Method for measuring violin sound radiation based on bowed glissandi and its application to sound synthesis

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This work presents a method for measuring and computing violin-body directional frequency responses, which are used for violin sound synthesis. The approach is based on a frame-weighted deconvolution of excitation and response signals. The excitation, consisting of bowed glissandi, is measured with piezoelectric transducers built into the bridge. Radiation responses are recorded in an anechoic chamber with multiple microphones placed at different angles around the violin. The proposed deconvolution algorithm computes impulse responses that, when convolved with any source signal (captured with the same transducer), produce a highly realistic violin sound very similar to that of a microphone recording. The use of motion sensors allows for tracking violin movements. Combining this information with the directional responses and using a dynamic convolution algorithm, helps to improve the listening experience by incorporating the violinist motion effect in stereo.

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I. INTRODUCTION

In a simplified model of violin sound production and radiation, we can consider two main vibrating elements: the bowed-string interaction, which is the nonlinear component, and a system comprised of all the elements in the signal path from bridge vibration to sound perception, which may be considered as nearly linear (Cremer, 1984). This linear part is often referred to as the violin body, but it is actually a coupled system whose main components are the bridge, the violin body and even the room. Understanding the properties of the violin body is of great interest as it largely influences the sound quality of the instrument. It is usually modeled as a linear filter (Farina et al., 1995, 1998; Bissinger, 2006) by measuring its response to a known excitation and computing the body frequency response (BFR) or body impulse response (BIR). There exist many different methods to obtain BFRs, which are adapted to the particularities of the research. There can be found works aimed at the comprehension of specific physical phenomena (Jansson et al., 1970; Cremer, 1984; Marshall, 1985; Jansson et al., 1994; Wang, 2000) and perceptual features related to the properties of the violin body (Woodhouse, 2005; Fritz et al., 2006, 2007). Some studies report progress in violin making (Schleske, 2002). Finally, there can be found works making use of BFRs for sound synthesis (Smith, 1983; Cook and Trueman, 1998; Vos et al., 2003; Waxman, 2005). However, their main objective has not been highly realistic synthesis, which is the goal of this work.

The aim of this study is to develop a general technique to obtain BFRs (and their respective BIRs) between the signal captured with a pickup on the violin and the radiated sound pressure in different directions relative to the body. The pickup is built into the bridge and we consider both the bridge and the pickup to be part of the body. The intention of our approach is not physical modeling, so the precise point in the signal path at which the pickup is mounted and the correlation between the pickup signal and some physical quantity are irrelevant. The main requisite for the obtained BIRs is that they must produce a highly realistic sound when convolved with a source signal [recorded with the pickup or synthesized with a signal model (Perez, 2009)]. The proposed methodology is based on bowed glissandi that are recorded with the pickup (excitation) and several microphones (responses). The frequency responses are obtained by computing the transfer functions between the pickup and each microphone as a weighted average of the deconvolution of short temporal frames. From the measurement and analysis described above, the estimated BFRs are incorporated into an enriched sound synthesis of violin performance. The use of BFRs allows us to imitate the sound of different violins as in the work by Fritz et al. (2006), simulate stereo listening as reported in Waxman (2005) and simulate different positions of the listener. Furthermore, provided with a motion sensor that tracks the position of the violin, we can simulate the effect of a moving performer. This is achieved by applying the convolution at each instant with the respective BIR.
II. RELATED WORK

Many different methods are used to obtain the violin BFR, usually based on measuring the body’s response to a known excitation. They can be differentiated by the place where the excitation is applied and type of excitation signal, by the place where the response is measured and the type of response, and by the algorithm to compute the transfer function from excitation and response signals, which typically is based on signal deconvolution (Smith, 1997) or system identification (Smith, 1983).

The excitation signal may be impulsive or continuous. Impulsive excitations can be performed by hitting the violin bridge with a force hammer (Woodhouse, 2005). The impulse may be applied directly to the strings in order to obtain a different BFR for each string (Cook and Trueman, 1998). Non-impulsive excitations, e.g., sine sweep or MLS (Farina et al., 1998), are usually less affected by noise and measurement errors and may be applied by means of a shaker or wire force transducer (Farina et al., 1998) or following the “indirect” method with a loudspeaker (Weinreich and Arnold, 1980; Farina et al., 1995; Vos et al., 2003). Other possible methods are finger tapping, traditionally used by violin makers, or playing the instrument as in Karjalainen et al. (2000), where a guitar is excited by playing musical excerpts.

Violin body transfer functions can be mechanical or acoustical depending on whether the response is measured as mechanical vibration or sound pressure. Mechanical vibration is usually measured as velocity or acceleration at the bridge with an optical laser (Woodhouse, 2005), with an accelerometer (Morset, 2001), or with a magnet inside a coil (Jansson, 2002). Acoustical responses are measured with a microphone, ideally in an anechoic chamber in order to avoid the room reverberation.

In this work, sound radiation is measured at different angles around the violin, making it possible to obtain radiation patterns (Weinreich and Arnold, 1980; Williams et al., 1980; Cook and Trueman, 1998; Otčenášek and Syrový, 1999; Wang and Burroughs, 2001; Bissinger, 2008a). Weinreich (1997) and Bissinger (2008b) have extensively studied the directional radiation from violins. In general, previous studies concentrate on a limited range of low frequencies but for our synthesis purposes we need to obtain the response over the full audio frequency range.

Violin sound synthesis based on a source plus filter approach is often used in the field of violin acoustics, but the final objective has not been highly realistic synthesis, as is the case in our work. For example, Otčenášek and Syrový (1999) report a perceptual analysis about directivity of the violin timbre, and Fritz et al. (2006) carry out psychoacoustic experiments with virtual violins. Holm and Välimäki (2000) investigate modifications of the resonances of violin body modes and Smith (1983) applies filter design techniques to model violin body resonances.

III. METHODOLOGY

The methodology for obtaining the violin BFR is based on a similar procedure as reported by Karjalainen et al. (2000), who present a method to compute guitar acoustic transfer functions. They excite the guitar by playing tones and measure the excitation with an under-saddle pickup and the response with a microphone, and finally compute the BFR by signal deconvolution. Our approach is based on playing glissandi that were recorded with an instrument pickup (excitation) and a microphone (response), and we estimate the BFR with an algorithm based on signal deconvolution.

A. Excitation and response measurement

Several typical excitation methods were tried and abandoned in favor of playing musical excerpts with the bow. The main reason is that common methods for obtaining BFRs are based on applying a known excitation signal (e.g., an impulse, a pseudo-random noise sequence, or a sine sweep), but in our case, the excitation signal is not the applied signal, but the signal captured by the pickup. Furthermore, the commonly used methods are not adapted to our specific needs. For instance, by striking the instrument with a force hammer, we obtained a non-noisy BFR up to around 10 kHz and the noise of the hammer hitting the bridge becomes a problem as it is captured by the microphone. The use of a shaker adds an extra mass to the bridge, and we could only obtain an acceptable response at low frequencies. Finally, it was decided to excite the violin by bowing as it does not interfere when measuring the response with a microphone and it is the natural way of exciting the violin. Moreover, the measurement conducted this way includes the influence of the performer, which is desirable for our synthesis purposes. Playing glissandi was preferred rather than performing other musical exercises, as it is a simple method to sweep and homogenously excite all frequencies in a wide frequency range.

Glissandi were performed on the G-string, beginning at the open string and with a range slightly wider than one octave. Using the G-string we obtain the lowest playable pitch of 196 Hz on the violin with the tone G3. A glissando over the lowest playable octave is sufficient for exciting the entire frequency range of interest, and no significant difference in sound quality was perceived when using BFRs obtained from glissandi with higher pitch extensions. It is possible to estimate individual BFRs for each string. However, this idea was abandoned for simplification and because the result of the convolution was perceptually found to be nearly identical according to informal listening. Glissandi lasted for approximately 50 s as they must be played slowly in order to be nearly stationary during each temporal frame (see next subsection). They were performed with an almost constant rate of increasing pitch (semitones/s) that would approximately correspond to a logarithmic frequency sweep. Approximately fifteen bow strokes were necessary to complete an entire glissando.

The excitation was measured with a pickup. Several sensors and commercial pickups were tested and the selected device (a piezoelectric pickup integrated with the bridge) was chosen based on the quality of the corresponding convolution with a source signal. The response was measured with several Rode-NT1A microphones, and a compensation filter was applied in order to achieve an ideal flat response in
the same manner as in Pätyinen et al. (2008). We evaluated
the spectral richness of the excitation glissando from histo-
grams of the energy contribution to each spectral bin. In
Fig. 1 we show the resulting histogram for a glissando
recorded with the bridge pickup and compare it to the spec-
trum of one of the first frames of the same glissando. The
spectrum envelope falls smoothly 70 dB until 10 kHz. A
clear partial structure is present in the spectrum of the single
frame up to that frequency. Above 10 kHz the spectrum level
is approximately flat.

Both distance and orientation of the microphones relative
to the violin affect the properties of the recorded sound,
and it is desirable to obtain a consistent estimation after
time-averaging individual frame estimations. For this reason,
we built a mechanical structure to keep the violin fixed, pre-
venting the performer from moving the instrument while
playing the glissando.

B. Energy weighted multiple frame deconvolution

Typical ways of obtaining violin body frequency
responses are by means of system identification techniques
(Smith, 1983) or by signal deconvolution (Farina et al., 1998;
Karjalainen et al., 2000). Applying an exact deconvolution by
division in frequency domain to signals, such as the proposed
glissando, would result in a very noisy frequency response.
There are a few reasons for this. First, the experimental mea-
surement has limited frequency resolution, and further, the
presence of noise and the use of a simplified model ignoring
potential non-linearities and time-variance lead to inaccuracies.
Hence, we compute the deconvolution on a frame-by-frame
basis followed by a global averaging across all frames of the
glissando in which each frame-deconvolution is weighted
according to the energy of the excitation signal in that frame.

The signal recorded with the microphone has a delay of
approximately 6 ms, corresponding to the distance of 2 m
from the violin. First, this delay is removed to align excita-
tion signal and response signal. Then, both signals are
windowed and expressed in the spectral domain, obtaining
two frame streams that are aligned in time \(X(f,k)\) and \(Y(f,k)\),
where \(f\) is the frame and \(k\) is the spectral bin. At this point,
we apply a frame-by-frame deconvolution. The magnitude is
obtained as

\[
|H_{\text{BFR}}(f,k)| = \frac{|Y(f,k)|}{\max(|X(f,k)|,b)},
\]

where the value \(10^{-5}\) was chosen for the threshold \(b\). The
phase is obtained as

\[
\phi[H_{\text{BFR}}(f,k)] = \phi[Y(f,k)] - \phi[X(f,k)] + 2\pi m,
\]

where \(m\) is the integer that wraps the phase difference in the
interval \([-\pi,\pi]\). Now, having estimated the BFR for individual
frames, separate BFRs are averaged in order to obtain an
accurate estimation of the overall BFR.

For the magnitude estimation, we use the typical
weighted average expressed as

\[
|H_{\text{BFR}}(k)| = \frac{\sum_{1 \leq f \leq N} w_f(k) \times 20\log_{10}|H_{\text{BFR}}(f,k)|}{\sum_{1 \leq f \leq N} w_f(k)} \text{ [dB]},
\]

where \(f\) is the frame number, \(N\) is the number of frames, \(k\) is
the spectral bin being estimated, and \(w_f(k)\) is the weight
being applied, which corresponds to the input signal energy
at frame \(f\) and bin \(k\).

Regarding the phase, due to its cyclic behavior, carrying
out a classical weighted average would not provide good esti-
mations. As a first attempt, we explored a method based on
constructing a histogram of the phase values estimated for each
spectral bin, weighted by their corresponding energy. However,
the resulting BIRs were not causal, so finally we computed the
minimum phase BIRs from the estimated BFR magnitudes by
using the cepstrum and converting anti-causal exponentials to
causal exponentials (Oppenheim and Schafer, 1975).

Glissandi were recorded at a sample rate of \(f_s = 44100\)
Hz and spectrally analyzed with the following parameters. A
Blackman-Harris window with length of 8192 samples (186
ms) is used with a window overlap of 50% and a zero-padding
factor of 8 (thus a FFT size of 65536). With these analysis
parameters we achieve a bin resolution in the computed BFR of
approximately 0.67 Hz. The effective length of the resulting
BIRs corresponds to half window length, that is, 93 ms.

The proposed deconvolution algorithm assumes that the
glissando is quasi-stationary within the analysis window in

![Excitation Richness of the Pickup](image)

**FIG. 1.** Observation of the spectral content of the input excitation signal from the bridge pickup used for
deconvolution. Histogram maxima
for all spectral bins across a gliss-
sando (gray) are compared with the
spectrum of one of the first frames
of the same glissando (black).
order to assure that the energy of the short-time spectrum of an arbitrary frame is mostly concentrated around harmonic frequencies. Therefore, the use of a relatively short analysis window (around 186 ms) implies that a slow glissando must be used as the input signal. In order to illustrate the effect of this energy distribution, let us consider an ideal case where the input signal \( x(t) \) is a complex exponential at frequency \( f_0 \), amplitude \( A \), and initial phase \( \phi_0 \)

\[
x(t) = Ae^{j(2\pi f_0 t + \phi_0)}.
\]

When this signal is filtered, the output signal obtained \( y(t) \) is equal to the input multiplied by a phasor determined by the filter frequency response,

\[
y(t) = Ae^{j(2\pi f_0 t + \phi_0)}H(f_0),
\]

where \( H \) is the frequency response of the filter.

In practice we must consider discrete signals, obtained by sampling at a fixed rate of \( f_s \) samples per second. The discrete Fourier transform (DFT) of a windowed complex exponential is given by

\[
X_w(k) = \sum_{n=0}^{N-1} w(n)x(n)e^{-j2\pi kn/N} = \sum_{n=0}^{N-1} w(n)Ae^{j(2\pi n/N + \phi_0)}e^{-j2\pi kn/N} = Ae^{j\phi_0}W(k - k_0),
\]

where \( n \) is the sample index, \( k \) the spectral bin, \( N \) the length of the DFT, \( w \) the analysis window, \( W \) its transform, and \( k_0 = (f_0/f_s)N \). The DFT is equal to the phasor of the exponential multiplied by the transform of the window shifted to the frequency of the exponential. The estimated filter \( \tilde{H} \) is computed by dividing the Fourier transforms of the windowed output and input discrete signals,

\[
\tilde{H}(k) = \frac{Y(k)}{X(k)} = \frac{N^{-1} \sum_{n=0}^{N-1} y(n)x(n)e^{-j2\pi kn/N}}{N^{-1} \sum_{n=0}^{N-1} w(n)x(n)e^{-j2\pi kn/N}} = \frac{Ae^{j\phi_0}W(k - k_0)H(k_0)}{Ae^{j\phi_0}W(k - k_0)} = H(k_0),
\]

that is, the estimated filter is constant for all spectral bins, and equal to the value of the actual filter at the exponential frequency. If we consider as input two complex exponentials of frequencies \( k_0 \) and \( k_1 \) cycles per sample, the estimated filter becomes

\[
\tilde{H}(k) = \frac{Y(k)}{X(k)} = \frac{N^{-1} \sum_{n=0}^{N-1} y(n)x(n)e^{-j2\pi kn/N}}{N^{-1} \sum_{n=0}^{N-1} w(n)x(n)e^{-j2\pi kn/N}} = \frac{Ae^{j\phi_0}W(k - k_0)H(k_0) + A_1e^{j\phi_1}W(k - k_1)H(k_1)}{Ae^{j\phi_0}W(k - k_0) + A_1e^{j\phi_1}W(k - k_1)}.
\]

which is a weighted sum of the actual filter values at the exponential frequencies, where the weighting depends on the transform of the analysis window. For typical windows the energy of the transform is concentrated around its main lobe. Consequently, the estimated filter is approximately constant around each exponential frequency.

When we combine several frame estimations as discussed above, each harmonic contributes to the averaged response with a weight, which is determined mostly by its energy and the transform of the analysis window. Hence, the ratio between output and input signals can be considered as a sampling of the target filter at harmonic frequencies. This is shown in Fig. 2.

When the analysis window is longer or the glissandi are performed faster it may happen that the excitation is no more quasi-stationary within the analysis window. This would have the effect of widening the harmonic peaks especially at mid and high frequencies (the harmonic frequency shift is larger at high frequencies than at low frequencies), negatively affecting the estimation of the filter. This is so due to the fact that we are performing a short-time analysis with a window that does not include the whole excitation and response signals. Therefore, it is recommended to use slow glissandos. Besides, for a good filter estimation, it is necessary that the sweeping harmonics along the whole glissando excite all the spectral bins in order to avoid zeros in the excitation spectrum.

In Fig. 3 we show the obtained BFR magnitude at the same microphone and violin position for four different glissando recordings to show the repeatability of the proposed method. BFRs are seen to be very similar, although there can be found differences among them that seem to vary along frequencies. The reason is that the responses are affected by the exact position of the violinist when playing, especially his head.

**IV. MEASUREMENTS**

The objective of the measurements is to obtain sound radiation levels around a violin at all frequencies covered in normal playing by bowing. This is achieved by computing the transfer function between the bridge pickup and a set of microphones placed at different angles. BFRs can be convolved subsequently with a source signal in order to simulate...
the position of the listener with respect to the violinist. The BFR at each angle is estimated as described in the previous section. The measurements were performed using a violin of professional quality.

A. Setup and calibration

The measurements were conducted in an anechoic chamber at Aalto University with anechoic conditions down to about 125 Hz. Setup and microphone calibration were performed following the same basic concept as in the work by Pätynen et al. (2008). The measurement room was equipped with 21 Røde microphones placed around the center of the room in a spherical-like arrangement (two pentagonal anti-prisms) at distances from the center between 1.7 and 2.4 m. The microphones were grouped in four horizontal planes at nominally 10 and 50 deg above and below the equatorial plane, with an additional microphone close to the top. In each horizontal plane there were five microphones evenly spaced (around 72 deg between consecutive microphones). All angles and distances relative to the center of the room were accurately measured with laser equipment.

Due to the different positions of the microphones in relation to the absorbing wedges on the walls, varying distances from the room center, and the inherent differences in frequency response between microphones, the same procedure as in Pätynen et al. (2008) was performed in order to equalize the responses of the recording microphones.

We built a mechanical structure to prevent the performer from moving the violin while playing the glissandi. The structure is thin to minimize its influence on the measurements and it can hold the violin with two standard shoulder rests (Fig. 4). The support of the violin may be assumed to be similar to normal playing conditions when the performer is standing.

B. Coordinate systems

The violin was placed at the center of the room by defining 3D coordinate systems for the violin and room, respectively, and matching the origins of the two systems with the help of an OptiTrack system² with six cameras.

The origin of the room system was set at the center of the room and was determined with the help of a laser pointer and marks on the walls at the same height, separated by 90 deg. The room x-y plane was the equatorial plane of the microphone array, which is parallel to the floor (measured with levels).

The origin of the violin coordinate system (see Fig. 4) was defined as the midpoint between the projections of the E and G string-to-bridge terminations on the top plate plane. The violin x axis was in the direction of the neck with the x-y plane being parallel to the top plate.

The violin was mounted on the supporting structure and fluorescent markers were attached to it (Fig. 4). With the OptiTrack visualization tool we make violin and room coordinates coincide, making it possible to measure the angle differences between the two coordinate systems.

The supporting structure can be easily rotated to fix the violin at a specific orientation (see Fig. 4), determined by the violin azimuth (azv) and inclination (incv) with respect to the

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room system (Euler angles). The azimuth is changed through rotation of the violin around the z-axis (by rotating the supporting structure around the vertical axis of the stand), and the inclination through rotation around the violin x axis (by rotating around the pivoting axis of the shoulder rest). Note that the shoulder rest axis is parallel to, but not coinciding with the violin x axis, so each time the inclination is changed, violin and room systems need to be realigned. In addition to the OptiTrack system, the violin azimuth is measured with a protractor, placed at the base of the supporting structure. Violin elevation ($e_l$) was kept constant in the measurements at 0 deg, meaning that the violin z-axis always coincided with the equatorial plane of the microphone array. In Sec. V B the sampled microphone (or listener) positions are expressed in spherical coordinates with respect to the violin, which are determined by the listener azimuth ($a_{zl}$) and elevation ($e_l$), see Fig. 8. Note that moving the violin by rotating the structure, is equivalent to moving the listener to the opposite direction.

### C. Directional radiation

Radiation is understood as the quantity of radiated sound energy, which varies spatially around the violin. Directivity is usually defined as the radiation ratio among different directions at a certain frequency (Weinreich, 1997). As a result of the measurements, we obtained several BFRs around the violin. Although the estimated BFRs are not directly related to a physical property of the violin (because of the pickup), the ratio between BFRs shows the radiation ratio at different directions.

The measurements were conducted as follows. The violin was mounted in the structure and set to a certain inclination ($inc_v$). For that inclination, the 360 deg of azimuth ($a_{zv}$) were sampled in steps of 5 deg. As each horizontal plane has five microphones, it was sufficient to record 15 glissandi at steps of 5 deg going from one microphone to the next in the same horizontal plane. The same process was repeated for inclinations at 0, 30, 60, and 90 deg in order to obtain the sampled space represented in Fig. 5, in all including 1260 sampled positions (each cross indicates a sampled position). Violin elevation was kept constant at 0 deg, as mentioned. However, by measuring the violin with different azimuth and inclination angles, also changes in elevation can be derived from the performed measurements.

Figure 6 shows the radiation ratio between two randomly selected directions (with the same $e_l$ and separated by 72 deg $a_{zv}$). Although the ratio is different for each pair of directions, the plot is representative for the general behavior. Below 600 Hz, radiation is nearly identical in both directions (the difference is around 1 dB). From 600 to 1000 Hz the radiation ratio is small although some peaks are present. The highest differences appear in the range 1000–4500 Hz, reaching in some cases 20 dB.

In Fig. 7 we visualize 2D polar directivity curves in dB at specific frequency ranges measured with five microphones located in the same horizontal plane (10 deg above the equatorial plane of the microphone array). The center of each polar graph (0 dB) represents the minimum radiated value at that frequency. Curves close to the center (i.e., a short radius) mean small differences in energy radiation among different angles, while a long radius indicates a high radiation ratio.

The graphs in Fig. 7(a) (violin inclination 0 deg) show the radiation in a plane parallel to and 35 cm above the x-y plane and those in Fig. 7(b) (inclination 90 deg) in a plane parallel to and 35 cm to the left of the x-z plane (bass bar side). The directivity patterns are seen to change greatly,
even between very close frequencies. A video showing the evolution of the patterns along the frequency can be found in a dedicated web page.

In both planes in Fig. 7 we can observe how at low frequencies (196–489 Hz) directivity is very low. The radiation is almost omnidirectional except at the position of the player who introduces a shadowing effect. In the x-y plane in the range 580–700 Hz, shapes seem to be symmetrical with respect to an axis at −60 deg, forming a clear pattern-of-eight which indicates a dipole source. Again, lowest radiation values are found at the position of the violinist. In the range 760–828 Hz most of the energy is radiated in the entire frontal direction of the violin, and in the range 1050–1200 Hz to the right side. For the x-z plane in the range 580–700 Hz, most of the energy is radiated from the back, whereas in the range 760–828 Hz it is rather symmetrical with respect to the violin x axis (neck direction), but with a stronger radiation from the top plate. The radiation in both directions of the x axis is low. At higher frequencies 1050–1200 Hz we observe a five-arm star shape.

In general, the observed behavior is consistent with previous studies (Weinreich, 1997). The main difference is that the performer is included in our measurements. It can clearly be observed that the lowest radiation values appear in the direction he is standing.

D. Discussion

The brief description of the radiation patterns above shows that the observations are in agreement with previous studies (Weinreich, 1997). A more comprehensive analysis of the patterns will be presented elsewhere, as in this presentation we want to focus on their application for sound synthesis. Furthermore, the observed results were obtained with only one violin of professional quality. By analyzing a set of violins of different quality we may find specific features that distinguish fine violins from others and, at the same time, we may find properties common to all violins, which would allow for defining a set of generic violin BFRs.

V. APPLICATION TO SOUND SYNTHESIS

Given a source signal, either recorded with the bridge-pickup or synthesized (Perez, 2009), we can obtain a realistic sound close to that captured by a microphone, by convolving it with the estimated body impulse responses (BIRs). Convolution is performed via the frequency domain following the procedure reported in Arfib et al. (2002). A rectangular window is used to select an input signal frame of the size of the BIR, which is zero padded and multiplied in the frequency domain with a zero padded BFR. The result, which is twice the length of the window, is transformed back to the time domain and overlap-added with previous data. The window is shifted by the length of the BIR and the same procedure is repeated.

A. Simulation of stereo and performer motion

As we computed several BFRs around the violin we can recreate the sound perceived by a listener at different directions and provide a better listening experience by simulating effects such as stereo or the movement of the player. As a simplification we consider the listener to be seated still at a distance of 2 m (the same distance as from the microphones to the violin during the measurements), and we assume that the ear-to-ear angle difference as viewed from the violin can be approximated as five degrees at that distance. In this manner, stereo can be simulated directly by using two consecutive BIRs in any horizontal plane, one for each ear.

The movement of the performer is another effect that is necessary to be included in a realistic simulation. Ancillary gestures are constantly changing the position and orientation of the violin, so the listener is perceiving the sound radiated
at continuously changing directions. To simulate this motion we need to know the position of the listener relative to the violin and apply the convolution with the corresponding BIR at each instant. This effect is more clearly perceived in stereo, so the same procedure must be performed for both ears. In order to demonstrate this effect, we made use of a multimodal database of real violin performances that contains recordings of the pickup signal together with the position and orientation of the violin. Any recorded pickup signal may be then convolved with the BFRs corresponding to the coordinates of the violin. The method is briefly explained in the following subsection and some sound samples are found in a dedicated web page.3

B. Listener coordinates

The position of the listener with respect to the violin is defined by the listener azimuth (azl) and elevation (ell) for each ear. Violin coordinates during performance were acquired with a Polhemus4 3D-motion tracker, consisting of an electro-magnetic field source and a small sensor attached to the violin. The system tracks the position and orientation of the sensor with respect to the source with a frame rate of 240 samples/s. Therefore, the problem consists of converting the recorded sensor coordinates to listener coordinates with respect to the violin system. Figure 5 shows the trajectory of the listener relative to the violin during a performance, together with the position of all sampled BFRs.

Let us consider three coordinate systems (Fig. 8), the source given by the axis vector [X, Y, Z], the sensor (S) with axes [Xs, Ys, Zs] and the violin (V) with axes [Xv, Yv, Zv]. The sensor was attached to the violin, so their coordinate systems have a fixed position and orientation with respect to each other. Similarly we consider the position of the listener relative to the source to be constant. Each coordinate system is represented as a tuple with position and rotation matrices.

Coordinates of a point in any system can be expressed in a second system by knowing the tuple of the second system with respect to the first.

Using standard techniques, updated violin coordinates are computed relative to the source system for every new position and orientation of the sensor [see Maestre et al. (2007) and Perez (2009) for details]. The listener position is then translated to the new violin coordinate system, and finally converted to spherical coordinates (azl and ell) with respect to the violin system. Figure 5 shows the trajectory of the listener relative to the violin during a performance, together with the position of all sampled BFRs.

C. BIR interpolation

Listener position coordinates may not coincide with any of the estimated BFRs, so we need to interpolate among the closest measured locations. Euclidean distance is computed to each sampled BFR in Cartesian coordinates. The n closest BFRs to the listener are selected (n = 10 was used in our tests). The interpolation is a weighted sum of the BIRs (in the time domain), where the weight of each BIR is inverse to the distance and the sum of all weights is unity.

D. Dynamic convolution

In order to simulate the effect of the violinist moving we need to apply a convolution with a different BIR at each instant (the BIR corresponding to each new listener position). If we follow the previously introduced convolution algorithm Arfbl et al. (2002), the maximum rate at which a change of BIR can be applied is given by its duration (93 ms). This may result in significant sound changes between consecutive convolutions. If we want to simulate faster changes in the radiation characteristics we cannot use full-length BIRs. We propose to use a dynamic convolution algorithm that involves breaking up a long BIR into smaller chunks and cross-fading among the smaller BIRs in a similar way to Foster et al. (1991), Gardner (1995) and Garcia (2002).

BIRs are divided into n sub-BIRs, and each frame is convolved with all n fragments of its corresponding BIR, and finally the sum of all convolutions is computed. When there is a BIR change, the current frame has to be convolved with both the previous and current BIRs and cross-faded. Crossfading is performed at each frame: If there is no change of BIR between two frames, the result is the same as if the crossfade was not applied. Let n be the number of sub-BIRs, fi the input frame, hi,j the jth sub-IR of the BIR at frame fi and fcr,i,mout the crossfade function. The value of the output frame fi,out is

\[ f_{i, out}^{out} = \sum_{k=0}^{n-1} f_{i, out}^{out} \left( f_{i-k} \ast h_{i-k, k}, f_{i-k} \ast h_{i-k-1, k} \right). \] (9)

The crossfade function is defined in Eq. (10). a(t) is faded-in and b(t) faded-out, l_subIR is the length of the sub-IR and t is time:
in these samples the number of sub-BIRs is 16 (2^4), the input frame as a result of the convolution, thus they are dynamic convolution. Each output frame is twice the size of the whole BIR (2^{14} samples) divided by the number of sub-BIRs, that is, l_subIR = 2^{10}.

VI. CONCLUSION

In this paper we have introduced a new method to compute transfer functions between a violin bridge-pickup signal and radiated sound by playing glissandi, which are recorded simultaneously with a bridge-pickup (excitation) and a microphone in an anechoic chamber (response). The body impulse response is obtained by performing a frame-weighted deconvolution of the pickup and microphone recordings. Using multiple microphones, it is possible to simultaneously measure a set of body impulse responses in various directions around the violin. The impulse responses can be convolved with a source signal obtained with the bridge-pickup in order to obtain realistic violin sounds. Alternatively, the source signal can be generated using a string sound synthesizer. In this case, the source signal can be used here for the violin, which is a general procedure that can be applied to other musical instruments as well. The requirement is that there must be a way to measure an excitation signal together with the response signals.

The multiple impulse responses at different directions make it possible to enhance sound synthesis with spatialization effects. Performances can be recorded while keeping track of the position of the violin, thereby enabling the simulation of the moving effect of the performer. To demonstrate some of these possibilities, simulations of stereo listening and performer motion were presented. Additionally, we proposed an algorithm for dynamic convolution, in which each impulse response is divided into short frames, allowing for real-time cross-fading of impulse responses in a virtually continuous manner.

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\[ cf[a(t), b(t)] = \begin{cases} a t + b(1 - 2t), & 0 < t \leq \frac{l_{\text{subIR}}}{2} \\ a, & \frac{l_{\text{subIR}}}{2} < t \leq l_{\text{subIR}} \end{cases} \]  

Figure 9 shows the process to compute the proposed dynamic convolution. Each output frame is twice the size of the input frame as a result of the convolution, thus they are overlap-added.

Some sound samples can be found in a dedicated web page. In these samples the number of sub-BIRs is 16 (2^4), and the length of the sub-BIRs(l_{\text{subIR}}) is the length of the whole BIR (2^{14} samples) divided by the number of sub-BIRs, that is, l_{\text{subIR}} = 2^{10}.

\[ \sum_{i} \omega_{i} \]