CONCLUSION**

In this work, we have successfully integrated a FL single-element ultrasound transducer in a steerable catheter together with an OSS fiber and we have demonstrated the feasibility of achieving FL imaging. The integrated device acquires ultrasound A-lines and OSS data simultaneously at 60 Hz, while manually steering the catheter tip. The manual scanning leads to a sparse and irregularly sampled data set, which requires interpolation to reconstruct the features at the front surface of the imaging target. We have shown that applying Normalized Convolution with an adaptive anisotropic Gaussian kernel on both the intensity and the distance data enables to generate surface plots of the imaging target. By applying adaptive NC on the phantom data, channels and strong reflectors can be identified, whereas in the carotid plaque image the strong calcified areas are successfully reconstructed.

We have shown that, with only 5 % of the original data, it is possible to reconstruct the salient features of the image, such as the location of the channels and the strong reflectors. The chosen combination of adaptive Gaussian parameters,  $\sigma_n=250 \ \mu\text{m}$  and  $\sigma_t=100 \ \mu\text{m}$ , leads to image reconstruction of the phantom surface with a contrast ratio of 0.9.

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# Chapter 7

# Speckle in Ultrasound

Based on:

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# ABSTRACT

Speckle is inherent to imaging with coherent waves. In this chapter, we introduce the basic concepts of speckle as it appears in randomly scattered wave fields. Using concepts from laser speckle originating in optics, we show how the Rayleigh distributed magnitude appears as a very general and robust result from a random phasor sum, with minimal assumptions on the statistics of amplitude and phase of the contributing phasors. Subsequently, the statistics of a scattered field with a constant background and non-uniform phase distributions are demonstrated.

By writing ultrasound scattering in a linear systems formalism, the relations that govern speckle appear straightforwardly in the field scattered by a random structure. We computed the second-order statistics to relate speckle spot size to ultrasound imaging parameters and found that the speckle size is similar, but not identical to the resolution cell dimensions. We outline the case of a small number of contributing waves, leading to the K-distribution, which is one of several heuristic models that have been implemented to descirbe various speckle scenarios. Finally, we discusse the impact of dynamic range compression on speckle statistics, which is important to describe the signal distribution in practical imaging applications.

## 7.1 INTRODUCTION

Ultrasonic B-mode imaging is fundamentally grayscale imaging: it forms an image based only on the amplitude of the detected echo. It processes the available information in the recorded echoes, which arise from scattering of the transmitted wave by acoustic discontinuities or heterogeneities in the imaged object. The translation from object features to image features is subject to the physics of the wave propagation, scattering, and detection processes that carry the information from object to the imaging system. The imaging physics introduces familiar imaging concepts like image resolution, related to system bandwidth, imaging contrast and penetration depth, related to the acoustic properties of the medium, object, and system parameters like sensitivity and noise. Image interpretation is principally an inversion of these steps: what can we infer about the object from the received echo, combined with some prior knowledge of the imaged object and the imaging system.

Speckle is one of the image features arising in the translation of object structure to an ultrasound image. It manifests itself as "spotty" variations in image intensity, giving rise to a granular appearance instead of homogeneous, flat shades of gray, as is visible in Figure 7.1. As such, speckle can severely compromise interpretation of ultrasound images, particularly in discrimination of small structures. As we will see in this chapter, it is fundamentally impossible to separate the "true" image information from speckle corrupting it. For this reason, the common term "speckle noise" is a misnomer, as there is no way to discriminate the speckle as a "noise" source from the "signal", that we are actually interested in. We will therefore not adopt the notion of "speckle as short-range correlated multiplicative noise" which is frequently applied, but does not adequately represent the complex nature of the resulting scattered field, nor does it encapsulate the statistical properties of amplitude and phase of physical speckle. The concept of speckle as noise also forgoes its use in detection of time varying phenomena: blood vessels, for instance, can be identified by their large speckle variance or short speckle decorrelation time. Nevertheless, quoting the optical physicist Joseph Goodman, to whom many of the concepts we present in this chapter are due: "Although a number of beneficial uses can be made of speckle, it usually is more of a hindrance than a help" [113].

In this chapter, we will discuss the basic physical origins of speckle and present different, but related, mathematical descriptions of speckle in simple model systems. We will start from a general description of scattered waves as phasors, deriving first order statistics applicable to both laser speckle and ultrasound speckle. Subsequently we discuss the correspondences and differences between laser speckle and ultrasound speckle, before presenting higher-order statistics and imaging implications in an ultrasound context. We also introduce the impact of the imaging system characteristics on speckle appearance and statistics.



**Figure 7.1** Ultrasound speckle. (a) Image of a CIRS ultrasound phantom, obtained with a phased array transducer with an element pitch of 295  $\mu$ m and a frequency bandwidth of 1–4 MHz. It shows speckle instead of homogeneous gray areas. (b) IVUS image of a human coronary artery specimen ex vivo, rotating single element with a bandwidth of 20–38 MHz.

# **7.2 SPECKLE OBSERVATIONS**

#### 7.2.1 Speckle in optics

Speckle arises from interference of many component waves with random phase, summed to a resultant vector with a stochastic distribution of amplitude and phase. This phenomenon was first described in the scintillation of stars in the late 19<sup>th</sup> century [114] caused by atmospheric disturbance of the planar wavefront arriving from point sources. The term "speckle" was first coined after the invention of the laser, where it is readily observable as the grainy appearance of laser light scattered off a rough surface [115], as shown in Figure 7.2.



Figure 7.2 Speckle generated by a Helium-Neon laser transmitted through a diffuser.

These observations illustrate that optical speckle arises in two situations: the first is when coherent light illuminates an irregular surface, with a roughness on scale of the optical wavelength; the second situation is when coherent light travels through a medium with variable refractive index [115]. In both cases, the coherence of the illuminating source is a condition for the observation of speckle. A wave's degree of coherence expresses the degree of correlation in amplitude fluctuations in space (spatial coherence) or time (temporal coherence) [116]: within the coherence time or coherence length, the phase relations between different points in the wave field are fixed. The coherence of a wave determines the ability to generate interference effects. Partial coherence refers to a situation where the observation time or area is larger than the coherence domain. By sampling multiple coherence windows in space or time, some of the interference is washed out.

When coherent light illuminates a rough surface, the returning optical wave is the result of the interference of coherent wavelets with random phase that are scattered from the microscopic structures of the surface. If the interference is constructive, a bright spot will characterize the image, whereas, if the interference is destructive, the spot will appear dark [117]. The combination of constructive and destructive interferences leads to the speckle granularity in the image. Speckle appears from a set of many coherent wavelets with random phase. Random scattering of a coherent wavefront creates this system quite readily, as we will see. Complete randomization by heterogeneity in the propagation medium requires a long interaction length, such as occurs when starlight traverses a turbulent atmosphere. In practice, random scattering of coherent waves explains most observations of speckle in ultrasound, and we will focus our

discussion on that situation. It is important to realize that, although the scattering randomizes the phase, this is a deterministic process: as long as the incident wave and the realization of the random scattering structure do not change, the observed speckle remains constant.

#### 7.2.2 Mapping optical speckle concepts to ultrasonic imaging

Optical speckle is usually observed as intensity variations due to far-field interference in a wavefront with random spatial phase disturbances. Ultrasound transducers are sensitive to the amplitude of the incident wave, and derive their contrast from temporal variations in the signal. Speckle manifests as amplitude variations due to temporal phase disturbances. These differences have to be accounted for in the application of speckle concepts from optics to ultrasonic imaging. Figure 7.3 illustrates that, both in the laser and in the acoustic case, speckle is caused by the superposition of wavelets having a random phase.

In classical laser speckle imaging, one can think of the wavelets as being positioned on a plane perpendicular to the direction of wave propagation. Invoking Huygens' principle, the resultant field on an observation screen is the sum of the propagated fields originating from wavelets within a coherence area of the incident field. For separations greater than the (transverse) coherence length of the field, the phase relations between the wavelets are not deterministic, and the interference pattern is washed out. In a ranging technique like ultrasonic imaging, the contributing wavelets are positioned at different, random, distance on the axis of propagation. The resultant amplitude now is the superposition of the waves along the time dimension, and interference results from waves separated in time by less than the coherence time of the field. In ultrasound, the coherence time is equal to the pulse duration: if the pulse spatial length is greater than the distance between two scatterers, the phase is correlated and the pulse can be considered as coherent [117]. These two realizations of speckle are mathematically equivalent, and many results obtained in the rich literature on optical speckle can be applied to ultrasound imaging.

Optical detectors are sensitive to the energy of the incident field, which is proportional to the amplitude squared, unlike ultrasound transducers, which record the incident amplitude itself [64]. This means that in optical speckle the phase is generally not available for analysis. It has to be inferred from auxiliary measurements, which are outside of the scope of this chapter; see [118] for an extensive review. The probability density functions and other statistics for the speckle field are identical for light and sound, but need to be converted into intensity statistics to describe measurements of optical speckle.

Optical coherence tomography (OCT) [119] is often introduced as the "optical analogue of ultrasonic imaging", and as such deserves special mention. In optical coherence tomography, a depth-resolved image of scattering tissue is formed by interferometry. A light source with a short coherence time is used to achieve high axial resolution, much like short pulse in ultrasound imaging. Indeed, the speckle that manifests in OCT, shown in Figure 7.4, is very

similar to that observed in ultrasound, and we have used concepts and results from the OCT literature in this chapter [120-122]. The main difference between OCT and ultrasound speckle is not in the physics, but in the beam "optics" (aperture, lenses) used to transmit and receive the sampling wave.



**Figure 7.3** The detected field is the result of interference of Huygens wavelets with random amplitude and phase within one resolution. Top: the contributing wavelets are on the line of sight at slightly different delays, as in the case of ultrasound and OCT. Bottom: the contributing wavelets are distributed on a screen perpendicular to the wave propagation direction, as in the case of laser speckle.



**Figure 7.4** OCT image of a finger tip, showing speckle both the darker epidermis and bright dermis layer. Axes in mm.

### 7.3 SPECKLE AS A 2D RANDOM WALK

The classical explanation for speckle is to conceive the sum of wavelets with random phase as a two-dimensional random walk [115]. For this discussion, we will adopt the conventional term of "phasor", which is a vector  $\alpha_n e^{i(\omega_0 t + \varphi_n)}$  that rotates at an angular frequency  $\omega_0$ , where  $\omega_0 = 2\pi f_0$  and  $f_0$  is the frequency of the incident wave. Its initial phase  $\varphi_n$  is a random variable and t is the propagation delay from its origin to the detector. The amplitude  $a_n$  is a positive random variable that reflects the scattering strength. In the simplest case, we assume that  $\varphi_n$  is uniformly distributed on [- $\pi$ , $\pi$ > and that there are sufficient contributing phasors to adequately sample the phase distribution [64]. The variation in t is only due to non-random, slowly varying geometric factors (we are not considering phase randomness due to variations in the medium, which affect t). Both  $\alpha_n$  and  $\varphi_n$  are uncorrelated.

If there are *N* scatterers within a coherence volume of the incident wave, each projecting a wavelet, the resulting complex field *S* at the detector can be written as the sum of *N* phasors:

$$S = Ae^{i\Phi} = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} a_n e^{i\varphi_n}$$

$$7.1$$

The random amplitudes  $\alpha_n = a_n / \sqrt{N}$  have been normalized for energy conservation, and the propagation delay  $e^{i\omega \rho t}$  has been suppressed. Figure 7.5 illustrates analogy between the vector sum of a few phasors and a 2D random walk in the complex plane. This representation of the field, combined with the simple assumptions on  $\varphi_n$  and  $\alpha_n$ , allows us to derive some of its statistical properties. Note that in the following derivation, we do not need any information on the distribution of  $a_n$ , which makes the resulting speckle statistics robust in homogeneous random media and independent of the details of the scattering structure.

#### 7.3.1 First order statistics

In order to compute statistics for amplitude and phase of the resultant vector, it is convenient to first decompose it into real and imaginary parts. These are given by:



**Figure 7.5** Phasor sum as a random walk. The resultant field **S** (black arrow) is the sum of many phasors (smaller gray arrows), each with random amplitude and uniformly distributed random phase.

$$S_r = Re(S) = \frac{1}{\sqrt{N}} \sum_{n=1}^N a_n \cos \varphi_n$$
7.2

$$S_i = Im(S) = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} a_n \sin\varphi_n$$
7.3

Since both the real and imaginary parts are the sum of independent random components, the central limit theorem enables us to conclude that they are, for large *N*, Gaussian random variables. The  $\overline{S}_r$  mean and  $\overline{S}_i$  are:

$$\overline{S_r} = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} \overline{a_n \cos \varphi_n}$$
7.4

$$\overline{S_{i}} = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} \overline{a_{n} \sin \varphi_{n}}$$

$$7.5$$

and, since the phases are uniformly distributed over  $[-\pi,\pi\rangle, \overline{S}_r = \overline{S}_i = 0$ . The variances  $\sigma_R^2$  and  $\sigma_I^2$  can be then calculated as the second moments  $S_R^2$  and  $S_I^2$ , which are given by:

$$\overline{S_r^2} = \frac{1}{N} \sum_{n=1}^N \sum_{k=1}^N \overline{a_n a_k} \, \overline{\cos\varphi_n \cos\varphi_k}$$
7.6

$$\overline{S_{\iota}^{2}} = \frac{1}{N} \sum_{n=1}^{N} \sum_{k=1}^{N} \overline{a_{n} a_{k}} \,\overline{\sin \varphi_{n} \sin \varphi_{k}}$$

$$7.7$$

and again due to the uniform distribution of the phases over [- $\pi$ , $\pi$ >, we have:

$$\overline{\cos\varphi_n \cos\varphi_k} = \overline{\sin\varphi_n \sin\varphi_k} = \begin{cases} 0 & k \neq n \\ \frac{1}{2} & k = n \end{cases}$$
7.8

Therefore,

$$\sigma_R^2 = \sigma_I^2 = \overline{S_r^2} = \overline{S_\iota^2} = \frac{1}{2N} \sum_{n=1}^N a_n^2 = \frac{\bar{A}^2}{2}$$
 7.9

The correlation between  $S_r$  and  $S_i$  is defined as:

$$R_{S_rS_i} = \overline{S_rS_i} = \frac{1}{N} \sum_{k=1}^{N} \sum_{n=1}^{N} \overline{a_k a_n} \ \overline{\cos \varphi_k} \overline{\sin \varphi_n}$$
7.10

Terms with  $k \neq n$  vanish because of the uncorrelated amplitudes and phases in the sum, while, since sin  $2\varphi=2 \sin \varphi \cos \varphi$ , we get:

$$\overline{\cos \varphi_k} \,\overline{\sin \varphi_n} = 0 \tag{7.11}$$

Therefore,  $S_r$  and  $S_i$  are uncorrelated.

To summarize, the real and imaginary parts of the resultant vector sum *S* are uncorrelated Gaussian random variables. They have identical variance and zero mean, as expected from the circular geometry of the problem.

To find the probability density function of the magnitude  $A=|S|=\sqrt{(S_r^2+S_i^2)}$  and phase  $\Phi$  of the sum of the phasors, we need to define the joint probability density function of the real and imaginary parts  $p_{srsi}$ . Again using the central limit theorem, and applying the first and second moments of  $S_r$  and  $S_i$  computed above, we can write for the joint probability density function

$$p_{S_r S_i}(S_r, S_i) = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{S_r^2 + S_i^2}{2\sigma^2}\right)$$
 7.12

with  $\sigma^2 = \sigma_r^2 = \sigma_i^2$ . The circular symmetry in  $p_{s_r s_{i'}}$  following from the uniform phase distribution, leads to the term "circular complex Gaussian" variate for the phasor sum *S*. From  $p_{s_r s_i}$  we can indeed compute the joint probability function of the magnitude and phase by making the appropriate variable transformations [115]. We then derive the marginal distributions of *A* and  $\Phi$ . First we define and as follows:

$$A = \sqrt{S_r^2 + S_i^2} \tag{7.13}$$

$$\Phi = \tan^{-1} \left( \frac{S_i}{S_r} \right) \tag{7.14}$$

 $S_r$  and  $S_i$  follow accordingly:

$$S_r = A\cos\Phi \qquad 7.15$$

$$S_i = A\sin\Phi \qquad 7.16$$

then we calculate the Jacobian of the transformation

$$\|J\| = \left\| \frac{\delta S_r}{\delta A} \quad \frac{\delta S_r}{\delta \Phi} \\ \frac{\delta S_i}{\delta A} \quad \frac{\delta S_i}{\delta \Phi} \\ \| = \left\| \cos \Phi \quad -A \sin \Phi \\ \sin \Phi \quad A \cos \Phi \\ \| = A \right.$$
7.17

where  $|| \dots ||$  denotes the modulus of the determinant. The joint density function for *A* and  $\Phi$  is:

$$p_{A,\Phi} = p_{S_r,S_i}(S_r = A\cos\Phi, S_i = A\sin\Phi) \cdot A$$
7.18

and if we use 7.12, we obtain

$$p_{A,\Phi} = \begin{cases} \frac{A}{2\pi\sigma^2} \exp\left(-\frac{A^2}{2\sigma^2}\right) & A \ge 0, -\pi \le \Phi < \pi \\ 0 & \text{otherwise} \end{cases}$$
 7.19

We compute the marginal density for the magnitude A by integration over  $\Phi$  to get

$$p_{A} = \begin{cases} \frac{A}{\sigma^{2}} \exp\left(-\frac{A^{2}}{2\sigma^{2}}\right) & A \ge 0\\ 0 & \text{otherwise} \end{cases}$$
 7.20

This function is known as the Rayleigh density function, displayed in Figure 7.6. Note that the probability of A=0 vanishes. Similarly, we can compute the marginal density for the phase  $\Phi$  as follows

$$p_{\Phi} = \begin{cases} \frac{1}{2\pi} \int_{0}^{\infty} \frac{A}{\sigma^{2}} \exp\left(-\frac{A^{2}}{2\sigma^{2}}\right) dA & -\pi \le \Phi < \pi \\ 0 & \text{otherwise} \end{cases}$$
 7.21



Figure 7.6 The Rayleigh distribution, describing the amplitude statistics of fully-developed speckle.

We can see that the integral argument is the Rayleigh probability density function. Therefore, the integral must be equal to 1. Thus, we obtain

$$p_{\Phi} = \begin{cases} \frac{1}{2\pi} & -\pi \le \Phi < \pi \\ 0 & \text{otherwise} \end{cases}$$
 7.22

Clearly,  $p_{A,\Phi}=p_A p_{\Phi}$ , as it should be for independent variables.

With these distributions at our disposal, we can compute the mean and variance of the scattered amplitude:

$$\bar{A} = \sqrt{\frac{\pi}{2}}\sigma$$
7.23

$$\sigma_A^2 = \left[2 - \frac{\pi}{2}\right]\sigma^2 \tag{7.24}$$

If we divide the standard deviation by the mean, we obtain the speckle contrast  $\kappa$ , the inverse of which is often loosely referred to as signal-to-noise ratio SNR [117,123].

$$\kappa = \frac{\sigma_A}{\overline{A}} = \sqrt{\frac{4-\pi}{\pi}} \approx 0.52; SNR = \frac{1}{\kappa} \approx 1.91$$
 7.25

Figure 7.7 displays the result of an imaging experiment on a tissue-mimicking phantom, scanning a single element transducer along the phantom. The experiment was repeated 15 times, each at a slightly different location, to sample a different realization of the randomness that generates the speckle pattern. Each image has an SNR close to the theoretical value of 1.91.



**Figure 7.7** Scanning probe measurement: amplitude signal of one acquisition (a), corresponding phase (b) and SNR for all 15 acquisitions (c).

For ultrasound speckle, the statistics of the amplitude is the most relevant quantity. For completeness, allowing the same approach used for the amplitude, we also compute the probability density function of the intensity  $I=A^2$ . First we relate the intensity and phase to the real and imaginary part by

0.0

$$S_r = \sqrt{I}\cos\Phi \qquad 7.26$$

$$S_i = \sqrt{I} \sin \Phi \tag{7.27}$$

The Jacobian of the transformation is:

$$\|J\| = \left\| \frac{\frac{\partial S_r}{\delta I}}{\frac{\partial S_i}{\delta I}} \frac{\frac{\partial S_r}{\delta \Phi}}{\frac{\partial S_i}{\delta \Phi}} \right\| = \frac{1}{2}$$
7.28

and the joint probability density function can be expressed as:

$$p_{I,\Phi} = p_{S_r,S_i} \left( S_r = \sqrt{I} \cos \Phi , S_i = \sqrt{I} \sin \Phi \right) \cdot \frac{1}{2}$$

$$7.29$$

By substituting 7.19 in 7.29 we get:

$$p_{I,\Phi} = \begin{cases} \frac{1}{4\pi\sigma^2} \exp\left(-\frac{l}{2\sigma^2}\right) & l \ge 0, -\pi \le \Phi < \pi \\ 0 & \text{otherwise} \end{cases}$$
 7.30

We can then integrate over the phase, obtaining

$$p_{I}(I) = \begin{cases} \frac{1}{2\sigma^{2}} \exp\left(-\frac{I}{2\sigma^{2}}\right) & I \ge 0, \\ 0 & \text{otherwise} \end{cases}$$
 7.31

which is defined as the exponential probability density function, which has its maximum value at *I*=0. The phase is obviously unaffected by the transformation and is still uniformly distributed

on  $[-\pi,\pi$ >. Note that for the intensity,  $\overline{I}=\sigma_I=2\sigma^2$  and  $\kappa=1$ , a result that seriously complicates imaging with coherent light! The physical interpretation of these results can be summarized as follows: if we have a medium that randomizes the phase of an incident wave to such a degree that the received signal can be represented as the sum of a large number of phasors, the amplitude of the complex scattered field follows a Rayleigh distribution, while the phase is uniformly distributed. The only assumptions we made are: uncorrelated phasor amplitudes and phases, uniformly distributed phases; and large *N*, which allowed us to apply the central limit theorem that straightforwardly produces the Gaussian circular statistics for the field. Speckle fulfilling these criteria is called "fully-developed", and allows us to describe the underlying physics in a concise manner. A scattered field consisting of few phasors, phasors with nonuniformly distributed phases, or correlated amplitudes, does not obey these simple statistics, as we will show later on in this chapter. Partially developed speckle is commonly encountered in ultrasound imaging. Its analysis is more complex and requires more knowledge about the scattering structure.

#### 7.3.2 Sums of speckle patterns

We mentioned in Section 7.2.1 that speckle is a deterministic random process: the signal scattered by a random structure, probed by a wave of the same frequency content and directionality, will always yield the same speckle pattern. As will become clear in the remainder of this book, many strategies to mitigate the effects of speckle rely on some form of averaging, where one control variable is changed to produce different speckle patterns that can then be summed or compounded.

Consider two random phasor sums,  $S_1 = \frac{1}{\sqrt{N}} \sum_{n=1}^N a_n e^{i\varphi_n}$  and  $S_2 = \frac{1}{\sqrt{M}} \sum_{n=1}^M b_n e^{i\vartheta_n}$ . If both satisfy the conditions for fully developed speckle,  $a_n$  and  $b_n$  are uncorrelated random variables, as are the phases  $\varphi_n$  and  $\vartheta_n$ . The sum of these two fields  $S = S_1 + S_2 = \frac{1}{\sqrt{N}} \sum_{n=1}^N a_n e^{i\varphi_n} + \frac{1}{\sqrt{M}} \sum_{n=1}^M b_n e^{i\vartheta_n}$  can be written as  $S = \frac{1}{\sqrt{N+M}} \sum_{n=1}^{N+M} c_n e^{i\gamma_n}$ , where  $c=a \cup b$  and  $\gamma = \varphi \cup \vartheta$ . This again has the form of a random phasor sum: it represents a longer random walk. Addition of speckle patterns on a complex amplitude basis does *not* alter the statistics.

If we compute, on the other hand, the incoherent sum of the resultant field, based on the signal envelopes,  $A_n = |S_n|$  the distribution does change. The sum of N speckle fields is:

$$A_S = \sum_{n=1}^{N} A_n$$
 7.32

Before going into mathematical detail, the net gain in SNR can be estimated from general results in probability theory. A sum  $A_s$  of N real-valued, identically distributed, uncorrelated random variates, each with a mean  $\overline{A}$  and standard deviation  $\sigma_A$ , has an expected value  $\overline{A}_s = N\overline{A}$  and a standard deviation  $\sigma_{AS} = \sqrt{N\sigma_A}$ . The SNR of the summed fields thus scales as  $1/\sqrt{N}$ ; a well-known and commonly applied result.

If  $\alpha$  and  $\beta$  are two random variables with probability density functions  $p_{\alpha}(\alpha)$  and  $p_{\beta}(\beta)$ , the probability distribution of the sum  $\alpha+\beta$  is given by the convolution of their individual distributions,  $p_{\alpha+\beta}(\alpha+\beta)=p_{\alpha}(\alpha)\otimes p_{\beta}(\beta)$ . Hence, for a sum N=2 of fully-developed speckles with identical magnitude (same  $\sigma$ ), an exact result for the probability distribution of the resulting magnitude can be found from the convolution of a Rayleigh distribution with itself:

$$p_{S,N=2}(A_s) = \frac{A_s}{2\sigma^2} e^{-\frac{A_s^2}{2\sigma^2}} + \frac{\sqrt{\pi}}{4\sigma} e^{-\frac{A_s^2}{4\sigma^2}} \left(\frac{A_s^2}{\sigma^2} - 2\right) \operatorname{erf}\left(\frac{A_s}{2\sigma}\right)$$
(7.33)

where  $\operatorname{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$  is the error function, and the integration over the phase has been implicitly performed.

The distribution of the sum  $A_s$  can be computed for general N, in principle, using the characteristic function. The characteristic function  $M_x(\omega)$  of a random variable x is the expectation value of  $e^{i\omega x}$ , or equivalently, the inverse Fourier transform of its probability density function. It can be found using symbolic manipulation software:

$$M_{A}(\omega) = \int_{0}^{\infty} e^{i\omega A} p_{A}(A) dA = \frac{1}{2\pi} \int_{0}^{\infty} e^{i\omega A} \frac{A}{\sigma^{2}} e^{-\frac{A^{2}}{2\sigma^{2}}} dA$$

$$= 1 - \sqrt{2}\sigma\omega D_{+} \left(\frac{\sigma\omega}{\sqrt{2}}\right) + i\sigma\omega \sqrt{\frac{\pi}{2}} e^{-\frac{\sigma^{2}\omega^{2}}{2}}.$$
7.34

Here,  $D_+$  is the function known as Dawson's integral, defined as  $D_+(x) = e^{-x^2} \int_0^x e^{t^2} dt$ . The characteristic function of  $A_s$  then is:

$$M_{A,S} = \prod_{n=1}^{N} M_{A,n} = [M_A(\omega)]^N.$$
7.35

The last equality holds if we assume that all contributing fields have the same mean amplitude (and thus, the same  $\sigma_n = \sigma$ ). We can write the probability distribution of  $A_s$  as the Fourier transform of  $M_s$ . It turns out, however, that for N>2, the analytic form of  $p_{S,N}$  cannot be straightforwardly computed, nor is it known in literature. The moments of the distribution can be found by serial differentiation of  $M_s$  (from which the probability distribution function itself can be approximated in principle by construction of a moments series). Again, by the central limit theorem, the distribution of magnitudes approaches a Gaussian for  $N \rightarrow \infty$ .

For some applications [124], the statistics of the signal power in a sum of speckles is of interest. The probability distribution function of a sum of intensities:

$$I_S = \sum_{n=1}^{N} I_n = \sum_{n=1}^{N} |A_n|^2$$
 7.36

can be computed using the characteristic function of the exponentially distributed intensity [118]:

$$M_I(\omega) = \frac{1}{1 - i\omega\bar{I}}$$
7.37

where  $\bar{I}=2\sigma^2$  is the average intensity as in defined in Section 7.3.1. As above,  $M_{I,S} = \prod_{n=1}^{N} M_{I,n} = [M_I(\omega)]^N$ , for an ensemble of speckles with equal mean intensity, which can be Fourier transformed to yield:

$$p_{S}(I_{S}) = \frac{I_{S}^{N-1}}{\Gamma(N)\bar{I}^{N}} e^{-I_{S}/\bar{I}}$$
7.38

where  $\Gamma(N)$  is the gamma function and  $\overline{I}$  is the mean intensity of each contributing field. For large *N*, the distribution again tends towards a Gaussian.

#### 7.3.3 Speckle pattern plus a non-random phasor

The preceding sections discussed the case of fully developed speckle, arising when many wavelets with random phase interfere. In practical imaging situations, scattering media are often heterogeneous, containing strongly reflecting boundaries that can be resolved amidst a speckle generate by unresolved, microscopic scattering. The reflections generate a constant background field. The scattered field can now be described as the sum of a random walk and the constant, known phasor. The approach here follows [118]. Aligning the constant phasor  $A_0$  with the real axis, we can add it to the real component of the field in 7.12 to yield

$$p_{S_r S_i}(S_r, S_i) = \frac{1}{2\pi\sigma^2} \exp\left(-\frac{(S_r + A_0)^2 + S_i^2}{2\sigma^2}\right)$$
7.39

We again perform the transformation of variables 7.15–7.17 and find:

$$p_{A,\Phi}(A,\Phi) = \begin{cases} \frac{A}{2\pi\sigma^2} \exp\left(-\frac{A^2 + A_0^2 - 2AA_0\cos\Phi}{2\sigma^2}\right) & A > 0, -\pi < \Phi \le \pi \\ 0 & \text{otherwise} \end{cases}$$
7.40

The probability distribution of the magnitude of the resultant phasor can be obtained by integration of above equation with respect to the phase  $\Phi$ . Therefore,

$$p_{A}(A) = \begin{cases} \frac{A}{2\pi\sigma^{2}} e^{-\frac{A_{0}^{2}+A^{2}}{2\sigma^{2}}} \int_{-\pi}^{\pi} e^{-\frac{AA_{0}\cos\Phi}{\sigma^{2}}} d\Phi = \frac{A}{\sigma^{2}} e^{-\frac{A_{0}^{2}+A^{2}}{2\sigma^{2}}} I_{0}\left(\frac{A_{0}A}{\sigma^{2}}\right) \quad A > 0 \\ 0 & \text{otherwise} \end{cases}$$
7.41

where  $I_0$  is the modified Bessel function of the first kind and order zero. The above expression is known as the Rician density function; Figure 7.8 shows the distribution for a number of different background amplitudes  $k=A_0/\sigma$ . It reduces to the Rayleigh distribution for  $A_0=0$ , as expected. The mean and the variance can be calculated from the first two moments of the density function. Thus:

$$\bar{A} = \int_{0}^{\infty} \frac{A}{\sigma^{2}} e^{-\frac{A_{0}^{2} + A^{2}}{2\sigma^{2}}} I_{0}\left(\frac{A_{0}A}{\sigma^{2}}\right) dA$$

$$= \sqrt{\frac{\pi}{2}} \sigma e^{-\frac{k^{2}}{4}} \left[ \left(1 + \frac{k^{2}}{2}\right) I_{0}\left(\frac{k^{2}}{4}\right) + \frac{k^{2}}{2} I_{1}\left(\frac{k^{2}}{4}\right) \right]$$

$$\bar{A}^{2} = \sigma^{2} [2 + k^{2}]$$
7.43

where  $I_1$  is the modified Bessel function of the first kind and order one.

If k is large, then the density function of the magnitude approximates the Gaussian density function

$$p_A(A) \cong \frac{1}{\sqrt{2\pi\sigma}} e^{-\frac{(A-A_0)^2}{2\sigma^2}}$$
 7.44

with  $\overline{A} \cong \overline{A}_0$  and  $\overline{A^2} \cong \sigma^2$ .





To find the density function of the phase, we integrate with respect to the magnitude and we obtain

$$p_{\Phi}(\Phi) = \frac{e^{-\frac{k^2}{2}}}{2\pi} + \frac{k\cos\Phi}{\sqrt{2\pi}}e^{-\frac{k^2\sin^2\Phi}{2}}\left(\frac{1+\operatorname{erf}\left(\frac{k\cos\Phi}{\sqrt{2}}\right)}{2}\right)$$
 7.45

for  $-\pi < \Phi \le \pi$ .

Note that for k=0 the distribution is uniform, while for  $k\to\infty$  the function converges to the delta function at  $\Phi=0$ . If k is large, the density function of the phase approximates a Gaussian function [118]

$$p_{\Phi}(\Phi) \cong \frac{k}{\sqrt{2\pi}} e^{-\frac{k^2 \Phi^2}{2}}$$
7.46

with  $\overline{\Phi} \cong 0$  and  $\Phi^2 \cong \frac{1}{k^2}$  the first and second moment, respectively. The variance equals the second moment because  $\overline{\Phi} \cong 0$ .

#### 7.3.4 Non-uniform phase distribution

Like the constant background phasor presented in the previous sections, deviations from uniform, uncorrelated, isotropic randomness in the scattering structure can give rise to nonuniformly distributed phases of the component phasors; one example in ultrasonic imaging is long-range correlation of scatterer position. If we relax the assumption of uniform phase distribution, there is no general closed from for the amplitude distribution, but maintaining the statistical independence of amplitudes and phases, we can still compute some properties of the scattered field (see [125]; sect 2.9 and Appendix B). We start from equations 7.4 and 7.5:

$$\overline{S}_r = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} \overline{\alpha_n} \, \overline{\cos \varphi_n}$$
7.47

$$\overline{S}_{i} = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} \overline{a_{n}} \, \overline{\sin \varphi_{n}}$$
7.48

and recognize that the mean trigonometric functions can be expressed in terms of the characteristic function of the phase:

$$\overline{\cos\varphi_n} = \frac{1}{2}(\overline{e^{\iota\varphi}} + \overline{e^{-\iota\varphi}}) = \frac{1}{2}[M_{\varphi}(1) + M_{\varphi}(-1)]$$
7.49

$$\overline{\sin \varphi_n} = \frac{1}{2i} (\overline{e^{i\varphi}} + \overline{e^{-i\varphi}}) = \frac{1}{2i} [M_{\varphi}(1) - M_{\varphi}(-1)]$$

$$7.50$$

With this result, we can express the mean real and imaginary components of the resultant field as

$$\overline{S}_r = \frac{\sqrt{N}\overline{a}}{2} \left[ M_{\varphi}(1) + M_{\varphi}(-1) \right]$$
7.51

$$\overline{S}_{\iota} = \frac{\sqrt{N}\overline{a}}{2i} \left[ M_{\varphi}(1) - M_{\varphi}(-1) \right]$$

$$7.52$$

Again using Euler's formulas, the variances and covariance can be written in similar terms:

$$\sigma_r^2 = \overline{S_r^2} - \overline{S_r}^2 = \frac{\overline{a^2}}{4} [2 + M_{\varphi}(2) + M_{\varphi}(-2)] - \frac{\overline{a^2}}{4} [2M_{\varphi}(1)M_{\varphi}(-1) + M_{\varphi}^2(1) + M_{\varphi}^2(-1)]$$
7.53

$$\sigma_{\iota}^{2} = \overline{S_{\iota}^{2}} - \overline{S_{\iota}}^{2} = \frac{\overline{a^{2}}}{4} [2 - M_{\varphi}(2) - M_{\varphi}(-2)]$$

$$- \frac{\overline{a}^{2}}{4} [2M_{\varphi}(1)M_{\varphi}(-1) - M_{\varphi}^{2}(1) - M_{\varphi}^{2}(-1)]$$

$$C_{ri} = \overline{S_{r}S_{\iota}} - \overline{S_{r}}\overline{S_{\iota}} = \frac{\overline{a^{2}}}{4i} [M_{\varphi}(2) - M_{\varphi}(-2)] - \frac{\overline{a}^{2}}{4i} [M_{\varphi}^{2}(1) - M_{\varphi}^{2}(-1)]$$

$$7.54$$

With a few extra assumptions, simpler expressions can be computed. First, in the important special case that  $p_{\varphi}(\varphi)$  is symmetric around  $\varphi = 0$  (and any probability distribution that is symmetric around a central value can be shifted to the origin by a rotation of the coordinate system), the characteristic function is even and real. As a result,

$$\overline{S_r} = \sqrt{N}\overline{a}M_{\varphi}(1)$$

$$\overline{S_i} = 0$$

$$\sigma_r^2 = \frac{1}{2}\overline{a^2} \left[1 + M_{\varphi}(2)\right] - \overline{a}^2 M_{\varphi}^2(1)$$

$$\sigma_i^2 = \frac{1}{2}\overline{a^2} \left[1 - M_{\varphi}(2)\right]$$

$$C_{ri} = 0$$

$$7.56$$

Inspection of these equations shows that the mean of the real part of the resultant field is now non-zero, and the variances of real and imaginary parts differ. Hence, the scattered field is no longer a circular complex random variable.

More specifically, if the phase is a zero-mean Gaussian variable with a probability density function  $p_{\varphi}(\varphi) = \frac{1}{\sqrt{2\pi}\sigma_{\varphi}} \exp\left(-\frac{\varphi^2}{2\sigma_{\varphi}^2}\right)$ , the characteristic function is

$$M_{\varphi}(\omega) = \exp\left(-\frac{\sigma_{\varphi}^2 \omega^2}{2}\right)$$
 7.57

Substitution in the relations above yields

$$\overline{S_r} = \sqrt{N}\overline{a}e^{-\frac{\sigma_{\varphi}^2}{2}}$$

$$\overline{S_l} = 0$$

$$\sigma_r^2 = \frac{1}{2}\overline{a^2} \left[1 + e^{-2\sigma_{\varphi}^2}\right] - \overline{a}^2 e^{-\sigma_{\varphi}^2}$$

$$\sigma_l^2 = \frac{1}{2}\overline{a^2} \left[1 - e^{-2\sigma_{\varphi}^2}\right]$$

$$C_{ri} = 0$$

$$7.58$$

For  $\sigma_{\varphi}^2 \rightarrow \infty$ , the probability density function of the phase approaches a uniform distribution, and we retrieve the familiar means and variances of section 2.3.1.

## 7.4 SPECKLE IN ULTRASONIC IMAGING

In ultrasound imaging, the signal arises from backscattered waves inside the imaged object. Acoustic heterogeneity gives rise to scattering structures, which in biological tissue span a large range of characteristic length scales; micrometers at the subcellular level to centimeters or decimeters at the organ level. At length scales smaller than the acoustic wavelength, biological tissue can, to a degree, be modeled as a random acoustic medium with a spectrum of correlation lengths, and scattering coefficients that depend on tissue type. Larger structures reflect ultrasound, rather than scatter it, and present as readily recognizable structures (see, for instance, the wire reflections in Figure 7.1(a). As these large structures do not exhibit randomness on the scale of a wavelength, they do not generate speckle.

In the treatment of speckle as an ultrasound scattering phenomenon, we will implicitly depend on the Born approximation, which states that the scattered field is small in comparison to the incident beam. This allows us to neglect multiple scattering and attenuation due to scattering, both reasonably accurate assumptions in most ultrasound imaging applications. Acoustic absorption of course exponentially attenuates the field amplitude, but is not essential for understanding ultrasound speckle, since path length differences contributing to the random interference are small in the single scatter approximation.

#### 7.4.1 Random ultrasound scattering in 1D

The phase randomness that is evident in the Rayleigh-distributed signal distribution arises because of the presence of many scatterers within a resolution cell, the scattered fields of which are added coherently to result in a random local phase. The random phasor sum 7.1 appears straightforwardly if we simplify the ultrasound scattering treatment to a one-dimensional linear systems description. In this framework, the field emitted by a scattering potential  $\Psi$  that is insonified with a finite-length pulse can be written as the convolution

$$S = \Psi \otimes h \tag{7.59}$$

Here *h* is the axial point spread function as determined by the temporal impulse response of the band-limited system, which we write as  $h(t)=H(t)e^{i2\pi f_0 t}$ , the product of a carrier wave of frequency  $f_0$  and the envelope H(t) that is the Fourier transform of the power spectral density of the pulse. For a single point-like scatterer located at position  $z_1$ ,  $\Psi=\Psi_1\delta(z-z_1)$ , and so

$$S_1(t) = \Psi_1 h(t - z_1/c) = \Psi_1 H(t - z_1/c) e^{i2\pi f_0 t} e^{-i2\pi f_0 z_1/c}$$
7.60

This expression shows that the amplitude response is limited in time by the extent of *H*, and that it incurs a phase  $2\pi f_0 z_1/c$ . Figure 7.9(a) shows a graph of the scattered field in response to a two-cycle Gaussian envelope excitation. If we now complicate the scattering structure by adding a few more point-like potentials at random positions,  $\Psi = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} \Psi_n \delta(z - z_n)$ , then

$$S_N(t) = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} \Psi_n H(t - z_n/c) e^{i2\pi f_0 t} e^{-i2\pi f_0 z_n/c} = \frac{1}{\sqrt{N}} \sum_{n=1}^{N} a_n e^{i\varphi_n}$$
 7.61

The  $a_n = \Psi_n H(t-z_n/c)$  form a set of random amplitudes at any given time , and the phase  $\varphi_n = 2\pi f_0 z_n/c$ . This expression has the exact form of 7.1 if the positions  $z_n$  are random. It follows that the scattered field from a random scattering potential is a speckle: it is a random zero-mean circular Gaussian variable. Its amplitude is a random variable that follows a Rayleigh distribution, and the phase is uniformly distributed on  $|-\pi,\pi\rangle$ , the random phase being a consequence of the random scatterer position. The one caveat is that the density of scatterers *per resolution cell* needs to be large enough for the Rayleigh distribution to emerge, as illustrated by Figure 7.9(c) (d). Sections of *S* that are separated in time by more than the width of H(t) are uncoupled and have independent statistics. This argument can be expanded to two and three dimensions, as we will do to compute speckle spot size.

Returning to the linear systems formulation of scattering and Fourier transforming it (indicated by hats;  $F(\cdot) = \uparrow$ ) yields a product in spatial frequency space

$$\widehat{S}(k) = \widehat{\Psi}(k) \cdot \widehat{h}(k)$$
7.62

The axial width of the point spread function is a consequence of the band limitation of the ultrasound imaging system, which appears here directly as  $k_z = \frac{f}{c}$ . For frequencies outside the system bandwidth  $f_0 \pm \Delta f/2$ , the response of the system vanishes. This relation shows that the recorded scattered field *S* only contains spatial frequencies within the band of  $\hat{h}$ . By the Wiener-Khinchine theorem, the frequency spectrum of *S* is the Fourier transform of the autocorrelation function  $R_S = \langle S(z_1)S(z_2) \rangle = R_S(\Delta z)$ . The scattered field only samples spatial frequencies of the scattering structure in a passband centered  $2/\lambda$ , with a bandwidth of the order of  $2\Delta f/_{c'}$  depending on the shape of the envelope of *h* [126,127].

Tissues contain a range of randomly arranged scattering structures, ranging in size from vesicles and cells (large spatial frequency) to blood vessels (small spatial frequency). A range of tissue acoustic models exists that predict the frequency dependence of the scattered field, which is to some degree tissue specific. The equivalence of the linear scattering formulation in real space and k-space shows that the randomness of speckle obscuring features of the imaged structure, and the limited sampling of spatial frequency space, are in fact two sides of the same coin.



**Figure 7.9** Response to a 2-cycle sine wave with a Gaussian envelope, simulated in a linear system as in Eq. (60). (A) Single scatterer (indicated by the vertical line. (B) Three randomly positioned scatterers within one resolution cell affect the amplitude and phase; note the shift of the trough over the centre scatterer. (C) Many scatterers (N=12 per resolution cell) exhibit speckle. (D) Increasing the number of scatterers and the length of the simulated trace yields a Rayleigh distribution of the envelope (simulation parameters are arbitrary).

#### 7.4.2 Second order speckle statistics in ultrasound

The second-order statistics enables to describe the statistical relation between two different pixels in the image. Among many other speckle features, they allow the computation of the speckle spot size, which is an important parameter in determining the degree to which speckle affects the image. To derive the autocorrelation and autocovariance of the complex ultrasound field *S*, we again adopt the linear systems description of ultrasound scattering [123]. Therefore,

$$S = \int_{-\infty}^{+\infty} \Psi(\mathbf{x} - \mathbf{x}') h(\mathbf{x}') d\mathbf{x}' = \Psi(\mathbf{x}) \otimes h(\mathbf{x})$$
 7.63

where  $\Psi$  is the complex scattering amplitude and *h* is the point spread function. The autocorrelation function  $R_s$  is:

$$R_s = E[S(\mathbf{x}_1)S^*(\mathbf{x}_2)] = R_{\Psi}(\mathbf{x}_1, \mathbf{x}_2) \otimes h(\mathbf{x}_1) \otimes h^*(\mathbf{x}_2)$$
7.64

Where  $R_{\Psi}(\mathbf{x}_1, \mathbf{x}_2) = E[\Psi(\mathbf{x}_1)\Psi^*(\mathbf{x}_2)]$  and  $E[\cdot]$  denotes the expectation operator. Since the real and imaginary parts of *S* are uncorrelated Gaussian random variables with zero mean,  $C_s = R_s$ , where  $C_s$  is the autocovariance of the speckle field.

As before, we assume weak interaction between the medium and the ultrasound field, discarding any multiple scattering. Furthermore, the coarse macroscopic scattering strength is slowly varying compared to the correlation of the microstructure. This diagonalizes the scattering amplitude autocovariance  $R_{\Psi}$  [123]:

$$R_{\Psi}(\mathbf{x_1}, \mathbf{x_2}) = |\Psi(\mathbf{x})|^2$$
 7.65

with  $x_1 = x \cong x_2$  and  $\Delta x = x_2 - x_1$ .

If the imaging target is a homogenous distribution of scatterers, or, more generally, the macroscopic scattering structure is slowly varying compared to the width of the point spread function, we can write it as a constant locally:  $|\Psi(\mathbf{x})|^2 = \Psi_0^2$ . This allows us to rewrite

$$C_{S}(\Delta \mathbf{x}) = \Psi_{0}^{2}h(-\Delta \mathbf{x}) \otimes h^{*}(\Delta \mathbf{x})$$
7.66

From 7.66, we can conclude that, when imaging objects composed of randomly dispersed fine particles, speckle contains information only about the point spread function.

Assuming that we are in the focal region of the ultrasound transducer, the point spread function can be separated along two orthogonal directions: the insonification direction z (longitudinal) and the transverse direction x (lateral) [128]

$$h(x,z) = h_x h_z 7.67$$

and the covariance matrix can be written as follows:

$$C_{S}(\Delta x, \Delta z) = C_{Sx}(\Delta x)C_{Sz}(\Delta z)$$
7.68

Therefore, it is possible to analyze the resulting two covariance functions separately.

For a rectangular transducer the transverse point spread function in the focal plane is defined as:

$$h(x) = B \operatorname{sinc}^{2}(\pi x \xi_{0})$$

$$\xi_{0} = D/\lambda z_{0}$$
7.69

with *B* a normalization factor, *D* the width of the transducer,  $\lambda$  the wavelength and  $z_0$  the focal zone distance from the transducer. The expression in 7.69 is obtained assuming continuous wave approximation, which can be used as a first order estimate of the spatial response in the focal zone.

The autocovariance matrix in the transverse direction is then:

$$C_{S_x}(\Delta x) = K_x sinc^2(\pi \Delta x \xi_0) \otimes sinc^2(\pi \Delta x \xi_0)$$

$$= [K_x/(\pi \Delta x)^2] [1 - sinc(2\pi \Delta x \xi_0)]$$
7.70

with  $K_x$  a normalization factor.

Regarding the autocovariance along the longitudinal direction, we should first define the longitudinal point spread function. If, for simplicity, we assume the shape of the pulse envelope to be Gaussian ~  $(0, \sigma_z^2)$ , then

$$h(z) = \frac{1}{2\pi\sigma_z^2} \exp(-\frac{z^2}{2\sigma_z^2})$$
7.71

and the autocovariance becomes

$$C_{SZ}(\Delta z) = K_z \exp\left[-\frac{\Delta z^2}{4\sigma_z^2}\right]$$
7.72

with  $K_z$  a normalization factor.

With the knowledge about the autocovariance matrix in the transverse and longitudinal direction, we can define the speckle spot size [123,128]. In fact, the speckle size along the transverse direction,  $\delta_x$ , and along the longitudinal direction  $\delta_z$ , are defined as follows

$$\delta_x = \int_{-\infty}^{+\infty} C_{S_x}(\Delta x) / C_x(0) d(\Delta x)$$
7.73

$$\delta_z = \int_{-\infty}^{+\infty} C_{S_z}(\Delta z) / C_z(0) d(\Delta z)$$
7.74

For a rectangular transducer the lateral and axial correlation cell size are [59]

$$\delta_x = \frac{0.87}{\xi_0} \tag{7.75}$$

$$\delta_z = 0.91 \frac{c}{\Delta f}$$
 7.76

with *c* the speed of sound and  $\Delta f$  the frequency bandwidth. From 7.75 and 7.76 we can see that the speckle size in the transverse direction is proportional to the beam width, while in the longitudinal direction is inversely proportional to the bandwidth. That means that the speckle size is comparable to the resolution cell size.

The point spread function is not spatially invariant. Far from the focal region, the factorization in 7.67 does not hold, which complicates the detailed mathematical description of the speckle correlation, but not the approximate equality of speckle spot and resolution cell sizes as a result of 7.66.

#### 7.4.3 Partially developed speckle and speckle from few scatterers

Thus far, we assumed that the density of microscopic scattering events was sufficiently high that the phase of the signal was randomized within one resolution cell. In practical ultrasonic imaging applications, this condition is often not met: the speckle is "partially developed." The assumptions of the central limit theorem are not fulfilled any longer, and the real and imaginary parts do not have a Gaussian distribution anymore. Different statistical models have to be applied to describe the signal [129]. One circumstance in which this occurs was discussed in the previous section: presence of strong, resolved reflectors or correlated scattering structures giving rise to a constant phasor background.

We briefly discuss here a few more special cases for which the speckle statistics can be computed.

If we have a finite number of random scatteres, we can consider the problem as an *N*-dimensional random walk problem, with *N* a finite number. Assume again that the *N* phasors have independent phase and amplitude and that the phase is uniformly distributed. Recalling 7.1-7.3, and using the same definitions as before, we can write the complex field *S* as

$$S = \sum_{n=1}^{N} \alpha_n e^{i\varphi_n} = S_r + iS_i = Ae^{i\Phi}$$
 7.77

The joint characteristic function of  $S_r$  and  $S_i$  is

$$M_{S_{r},S_{i}}(\omega_{1},\omega_{2},N) = E\left[e^{i(\omega_{1}S_{r}+\omega_{2}S_{i})}\right] = \int_{0}^{\infty} \int_{-\pi}^{\pi} e^{i(\omega_{1}S_{r}+\omega_{2}S_{i})} p_{\alpha_{n},\varphi_{n}} d\varphi_{n} d\alpha_{n}$$
 7.78

where  $p_{\alpha_n,\varphi_n}$  is the joint probability distribution of the amplitude and the phase of the individual scatterer; since  $S_r$  and  $S_i$  depend on  $\alpha_n$  and  $\varphi_n$  only, the integration over S can be replaced by the phasor variables directly. Note that the characteristic function is a function of the number of scatterers N.

If we define the polar coordinates in the plane ( $\omega_1 \omega_2$ ) through  $\omega_1 = \Omega \cos \chi$  and  $\omega_2 = \Omega \sin \chi$ , we can rewrite 7.78 as

$$M_{S_r,S_i}(\omega_1\omega_2,N) = \int_0^\infty \prod_{n=1}^N J_0(\alpha_n\Omega) p_{\alpha_n} d\alpha_n$$
7.79

with  $J_0$  the Bessel function of the first kind and order zero. Knowing that the scatterer amplitudes are independent and identically distributed, we obtain

$$M_{S_r,S_i}(\omega_1\omega_2,N) = \left[\int_0^\infty J_0(\alpha\Omega)p_\alpha d\alpha\right]^N$$
7.80

From 7.80 we can note that, in order to compute the distributions of the phase and amplitude of the complex field, we need to define the probability density function  $p_{\alpha}$  of the scatterer amplitude  $\alpha$ . This dependence arises because for small N we cannot rely on the central limit theorem to produce Gaussian statistics. The interesting consequence of this complication is that for non-Rayleigh-distributed speckle, we can infer properties of the scattering medium from the speckle distribution. It is possible to demonstrate that the K distribution is a suitable model for  $p_{\alpha}$  in many media. It is defined as follows

$$p_{\alpha} = \frac{2b}{\Gamma(\nu+1)} \left(\frac{b\alpha}{2}\right)^{\nu+1} K_{\nu}(b\alpha) \quad \text{with } \alpha \ge 0, \nu \ge -1$$

$$b = 2\sqrt{\frac{N(\nu+1)}{E[\alpha^{2}]}}$$
7.81

where  $K_v(\cdot)$  is the modified Bessel function of the second kind and order v and  $\Gamma(\cdot)$  is the Gamma function. By computing the characteristic function  $M_{S_r,S_i}$  ( $\omega_1 \ \omega_2 N$ ) using the K distributed  $p_{\alpha}$  and inverting to get the joint density function of the envelope A and phase  $\Phi$ , we obtain

$$p_{A,\Phi} = \frac{1}{2\pi} \frac{2b}{\Gamma(Q)} \left(\frac{bA}{2}\right)^Q K_{Q-1}(b\alpha) \text{ for } A > 0$$

$$7.82$$

where we have substituted Q=N(1+v), and so  $b = 2\sqrt{\frac{Q}{E[A^2]}}$ . Computing the marginal densities, we obtain

$$p_A = \frac{2b}{\Gamma(Q)} \left(\frac{bA}{2}\right)^Q K_{Q-1}(bA) \qquad \text{for } A > 0 \qquad 7.83$$

$$p_{\pi} = \frac{1}{2\pi} \qquad -\pi \le 0 \le \pi \qquad 7.84$$

Analysis of the parameters of the K distribution can provide information on the number of scatterers and on the scattering section. This is unlike the statistics of fully developed speckle, which only depends on the mean amplitude. The parameter v represents the skewness of the distribution. For v approaching infinity,  $p_{\alpha}$  approaches a Rayleigh distribution while for v approaching -1,  $p_{\alpha}$  approaches a lognormal distribution. Therefore, we can appreciate the generality of the K distribution since, it is possible to change the distribution of the scatterer amplitude by changing v. Knowing the meaning of the parameter v, it is possible to understand the physical significance of Q. For v approaching -1, Q will be very small even if N is very large. Therefore, Q represents the effective number of scatterers per resolution cell. Thus Q is a measure of both the number of scatterers and the variation in scattering cross-section [129]. Figure 7.10 shows plots of the K-distribution for different effective numbers of scatterers Q, demonstrating that for Q>10 the difference with the Rayleigh distribution becomes small.



Figure 7.10 K-distribution for different effective number of scatterers .

# **7.5** EFFECT OF POST-PROCESSING ON FIRST AND SECOND ORDER STATISTICS

In standard B-mode imaging, the ultrasound signal detected by the transducer is usually processed non- linearly in order to compress the incoming signal having wide range into a smaller signal with reduced range. This is done to overcome the problem of displaying a wide range signal using too many gray-scale levels that would not be distinguished by the observer [64]. Therefore, compression is needed in order to be able to display the images on the

commercially available systems. This non-linear postprocessing of the incoming signal will affect its statistics, leading to changes in the probability density functions. It is therefore important to study the effect of these compression mechanisms on the previously derived statistics. The most common way of compressing ultrasound data is to use a logarithmic amplifier. More in detail, the received signal amplitude *A* will be scaled and shifted according to the following expression:

$$A' = c \ln A + d \tag{7.85}$$

where A' is the post processing amplitude and c and d are constants associated to the dynamic range and the gain setting respectively [130]. If  $A_{min}$  and  $A_{max}$  are the minimum and maximum respectively of the input A which lead to  $A'_{min}$  and  $A'_{max}$  and of the output A', then

$$c = \frac{A'_{max} - A'_{min}}{\ln(A_{max} - A_{min})}$$
7.86

Knowing that the dynamic range is given by

$$DR = 20 \log \left(\frac{A_{max}}{A_{min}}\right)$$
 7.87

We can rewrite 7.86 such that

$$c = \frac{20}{DR\ln(10)} (A'_{max} - A'_{min})$$
7.88

Therefore, knowing the input and output dynamic range, it is possible to estimate the logarithmic amplifier parameter *c*.

To compute the probability density function of we can use the following expression [131]

$$p(A') = p(A) \left(\frac{dA'}{dA}\right)^{-1}$$
 7.89

with the probability density function of A given by 7.20. We then get

$$\frac{dA}{dA} = \frac{c}{A} = \frac{c}{e^{\left(\frac{A'-d}{c}\right)}}$$
7.90

And p(A') becomes

$$p(A') = \frac{1}{\beta} e^{(-g - e^{-g})}$$
7.91

with

$$g = \frac{\alpha - A'}{\beta} \qquad \alpha = \frac{c \log(2\sigma^2)}{2} + d \qquad \beta = \frac{c}{2}$$
 7.92

Equation 7.91 is a double exponential (Fisher-Tippet) [130]. Note that  $\alpha$ ,c, and d are functions of the standard deviation  $\sigma$  of the Rayleigh distribution. Therefore, we can compute the mean and variance of A' as

$$E[A'] = c\left(\frac{\ln 2}{2} + \ln \sigma - \frac{\gamma}{2}\right) + d$$
 7.93

$$E[(A' - E[A'])^2] = \frac{\pi^2 c^2}{24}$$
7.94

where  $\gamma$  is the Euler constant ( $\gamma \approx 0.5772$ ). Figure 7.11 shows the effect of log-compression on the Rayleigh distribution.

Knowing the variance of the log compressed data we can derive the values of the parameters c. Assuming the known variance is equal to v, we get:

$$c = \frac{2\sqrt{6}}{\pi}\sqrt{\nu}$$
 7.95

This means that, for fully developed speckle, we can derive the amplification parameter from the signal statistics (the variance).



**Figure 7.11** Rayleigh distribution and log-compressed distribution with c=2 and d=4.

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# Chapter 8

# Structured Ultrasound Microscopy

Based on:

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# ABSTRACT

We present a new form of acoustic microscopy, called Structured Ultrasound Microscopy (SUM). It creates a volumetric image by recording reflected echoes of ultrasound waves with a structured phase front using a moving single-element transducer and solving an inverse scattering problem. A priori knowledge of the acoustic field produced by the single element allows to relate the received echoes to a 3D scatter map within the acoustic beam itself, leading to an isotropic resolution at all depths. An aberration mask in front of the acoustic element imposes the phase structure, broadening the beam and breaking the spatial coherence between different voxels at equal acoustic propagation delay, increasing the uniqueness of the scattering problem. By translating the transducer across the 3D volume, we synthetically enlarge the imaging aperture by using multiple overlapping and spatially sparsely sampled measurements to solve for the entire image. In this paper we explain the SUM technique and successfully demonstrate microscopic imaging at 20 MHz of a 2.3×2.3×1.2 mm object in water, with an isotropic resolution below 100 µm. The proposed approach allows for wide-field 3D imaging at isotropic microscopic resolution using a very small unfocused ultrasound sensor and multiple spatially sparsely sampled measurements. This technique may find applications in many other fields where space is constrained, device simplicity is desired, and wide-field isotropic high-resolution imaging is required.

## 8.1 INTRODUCTION

High-resolution wide-field ultrasound microscopy is of great interest for different applications ranging from non-destructive testing of small electronic components [132] to tissue microscopy [133-136] and intravascular imaging [137-139]. Scanning acoustic microscopy (SAM) enables high resolution imaging using a high frequency, tightly-focused ultrasound transducer. Image data is acquired by scanning the transducer or the object under study in two dimensions while acquiring pulse-echo A-lines [140]. The imaged volume is obtained by mapping the backscattered echoes to a 3D image grid. The lateral resolution achieved at the focal distance; the axial resolution is proportional to the acoustic bandwidth. The depth-varying, anisotropic resolution, the large number of sampling points needed for full 3D coverage and the complexity of manufacturing high-frequency focused transducers limit the utility of SAM in many practical applications. In this work, we propose Structured Ultrasound Microscopy (SUM), a wide-field imaging technique with isotropic high resolution. SUM exploits the spatial a priori information of a structured ultrasound beam and achieves volumetric imaging with a set of measurements spatially sampled on a sparse grid.

Computational imaging has attracted broad interest recently, as it has enabled simultaneous wide field-of-view and high image resolution, often realized in compact, simple imaging systems. It forms an image by solving an inverse mathematical problem, using a priori knowledge of the imaging system. Some technologies developed in the past decade include synthetic aperture microscopy [141], Fourier ptychographic microscopy [142,143] and lensless imaging [144,145]. Optical synthetic aperture approaches rely on scanning the frequency space [141] using holographic [146] or interferometric [147,148] techniques. A computational step is then employed to reconstruct the image. Synthetic aperture techniques were developed for radar to improve resolution by moving the recording system (e.g. mounted on an airplane) and combining multiple overlapping signals [149]. Image resolution is determined by the compounded aperture, which is wider than the real aperture.

Lensless optical imaging systems achieve high-resolution, wide-field imaging [144,145] by replacing the lens with an aperture mask that modulates the illumination and encodes high-frequency spatial information in the available aperture. The use of aperture masks to encode spatial information has also been considered in ultrasonic imaging [150-152]. The mask scrambles the incident sound field, destroying the spatial coherence, similar to speckle formation. The resulting fine-scale structure in the beam enables enhanced resolution imaging, as in imaging through scattering media [153].

Recently, we showed that knowledge of the complex ultrasound field created by the aperture mask allows 3D imaging using only one sensor [152]. SUM is based on that concept: acoustic microscopy with phase-encoded insonification. It employs a 3D-printed mask to encode 3D

spatial information in the acoustic delay, and inverse modeling of the imaging system to achieve wide-field high-resolution volumetric imaging. In this way, we can overcome the limitations of SAM, where dense sampling is required and high resolution is achieved only at the focal region of the transducer. We also investigate synthetic aperture microscopy with computational reconstruction, without an aperture mask (which we call Reconstructed Ultrasound Microscopy; RUM), to independently assess the benefits of computational reconstruction and beam scrambling.

#### 8.2 METHODS

The mathematical model for computational reconstruction [152] describes the pulse-echo signal for each voxel in the imaging region for a given ultrasound field (with and without the mask), scanned over a sparsely sampled grid. We denote this as the system matrix H. Volumetric imaging is achieved by inversion of this model. We denote the measurement vector u, and we assume it is linearly related to the image v through the following equation:

$$\boldsymbol{u} = \boldsymbol{H}\boldsymbol{v} + \boldsymbol{n} \tag{8.1}$$

where n is additive noise. The image v, containing the scattering amplitude per voxel, can be computed with linear inversion methods, such as regularized least squares inversion using the LSQR algorithm [152].

As shown in Figure 8.1, a custom-built unfocused single-element piezoelectric ultrasound transducer (20 MHz, 1 mm diameter) was mounted on a rigid tube and clamped on a motorized positioning stage (Newport Corporation, CA, USA). The coding mask was fabricated using stereolithography (Nanoscribe, Karlsruhe, Germany). It consisted of square pillars (80 µm sides; comparable to the wavelength in water) with random heights (between 40 and 320 µm). This mask effects local delays of up to several wavelengths to the wave field. The speed of sound in the printed resin is 2500-2700 m/s [154]. The mask was slightly larger than the transducer to facilitate assembly. After mounting the mask on the transducer we covered it with silicone to improve the mechanical stability. The transducer was excited with a 10 ns, 60 V unipolar pulse (Avtech Electrosystems Ltd, USA). After amplification by 40 dB (MITEQ, USA) and bandpass filtering (10–30 MHz, 5<sup>th</sup> order Butterworth), the received echo was digitized at 400 MS/s and 12-bit resolution (Agilent Technologies, USA). A custom-built expander/limiter electrically separated the transmit and the receive signal path. All the measurements were performed in a water tank at room temperature.

The spatiotemporal impulse response of the two transducers (with and without mask) was mapped by first scanning a plane perpendicular to the propagation direction, using a 75  $\mu$ m needle hydrophone (Precision Acoustics, UK). The spatiotemporal impulses were acquired

on a 4 x 4 mm grid with dense 30  $\mu$ m spacing at a distance of 900  $\mu$ m from the transducer. Using the angular spectrum of the measurements in this plane, we then computed the forward field impulse responses of the voxels in any parallel plane [155]. Finally, assuming reciprocity, we autoconvolved the forward field impulse-response of each pixel to get their scattered echo signals. The resulting traces are then used to populate the system matrix *H*. All the signal processing is implemented in Matlab 2016b (The MathWorks, MA, USA). To reduce the computational complexity, we chose a sparse Fourier-domain representation of each voxel signal (bandlimited between 15 and 25 MHz).



**Figure 8.1** (a) Top view of the aperture mask imaged with a microscope, (b) schematic drawing of the aperture mask. (c) transducer with the mask mounted on the translation stage.

## 8.3 RESULTS

The ultrasound field transmitted by the single element transducer without mask exhibits a mild natural focus at approximately 3 mm (Figure 8.2). When the mask is placed in front of the transducer, the scrambled phase of the ultrasound field leads to a complex interference pattern. This interference pattern reduces the similarity between nearby voxels resulting in pulse-echo signals that are highly uncorrelated (Figure 3(a)), and increases the field-of-view (Figure 8.3(b)). For full volumetric imaging without (RUM) and with mask (SUM), we acquire measurements at different locations, similar to SAM. However, in SAM the width of the focal region determines the lateral resolution in the plane perpendicular to the ultrasound propagation axis (the *x*,*y* plane). This means that high lateral resolution is achieved by a narrow beam and a dense spatial sampling. In both RUM and SUM we resolve the *x* and *y* components in the beam by exploiting the ultrasound spatial information in the imaging plane, without the need for a tightly focused beam. In SUM, the mask produces a broader beam, and so fewer measurements are required to cover the same area compared to SAM and RUM. Even with sparse spatial sampling, there





**Figure 8.2** Ultrasound field transmitted from the sensor without (left column) and with the aperture mask (right column). Propagating medium is water.

We compute the absolute correlations between the pulse-echo signals H in that correspond to the imaging voxels in a single plane for a given measurement pattern. Figure 8.4(a) shows the correlations between the pulse- echo signal of one voxel (scatterer) and the pulse-echo signals of all other voxels in the analysis plane, with a scanning step size of 300 µm both in xand y directions for both the case without (RUM) and with (SUM) the mask. The correlation is always high for neighboring voxels, making it difficult to resolve two different scatterers, and decreases as one moves further away from the scatterer. We observe that with the mask, high image resolution is achieved by fast decorrelation for increasing distances from the scatterer, at the cost of a slight increase in background clutter. We show the absolute correlation for increasing voxel-to-scatterer distance, averaged in the scan plane, in Figure 8.4(b).


**Figure 8.3** (a) voxel-to-voxel correlation distribution between all the voxels in one *x*,*y* plane and (b) -20dB contour plot of the ultrasound beam in *x*,*y*-plane for both the case without and with the mask; (c) schematic of the imaging procedure.



**Figure 8.4** Analysis of the in-plane correlation. (a) Correlation of one reference voxel (located at x=-0.5 and y=0.5 mm) for the case without and with the mask, using a scan step size of 300  $\mu$ m; (b) average of all the voxel correlations versus pixel-to-scatterer distance (mean ± standard deviation).



Figure 8.5 (a) top view of the 3D object imaged with a microscope; reconstructed image using (b) conventional SAM-like imaging, (c) RUM and (d)-(e) SUM.

We experimentally demonstrate computational ultrasound for high-resolution wide-field imaging of a 3D printed object with dimensions  $2.3 \times 2.3 \times 1.2 \text{ mm}$  (Figure 8.5(a)). The object was glued on rubber and submerged in water at 3 mm from the transducer. Reconstruction of the image by mapping the pulse-echo signal envelope to a 3D imaging grid by scanning the region of interest (4×4 mm) with a 30 µm step (SAM-like imaging) results in a poor image (Figure 8.5(b)). Next, we perform a computational reconstruction without and with mask, using a randomly selected subset containing 40% of the densely sampled measurements and using the LSQR algorithm (Figure 8.5(c)-(e)).

Both RUM and SUM result in highly improved resolution. Even without the mask (RUM), the small imperfections in the transducer introduce spatial heterogeneity in the beam allowing reconstruction of the object with recognizable features down to ~150  $\mu$ m. When the mask is used (SUM), the image fidelity is further improved, with the smallest structures (80  $\mu$ m) successfully resolved (Figure 8.5(d)-(e)).

#### 8.4 DISCUSSION AND CONCLUSION

This result also demonstrates that exploiting the spatial information of the ultrasound field results in isotropic resolution. Increased resolution and field-of-view are obtained by encoding 3D spatial information on echo delays using an aperture mask. The gain in resolution comes at the expense of a reduction in SNR by approximately 10 dB. We attribute this SNR loss to the acoustic energy being emitted into a larger volume and to the impedance mismatch of the transducer-mask-water interfaces.

Improvement in the SNR can be achieved by optimizing mask materials and geometry. In this work we used a mask that was 3D printed, hence limiting the material choice to resins suitable for the specific printer. Beside the choice of mask material, accurate optimization of the mask geometry should be performed [156]. By optimizing the mask geometry more spatial diversity could be introduced, such that unique signals are associated to each voxel, leading to a model matrix *H* with highly uncorrelated columns. This will enable recovery of the 3D image with microscopic resolution using fewer measurements. Moreover, if the *H* matrix is characterized with high accuracy, better image reconstruction can be achieved. The hydrophone used in our measurements had a 75  $\mu$ m sensor, equal to the wavelength of the 20 MHz acoustic waves in water. Subwavelength sampling of the ultrasound field will lead to a more complete characterization of *H*, improving image resolution.

The imaging method that we propose here could also be applied to irregular scanning patterns if the transducer position is known at every scan. Compact tracking techniques, such as optics-based systems [48,49], may enable the implementation of SUM in minimally-invasive medical imaging devices. Catheter-based imaging has demonstrated to be a powerful tool for guidance

of cardiac and vascular interventions [157], but present-day imaging catheters do not achieve comprehensive, high-resolution microscopy. SUM hardware is small, cheap, simple, and it is compatible with miniature disposable devices.

In conclusion, SUM is a new form of acoustic microscopy that enables wide-field isotropic highresolution imaging with a simple unfocused ultrasound transducer and a cheap coding mask. This technique may find application in many fields where space is constrained and wide-field high-resolution imaging is required.

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**Discussion and Conclusion** 

#### 9.1 OVERVIEW

Minimally invasive interventions are the preferred choice in different medical fields due to the smaller incisions and the quicker recovery time compared to conventional open surgeries. Image guidance during minimally invasive interventions allows the physician to gain information over the working field and the location of the tools with respect to the anatomy. Imaging helps also to monitor the effects of certain procedures such as cardiac transseptal puncture, heart valve replacement and chronic total occlusion (CTO) crossing. Although many solutions have been investigated and some minimally invasive imaging devices have entered the market, minimally invasive forward-looking (FL) devices are still under development.

This thesis proposes several solutions to achieve minimally invasive 3D forward-looking ultrasound imaging focusing on two major approaches. The first approach consists in developing miniaturized multi-element transducers with optimal acoustic design and smart interconnections. The second approach consists in the development of simple single-element transducers that can be easily integrated in minimally invasive devices and that can be used to achieve forward-looking volumetric imaging by combining ultrasound information with steering mechanisms and a shape sensing system. In this chapter we give an overview of the results achieved and we discuss the applicability of the proposed solutions in image guidance of minimally invasive interventions. Some remaining challenges and future work are also discussed.

#### 9.2 TRANSDUCER GEOMETRY OPTIMIZATION

For every ultrasound transducer, independent on its application, geometry is crucial for optimal acoustic performances. The width-to-thickness ratio of a transducer element should be as small as possible for uniform surface motion. However, if the width is too small, poor radiation impedance is obtained. For high frequency transducers, which are usually preferred for image guidance in minimally invasive interventions, it is very difficult to achieve small width-to-thickness ratios, with the result that non-piston like modes can occur. To restore optimal vibration of wide elements, subdicing was investigated through finite element simulations (Chapter 2). Subdicing elements having a width greater than 0.6 times the wavelength improves the radiation impedance, the time and frequency response and the directivity. Moreover, to achieve the most significant improvements it is sufficient to subdice 70% of the total element thickness, hence making the transducer fabrication process mechanically stable. By optimizing transducer geometry, acoustically efficient transducers can be achieved resulting in images with improved signal-to-noise ratio and improved resolution.

# **9.3** 2D MATRIX ARRAY FOR MINIMALLY INVASIVE 3D FORWARD-LOOKING IMAGING

Several multi-element transducer arrays have been investigated for FL imaging, from linear arrays generating 2D images to 2D matrix arrays for volumetric imaging [29-35,38-44]. To electrically address multiple elements within the limited space of a catheter, electronic circuits with pulsers and multiplexers should be considered. In this way the number of cables that need to fit in the catheter shaft is reduced, leaving space for other components such as steering mechanisms, guidewires and other interventional tools. Application specific integrated circuits (ASICs) have been reported in the literature to reduce the cable count in FL devices, but they still require more than 10 connections [42,43,158]. Efforts have been made in modifying commercially available side-looking intravascular catheters to provide forward-looking imaging [34,76], but SNR and power dissipation are of concern, since only a limited aperture consisting of an annulus of array elements is used to achieve volumetric imaging.

In Chapter 3 we propose a 2D matrix array mounted on top of an ASIC that addresses 80 PZT elements with only 4 micro-coaxial cables. We simulated different transducer designs that could fit the limited aperture at the tip of a catheter and we fabricated the final design on a custom designed ASIC. We successfully characterized acoustically the transducer and demonstrated 3D FL imaging with only 4 cables. The chip and the transducer occupy a donut-shape area with an outer diameter of 1.5 mm and an inner lumen of 0.5 mm. These dimensions are suitable for catheter integration with the inner lumen available for a guidewire.

## **9.4** SINGLE-ELEMENT MINIMALLY INVASIVE 3D FORWARD-LOOKING IMAGING

Besides considering multi-element arrays, in this thesis we investigated approaches where only a single-element transducer is used to achieve 3D FL imaging. It is well known that a single-element transducer can only provide an A-line, hence additional source of information is required to achieve volumetric imaging. To this end, we exploited the combination of steering mechanisms and shape sensing devices together with single-element ultrasound transducers. We envisioned that by combining ultrasound and position information while steering the device, volumetric imaging could be achieved. Smart algorithms and processing methods are required to successfully elaborate the acquired information into volumetric images.

Commercially available minimally invasive instruments are not capable to successfully navigate through complex shapes. Several steerable devices have been proposed in the literature [96,159,160] but they are quite bulky. We therefore considered novel steerable devices and in Chapter 4 we present a novel miniature tip design with 8 Degrees of Freedom (DOF). A prototype of the device having an outer diameter of 2 mm has been manufactured and a single

element transducer together with an optical fiber for shape sensing have been successfully integrated. We demonstrated that, while manually steering the tip, ultrasound and shape data can be acquired simultaneously, providing information over the location of a wire phantom in 3D space. However, since steering is manual and limited by the design of the catheter, sparse and irregularly sampled ultrasound data are obtained. To interpolate the sparse ultrasound data, we implemented adaptive normalized convolution (Chapter 5), an algorithm that employs an anisotropic Gaussian kernel to fill the empty spaces in the sparse image obtained by spatially mapping the received echoes. After optimizing its dimensions, the kernel is rotated according to the local direction of the scanning pattern allowing to reconstruct features that are not recognizable in the sparse images. With this approach we successfully demonstrated FL imaging of the surface of a tissue mimicking phantom and an *ex-vivo* carotid plague using a clinically available steerable device. The use of a clinically employed steerable device ensures easier translation of the novel imaging concept into the clinics. We further extended the sparse image reconstruction method to 3D by imaging structures in an *ex-vivo* pig heart (Chapter 6). Here the kernel is dynamically changing depending on the spatial density of the information. The resolution of the images depends on the scanning pattern and on the kernel shape and size. The resolution changes spatially, being limited by the smoothing kernel in the sparse areas and by the scanning pattern in the dense areas. The effective resolution is in the range of the kernel dimensions and an analytical description is not possible due to the non-uniform sampling of the data.

A way to improve the resolution is to exploit further the ultrasound beam and relate the received echoes to a 3D scatter map within a structured acoustic beam (Chapter 8), of which speckle is a well-known phenomenon (Chapter 7). By placing a coding mask in front of a single-element transducer, the phase uniformity is lost and unique signals are associated to each imaging voxel. A mathematical model is then developed to exploit the scattering object information in the measurement and the LSQR algorithm is used to solve the problem. By translating the imaging sensor, we can enlarge synthetically the field of view achieving 3D FL imaging with microscopic and isotropic resolution.

#### **9.5 LIMITATIONS AND FUTURE PERSPECTIVES**

The solutions proposed in this thesis show the potential of achieving minimally invasive 3D FL ultrasound imaging in several applications.

The 2D matrix transducer presented in Chapter 3 requires only four coaxial cables and has a total outer diameter of 1.5 mm. The small dimensions and reduced cable count will enable easier integration into catheters or other instruments suitable for minimally invasive procedures. Integration of the ultrasound transducer is further simplified in the solutions presented in Chapters 4-6 and 8 where only a single element is employed. Although single-element

transducers require additional sources of information to achieve FL volumetric imaging, this thesis demonstrates that smart designs of novel steering mechanisms and the use of optical shape sensing systems can be used to explore possibilities for devices with small form factors. With future developments towards robust integration, it will be possible to apply the proposed solutions in *in-vivo* settings taking into consideration motion effects and the presence of additional instruments in the imaging field of view, e.g. guidewires.

In prospect, the solutions presented in this thesis may lead to simple and versatile image guidance with steerable catheters that dynamically build up a volumetric image of the working field in real time. FL volumetric imaging could provide accurate representation of the anatomy and generate information regarding the location of therapeutic devices. This will aid the interventionalist to navigate towards the site of intervention while moving the instruments.

The high resolution obtained with SUM in Chapter 8 indicates that accurate representation of the geometry of tiny structures can be achieved. For example, soft spots in a calcified CTO could be visualized and image guidance while maneuvering the guidewire towards it could be achieved.

Moreover, FL ultrasound imaging could be registered to pre-intervention CT scans by exploiting the shape sensing information, hence minimizing the radiation exposure during the intervention.

The efficient device designs and smart algorithms shown in this thesis together with future research and technological development may lead to image guidance in minimally invasive procedures with approaches that, both from the imaging and the technical point of view, look forward.



Figure 9.1 Envisioned FL-IVUS with SUM

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### Summary

Minimally invasive procedures are increasingly replacing traditional open surgeries due to their shorter recovery time, reduced patient pain, reduced risk of infection and less trauma. However, since the physician has no direct view of the working field, visualization of these complex interventions is critical for success. Forward-looking (FL) ultrasound image guidance can aid minimally invasive procedures providing visual feedback of the working field, instrument location and treatment progress. Currently there are no clinically available devices that can provide minimally invasive 3D FL imaging.

In this thesis we explored several innovative solutions towards miniaturized 3D FL imaging. We looked into methods to solve both hardware and image-related challenges resulting in mainly two approaches. The first approach consists in the realization of a complex multi-element transducer with an optimized design and an efficient interconnection and integration scheme. The second approach consists in the use of simple single-element transducers combined with steering catheters, optical shape sensing systems (OSS) and coding masks to achieve 3D FL imaging.

After introducing the importance of FL image guidance for minimally invasive procedures in **Chapter 1**, in **Chapter 2** we focused on the effects of element geometry on the vibration mode and radiation impedance of a 1D transducer array. We demonstrated, through finite element simulations, that subdicing elements having a width greater than 0.6 times the wavelength improves the acoustic performance. This is an important result for high frequency transducers (>10 MHz) since for those transducers small width-to-thickness ratios are difficult to achieve and non-piston like modes are more likely to occur. In the same chapter we also demonstrated that subdicing only 70% of the total element thickness is already enough to achieve optimal acoustic behavior.

In **Chapter 3** we described a 2D matrix transducer with an aperture size of 1.5 mm consisting of 80 elements interfaced with a front-end ASIC that requires only 4 micro-coaxial cables, making possible integration in intravascular catheters. We investigated optimal transducer design in simulations, we evaluated the effects of a practical implementation and we fabricated and tested the final design. Acoustical characterization and initial imaging results demonstrated the potential of the proposed device to achieve 3D FL imaging.

In the effort of reducing the hardware complexity of the ultrasound device, we moved towards investigation on how to perform 3D FL imaging using a single-element transducer. Since a single-element transducer provides only a single A-line, additional source of information is required to achieve volumetric imaging. We propose to combine the single-element transducer with an OSS system in a steerable device. The idea is to acquire multiple A-lines while steering

the device and to find an appropriate algorithm to combine them using the position information from OSS.

In **Chapter 4** we described a novel device with a steerable tip having 8 degrees of freedom. A prototype was manufactured with a single-element ultrasound transducer and an OSS fiber integrated at the tip. We demonstrated the feasibility of combining ultrasound A-lines and OSS position information while manually steering the device to reconstruct in 3D targets located ahead of the device tip.

Since steering of minimally devices is performed manually by the physician during the procedure and it is limited to specific patterns depending on the catheter design, sparse and irregularly sampled ultrasound data are acquired. To interpolate these sparse data we introduced in **Chapter 5** adaptive normalized convolution. We tested the method using a clinically available steerable catheter that was modified to accommodate a single element transducer and the OSS fiber. We successfully reconstructed the image of the surface of a phantom and an *ex-vivo* human carotid plaque. We extend the same method to 3D to reconstruct volumetric images of structures within an *ex-vivo* pig heart using a novel multi-steerable catheter designed for cardiac applications (**Chapter 6**).

In the effort of further improving the image resolution, we investigated and exploited the spatial information within a structured ultrasound beam. Speckle is a well-known phenomenon resulting from distorting an incident coherent wave until all spatial coherence is lost. The basic physics of this phenomenon was reviewed in **Chapter 7**. In this chapter we introduced the concept of speckle and showed that the magnitude of a random phasor sum can be statistically described with a Rayleigh distribution. We also statistically described the scattered field with a constant background and non-uniform phase distribution. In addition, we described the second-order statistics, that relates the speckle size to the resolution cell dimension, and the effect of dynamic range compression on speckle statistics.

In **Chapter 8** we explored the concept of structured ultrasound microscopy, where 3D forwardlooking imaging is achieved by solving an inverse scattering problem. To make the scattering problem better posed, we placed a coding mask in front of a single-element transducer, hence destroying the phase uniformity within the beam and resulting in a unique signal for each imaging voxel. We synthetically enlarged the imaging aperture by translating the transducer across the 3D volume allowing for wide-field 3D imaging at isotropic microscopic resolution using a very small unfocused ultrasound sensor and sparse spatial sampling.

In **Chapter 9** we summarized the achieved results and we discussed limitations and future work needed to successfully implement minimally invasive FL imaging in clinical practice.

### Samenvatting

Minimaal invasieve interventies vervangen steeds meer de traditionele chirurgisch ingrepen door de vele voordelen die hieraan verbonden zijn, zoals kortere hersteltijd, minder pijn voor de patiënt, een verminderde kans op infecties en minder schade. Het nadeel van zo'n procedure is echter dat de chirurg slechter zicht heeft tijdens de operatie. Een oplossing hiervoor zou een voorwaarts kijkende ultrageluid transducent kunnen zijn. Zo'n transducent kan tijdens een minimaal invasieve interventie visuele feedback geven over het weefsel, informatie geven over positionering van de instrumentatie, en de vorderingen tijdens de procedure. Op dit moment is er geen hulpmiddel beschikbaar in de kliniek wat in 3D en voorwaartse (3D FL) beeldvorming kan verzorgen tijdens minimaal invasieve procedures.

In dit proefschrift verkennen we meerdere innovatieve oplossing die kunnen leiden naar een 3D FL katheter. We hebben gekeken naar methoden om problemen omtrent de hardware en de beeldvorming op te lossen, wat heeft geresulteerd in twee benaderingen. De eerste benadering is gericht op het maken van een complexe multi-element transducent, waarbij we het ontwerp hebben geoptimaliseerd, efficiënte schakeling hebben ontworpen en het geheel hebben geïntegreerd. De tweede benadering is een katheter bestaande uit een bestuurbare transducent bestaande uit maar een element, gecombineerd met optische vorm waarneming (OSS) en een coderen masker over het element voor een 3D beeld.

In **hoofdstuk 1** wordt het belang van een 3D FL beeldvorming voor minimaal invasieve procedures verder toegelicht. Waarna in **hoofdstuk 2** het effect van de geometrie van een 1D multi-element transducent ontwerp werd getoetst aan de hand van de vibratiemodus en de stralingsimpedantie. Doormiddel van een eindige-elementenmethode laten we zien dat het opdelen van de transducent in elementen die groter zijn dan 0.6 keer de golflengte van het ultrageluid de akoestische prestaties verbetert. Dit is gunstig voor transducenten voor hoog frequent ultrageluid (>10MHz) omdat voor dit type transducenten geldt dat het lastig is om een kleine verhouding tussen de breedte en de dikte van de elementen te krijgen, wat en voor kan zorgen dat er meer vibraties ontstaan dan alleen een enkele vibratie over de dikte van het elementen hoeft te gebeuren om goede akoestische eigenschappen te krijgen.

In **hoofdstuk 3** beschrijven we een 2D matrix transducent met een apertuur grootte van 1.5mm bestaande uit 80 elementen verbonden met een toepassing specifieke geïntegreerde schakeling(ASIC). Deze ASIC heeft maar 4 micro coaxiale kabels nodig waardoor dit ontwerp kan worden geïntegreerd in een intravasculaire katheter. We onderzoeken het optimale ontwerp van een dergelijke transducent met behulp van simulaties. Hierbij hebben we gekeken naar de praktische implementatie en hebben we het uiteindelijke ontwerp geproduceerd en getest.

De akoestische eigenschappen en de eerste beelden laten zien dat dit ontwerp mogelijk kan worden toegepast voor 3D FL beeldvorming.

In een poging om het ontwerp vanuit een hardware oogpunt simpeler te maken hebben we gekeken naar de mogelijkheid om 3D FL beeldvorming te doen met een transducent met een enkel element. Een enkel element geeft maar een A-lijn, waardoor een andere bron van informatie nodig is om een volume in beeld te kunnen brengen. Daarom hebben we een transducent met een element gecombineerd met een OSS systeem en de katheter stuurbaar gemaakt. Het idee hierachter is dan we meerdere A-lijnen kunnen meten terwijl de katheter wordt gestuurd en zo met behulp van de informatie van de OSS een beeld kunnen reconstrueren.

In **hoofdstuk 4** laten we een ontwerp zien van een systeem met een stuurbare kop met 8 vrijheidsgraden. Een prototype van de systeem is ontwikkeld met daarbij een enkel element transducent en het OSS systeem in de kop geïntegreerd. We laten zien dat het mogelijk is om hiermee op de hierboven beschreven manier een 3D beeld te kunnen reconstrueren. Tijdens die experimenten werd de transducent kop met de hand gestuurd.

Het met de hand sturen van de transducent en de gelimiteerde bewegingsvrijheid van het ontwerp zorgen ervoor dat de data dun en onregelmatig verdeeld is over het volume. Om deze data te kunnen interpoleren introduceren we in **hoofdstuk 5** adaptieve genormaliseerde convolutie. We hebben dit getest met behulp van een klinische beschikbare stuurbare katheter die we hebben aangepast zodat het mogelijk was om de transducent en de OSS fiber te integreren. Hiermee hebben we laten zien dat het mogelijk is om succesvol het oppervlak van een humane carotide plaque in beeld te brengen. We hebben deze methode uitgebreid naar 3D en daarmee structuren in een ex-vivo varkenshart in beeld gebracht, **hoofdstuk 6**.

In een poging om de beeldresolutie te verbeteren hebben we gekeken naar de mogelijkheid om de spatiele informatie van een gestructureerde ultrageluid bundel te gebruiken. Speckle is een bekend fenomeen wat voortkomt uit vervormingen in een coherente golf totdat alle spatiele cohesie verdwenen is. De fysica hiervan is uiteengezet in **hoofdstuk 7**. In dit hoofdstuk introduceren we het concept en laten we zien dat de grootte van de som van een willekeurig fasor statistisch kan worden beschreven met een Rayleigh verdeling. Ook beschrijven we statistisch het verstrooide veld met een constante achtergrond en een niet uniforme fase verdeling. Daarnaast vervaardigen we ook de tweede orde statisiek, welke de speckle grootte relateert aan de pixel grootte. en het effect van het dynamische bereik compressie op de speckle statistiek.

In **hoofdstuk 8** onderzoeken we de mogelijkheid om gebruik te maken van een gestructureerd geluidsveld, waarbij we 3D beeldvorming bewerkstelligen door de inverse van een verstrooiingsprobleem oplossen. We noemen dit concept gestructureerde geluidsveld microscopie. We versimpelen het verstrooiingsprobleem door gebruik te maken van een masker wat op een transducent met een enkel element wordt geplaatst. Door dit masker krijgt het geluidsveld een uniek signaal in elke voxel. Daarnaast verplaatsen we de transducent over een dun verdeeld en niet uniform raster waardoor de hoek van de beeldvorming wordt vergroot. De combinatie van het bewegen en het masker zorgen ervoor dat we een 3D beeld kunnen krijgen met een isotrope en microscopische resolutie terwijl we verder alleen maar gebruik te maken van een piepkleine onfocuste sensor en weinig datapunten.

In **hoofdstuk 9** worden alle resultaten samengevat en bespreken we de beperking en grenzen van het onderzoek. We sluiten af met de stappen die nodig zijn om dit onderzoek verder te brengen naar een succesvolle implementatie in minimaal invasie voorwaarts kijkende katheter voor klinische toepassingen.

### **Publications**

#### **Peer-reviewed Papers**

**Jovana Janjic**, Maysam Shabanimotlagh, Gijs van Soest, Antonius F.W. van der Steen, Nico de Jong, Martin D. Verweij, "Improving the Performance of a 1-D Ultrasound Transducer Array by Subdicing", *IEEE Trans Ultrason Ferroelectr Freq Control. 2016 Aug; 63(8):1161-71*.

**Jovana Janjic**, Frits Mastik, Merel D. Leistikow, Johan G. Bosch, Geert Springeling, Antonius F. W. van der Steen, Gijs van Soest, "Sparse Ultrasound Image Reconstruction from a Shape-Sensing Single-Element Forward-Looking Catheter", *IEEE Trans. On Biomedical Engineering (accepted)*.

Aimee Sakes, Awaz Ali, Jovana Janjic, Paul Breedveld, "Novel Miniature Tip Design for Enhancing Dexterity in Minimally Invasive Surgery", Journal of Medical Devices (accepted).

Mingliang Tan, Chao Chen, Zhao Chen, **Jovana Janjic**, Verya Daeichin, Zu-yao Chang, Emile Noothout, Gijs van Soest, Martin D. Verweij, Nico de Jong, and Michiel A. P. Pertijs, "A Front-End ASIC with High-Voltage Transmit Switching and Receive Digitization for 3D Forward-Looking Intravascular Ultrasound Imaging" (*IEEE Journal of Solid State Circuits, accepted*).

Jovana Janjic, Pieter Kruizinga, Pim van der Meulen, Geert Springeling, Frits Mastik, Geert Leus, Johan G. Bosch, Antonius F. W. van der Steen, Gijs van Soest, "Structured Ultrasound Microscopy" (*submitted to Applied Physics Letters*).

#### **Papers in Preparation**

Jovana Janjic, Mingliang Tan, Emile Noothout, Chao Chen, Zhao Chen, Zu-yao Chang, Robert H.S.H. Beurskens, Gijs van Soest, Antonius F. W. van der Steen, Martin Verweij, Michiel A. P. Pertijs, Nico de Jong, "A 2D Ultrasound Transducer with Front-End ASIC and Low Cable Count for 3D Forward-Looking Intravascular Imaging: Performance and Characterization".

Maysam Shabanimotlagh, Shreyas B. Raghunathan, **Jovana Janjic**, Verya Daeichin, Emile Noothout, Michiel A.P. Pertijs, Johan G. Bosch, Nico de Jong, Martin D. Verweij, "Effect of Subdicing on Performance of an Ultrasound Matrix Transducer for Carotid Imaging: Numerical and Experimental Study".

Douwe van Willigen, Michiel Pertijs, **Jovana Janjic**, Eunchul Kang, Mingliang Tan, Zu-Yao Chang, Emile Noothout, Martin Verweij and Nico de Jong, "A Transceiver ASIC for a Single-Cable 64-Element Intra-Vascular Ultrasound Probe".

#### **Conference Proceedings**

**Jovana Janjic**, Maysam Shabanimotlagh, Martin D. Verweij, Gijs van Soest, Antonius F.W. van der Steen, Nico de Jong, "Quantifying the effect of subdicing on element vibration in ultrasound transducers", *Proc. IEEE International Ultrasonics Symposium (IUS), Oct. 2015, pp 1-4*.

Jovana Janjic, Merel D. Leistikow, Aimee Sakes, Frits Mastik, Nico de Jong, Johan G. Bosch, Antonius F.W. van der Steen, Gijs van Soest, "3D Imaging with a single-element forward-looking steerable IVUS catheter: initial testing," 2016 IEEE International Ultrasonics Symposium (IUS), Tours, 2016, pp. 1-4.

Maysam Shabanimotlagh, **Jovana Janjic**, Shreyas Raghunathan, Michiel Pertijs, Nico de Jong and Martin Verweij, "The role of sub-dicing in the acoustical design of an ultrasound matrix transducer for carotid arteries imaging," *2016 IEEE International Ultrasonics Symposium (IUS)*, Tours, 2016, pp. 1-4.

Mingliang Tan, Chao Chen, Zhao Chen, **Jovana Janjic**, Verya Daeichin, Zu-yao Chang, Emile Noothout, Gijs van Soest, Martin Verweij, Nico de Jong and Michiel Pertijs "A front-end ASIC with high-voltage transmit switching and receive digitization for forward-looking intravascular ultrasound," *2017 IEEE Custom Integrated Circuits Conference (CICC)*, Austin, TX, 2017, pp. 1-4.

#### **Book Chapter**

**Jovana Janjic** and Gijs van Soest, Speckle Physics, in *Handbook of Speckle Filtering and Tracking in Cardiovascular Ultrasound Imaging and Video*, Loizou, Christos P.; Pattichis, Constantinos S; D'hooge, Jan, The Institution of Engineering and Technology, London (UK), 2018, pp. 25-54.

### **Curriculum vitae**



Jovana Janjic was born in Banja Luka (Bosnia and Herzegovina) on April 1989.

With the start of the war in former Yugoslavia she and her family moved first to Belgrade and one year later to Italy, in Caorle, a small coastal town north of Venice.

She received the B.Sc. in Information Engineering in 2011 and the M.Sc. in Bioengineering in 2013 from The University of Padova, Italy. She was selected within the Erasmus Programme as an exchange student at the Royal Institute of Technology (KTH), Stockholm, Sweden, from August 2012 until June 2013. At KTH she completed her Master thesis on sonothrombolysis using contrast agents. In 2013 she joined the department of Biomedical Engineering at Erasmus University Medical Center, where she conducted the research that led to this PhD thesis.

### **PhD Portfolio**

#### Jovana Janjic

| Courses  | year | ECTS |
|--|------|------|
| IVUS, Virtual Histology & FFR, Theory & Practice                         | 2013 | 0.6  |
| Pathophysiology of Ischemic Heart Disease                                | 2014 | 1.5  |
| Micromachined Ultrasonics Transducers, Danish Technical University,      |      |      |
| Copenhagen, Denmark  | 2014 | 3.5  |
| Verasonics course, Taipei, Taiwan  | 2015 | 0.2  |
| Cardiovascular Imaging and Diagnostic                                    | 2015 | 1.5  |
| English biomedical writing and communication                             | 2015 | 4    |
| Signal Processing and System-on-Chip Designs for Ultrasonic Imaging,     |      |      |
| Detection and Estimation Application, IEEE Taipei, Taiwan                | 2015 | 0.2  |
| Advanced Ultrasound Imaging, Danish Technical University, Copenhagen,    |      |      |
| Denmark  | 2015 | 2.5  |
| Microultrasound Transducers – Materials, Instrumentation and             |      |      |
| Performance, IEEE, Tours, France   | 2016 | 0.2  |
| Elasticity Imaging: Methods and Applications, IEEE, Washington D.C., USA | 2017 | 0.2  |
| High-Performance Computing for Medical Imaging on Programmable           |      |      |
| Graphics Hardware (GPU)  | 2017 | 0.2  |
| Workshops and Seminars   |      |      |
| Seminar: New imaging strategies for the detection of atherosclerosis     | 2014 | 0.2  |
| Seminar: Current cardiac and vascular aging research at EMC              | 2014 | 0.2  |
| NVMU Autumn meeting, Maastricht  | 2013 | 0.15 |
| NVMU Spring meeting, Enschede  | 2014 | 0.15 |
| NVMU Autumn meeting, Delft   | 2014 | 0.15 |
| NVMU Autmn meeting, Eindhoven, (Oral)                                    | 2015 | 0.2  |
| Workshop: Photoshop and Illustrator CS5                                  | 2015 | 0.3  |
| iMIT meeting, Delft (Oral)   | 2013 | 0.6  |
| iMIT meeting, Eindhoven (Oral)   | 2014 | 0.6  |
| iMIT meeting, Rotterdam (Oral)   | 2017 | 0.6  |
| COEUR PhD-day 2014   | 2014 | 0.6  |
| COEUR PhD-day 2015   | 2015 | 0.6  |

#### International conferences

|  | тот       | 39.65 |
|--|-----------|-------|
| Ultrasound Contrast Imaging Symposium, Rotterdam, The Netherlands    | 2014-2018 | 2.7   |
| Optics in Cardiology   | 2015      | 0.6   |
| Optics in Cardiology (Poster)  | 2017      | 0.9   |
| Design of Medical Devices, Delft, The Netherlands (Oral)             | 2017      | 1.5   |
| Design of Medical Devices, Delft, The Netherlands (Oral)             | 2016      | 1.5   |
| Design of Medical Devices, Delft, The Netherlands (Poster)           | 2014      | 1.2   |
| Artimino Conference, Helsinborg, Sweden (Oral)                       | 2015      | 2.1   |
| Edinburgh, Scotland, UK (Oral)                                       | 2014      | 2.1   |
| International Conference on Ultrasonic Biomedical Microscanning,     |           |       |
| Kralendijk, Bonaire, The Netherlands (Oral)                          | 2016      | 2.1   |
| International Conference on Ultrasonic Biomedical Microscanning,     |           |       |
| IEEE International Ultrasonic Symposium, Washington D.C., USA (Oral) | 2017      | 2.1   |
| SPIE Medical Imaging, Orland, Florida, USA (Oral)                    | 2017      | 2.1   |
| IEEE International Ultrasonic Symposium, Taipei, Taiwan (Poster)     | 2015      | 1.8   |

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