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Autoregressive Parameter Estimation for Equalizing Vibrotactile Systems

ADITYA TIRUMALA BUKKAPATNAM*1, PHILIPPE DEPALLE†1, MARCELO M. WANDERLEY‡1

¹CIRMMT[§], McGILL UNIVERSITY

Musical performance, involving highly physical and cognitive expertise, can benefit from vibrotactile feedback systems offering independent control of amplitude and frequency of vibration over a wide frequency bandwidth. For advanced musical use, it is essential to perform the characterization of the amplifiers and actuators involved, as well as the equalization of their overall frequency response characteristics, a step typically implemented with the help of manually configured parametric equalizers. This paper proposes an autoregressive method that automatically estimates minimumphase filter parameters, which by design, remain stable upon inversion. We demonstrate this method with an example implementation and discuss the degree of equalization achieved using it.

INTRODUCTION

Haptic feedback has become commonplace in digital devices, particularly in the form of vibration, using different types of vibrating actuators (i.e. vibrotactile feedback). However, current systems are predominantly based on Eccentric Rotating Mass (ERM) motors, or more recently, Linear Resonant Actuators (LRAs), which result in systems with a limited frequency response range or exhibiting a coupled control of the amplitude and frequency of vibration. A plethora of such systems are available, including general tools such as the *TECHTILE Toolkit*, which enable prototyping and testing of vibrotactile stimuli [18].

Vibrotactile feedback has also been a popular area of interest for musical applications [20], implemented as either Digital Musical Instruments (DMIs) or general-purpose haptic interfaces [5]. Musical interaction involves highly developed physical and cognitive expertise, potentially benefiting from large bandwidth [8, 3], and decoupled control of the amplitude and frequency of vibration [6, 17]. This is made possible through the selection of specific actuators typically excited by Alternating Current (AC) input signals, and therefore requiring an additional amplifier to drive them [4, 16]. The implementation of some such systems has previously relied on commercial high-fidelity amplifiers such as *Bryston 2B-LP*[3, 4], or *Yamaha P2700*[19], which are bulky (8 kg, 24 kg respectively), and expensive (costing approx. 2500 US\$). Though providing excellent response, such implementations can hinder portability and lead to impractical implementations [14], thus becoming inaccessible to non-laboratory environments.

Vibrotactile actuators and their driving amplifiers have been characterized in terms of their frequency and amplitude responses, frequency resolution [16], and frequency content [4], primarily focusing on the bandwidth of 40-1000 Hz, where humans perceive vibration [24]. In the characterization of the frequency content of such systems, some low-cost (<10 US\$) amplifier boards were found to have a low-fidelity output, introducing harmonic distortion into the signal [4]. Such harmonic distortion

can influence the results of vibrotactile experiments [15, 22], and is commonly quantified in electro-mechanical vibrating systems in the form of Total Harmonic Distortion (THD) calculations [12]. Alongside the amplifiers, the actuators also commonly possess intrinsic non-linearities and frequency characteristics that can influence the accurate reproduction of an intended vibration signal [19].

Interfaces intended for musical applications often require high accuracy, especially where psycho-physical experiments are concerned. This is typically achieved through the characterization of the system and equalization of the vibrotactile output. The characterization of vibrotactile systems also helps provide verified data, which in-turn allows for refinement and better interpretation of experimental results [19]. While some of the existing toolkits address portability and simplicity [18, 9], they are not characterized and do not offer insights on factors like fidelity and frequency response, or lack an equalized output. An implementation of such a toolkit is being addressed more elaborately in the first author's master's thesis [2], where one of the primary objectives is to equip the New Interfaces for Musical Expression (NIME) and other Do-It-Yourself (DIY) music communities with a portable, affordable advanced haptic toolkit that would aid the exploration of more detailed haptic responses, particularly in music related haptic research.

In this paper, we focus on the equalization of the frequency response of the vibrotactile display system, which is often observed to have some resonant structure. This resonant structure has previously been equalized through manually configured parametric equalizers [6, 16, 19], which compensates the linear component in the system's frequency response. This form of equalization has been shown to improve frequency discriminability of the system [16], offer a more accurate reproduction of an intended vibrotactile output [6], and is often key to musical applications displaying simultaneous vibration frequencies [3]. Some alternative systems also attempt to correlate the output and input of a vibrotactile system to produce a "perceptually transparent" rendering of tactile sensations, by encoding the relation between the input control voltage of an ERM and the perceived intensity of vibration, in the form of a function [21]. However, this approach still relies on ERMs, where one and only one input parameter can exist, either perceived magnitude of vibration, or frequency.

Here, we use an auto-regressive system identification technique to estimate the frequency response of the vibrotactile system in the form of a parametric model. The technique models the system in the form of a minimum-phase filter that guarantees stability upon inversion, as opposed to manually configuring filter parameters. In addition to demonstrating the technique, we evaluate the degree of equalization achieved by measuring spectral flatness of the actuator's equalized response.

BACKGROUND

This section discusses issues on actuator selection and measurement, followed by the theory of the autoregressive Yule-Walker method being proposed for the equalization step.

^{*}aditya.tirumala@mail.mcgill.ca

[†]philippe.depalle@mcgill.ca

[‡]marcelo.wanderley@mcgill.ca

[§]Centre for Interdisciplinary Research in Music Media and Technology.

ACTUATORS: SELECTION, AND MEASUREMENT

Various types of actuator technologies are available for producing vibrotactile feedback, each offering pros and cons to different applications. LRAs are optimized to provide high energy output over a very narrow bandwidth, resulting in fixed frequency implementations [4], while ERMs alter the frequency of vibration with a changing of amplitude. However, these technologies do offer lower power consumption [4]. Voice Coil Actuators (VCAs) and Piezoelectric actuators, on the other hand, offer independent control of amplitude and frequency [16] but require an amplifier, while Piezo actuators require additional voltage biasing circuitry

Once an actuator is selected, its frequency characteristics are measured by observing the vibration output for an input sine signal of varying frequency [16] or from transfer function calculations obtained by inputting a logarithmic sine-sweep [4]. The vibration output of the actuator is typically observed with the help of a low-mass accelerometer affixed to the actuator [4, 19], which in turn is also sometimes mounted to a heavy rigid mass to decouple the measurement from structural vibrations [1].

AUTOREGRESSIVE PARAMETER IDENTIFICATION

Kay and Marple [11] surveyed and compared various methods to estimate the power spectral density (PSD) of a process. Among these, deterministic and stochastic processes found in practice are said to be well approximated as white noise filtered by a rational transfer function system, that relates an input sequence $\{x_n\}$ to an output sequence $\{y_n\}$ by the linear difference equation:

$$y_n = \sum_{l=0}^{q} b_l x_{n-l} - \sum_{k=1}^{p} a_k y_{n-k}$$

The transfer function derived from the above equation is the one of a linear filter (ARMA model), consisting of an Autoregressive (AR) part, and a Moving Average (MA) part, which writes:

$$H(z) = \frac{B(z)}{A(z)}$$

where.

$$B(z) = \sum_{l=0}^{q} b_l z^{-l}$$
 (Z-transform of the MA part)
 $A(z) = 1 + \sum_{k=1}^{p} a_k z^{-k}$ (Z-transform of the AR part)

$$A(z) = 1 + \sum_{k=1}^{p} a_k z^{-k}$$
 (Z-transform of the AR part)

For a discrete-time system, evenly sampled at Δt seconds, driven by a zero-mean white noise of variance σ^2 , the Power Spectral Density (PSD) of the ARMA model output is expressed as [11]:

$$P_{ARMA}(f) = P_y(f) = \sigma^2 \Delta t \left| \frac{B(e^{j2\pi f \Delta t})}{A(e^{j2\pi f \Delta t})} \right|^2$$

An ARMA process y_n is entirely determined by parameters a_k , b_l and σ^2 , where $a_0 = b_0 = 1$. When all $b_l = 0$ but $b_0 (= 1)$, the model reduces to an all-pole (AR)system of order p, which PSD simplifies as:

$$P_{AR}(f) = \frac{\sigma^2 \Delta t}{\left|1 + \sum_{k=1}^{p} a_k e^{-j2\pi f k \Delta t}\right|^2}$$
 (1)

The estimation of the PSD consists then in estimating $\{a_1, a_2, \cdots, a_p\}$ and σ^2 [11].

AUTOREGRESSIVE YULE-WALKER ESTIMATION

The Yule-Walker equations [27, 25] establish a relation between the AR parameters $\{a_1,a_2,\cdots,a_p\}$, σ^2 and the autocorrelation sequence of y_n . This relation is as follows [11]:

$$R_{yy}(k) = \begin{cases} -\sum_{l=1}^{p} a_l R_{yy}(k-l), & \forall k > 0, \\ -\sum_{l=1}^{p} a_l R_{yy}(-l) + \sigma^2, & \text{for } k = 0. \end{cases}$$
 (2)

By selecting p equations for k > 0 in Eq. (2), we can solve for $\{a_1,a_2,\cdots,a_p\}$, and then compute σ^2 for k=0. This can be expressed in matrix form as shown in Eq. (3), whose solution is found with p+1 estimated autocorrelation lags where $R_{yy}(-m)=R_{yy}^*(m)$ [11].

$$\begin{bmatrix} R_{yy}(0) & R_{yy}(-1) & \cdots & R_{yy}(-p) \\ R_{yy}(1) & R_{yy}(0) & \cdots & R_{yy}(-p+1) \\ \vdots & & & \vdots \\ R_{yy}(p & R_{yy}(p-1) & \cdots & R_{yy}(0) \end{bmatrix} \begin{bmatrix} 1 \\ a_1 \\ \vdots \\ a_p \end{bmatrix} = \begin{bmatrix} \sigma^2 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$
(3)

The linear system in Eq. (3) can be solved efficiently by using the Levinson-Durbin algorithm, which recursively computes the parameter sets $\{a_{1m}, a_{2m}, \cdots, a_{pm}\}$, for $m=\{1, 2, \cdots, p\}$ where p is the order of the AR process [11].

Parameters $\{a_{11},\cdots,a_{pp}\}$ can be interpreted as reflection coefficients, namely $\{K_1,K_2,\cdots,K_p\}$. It appears that the necessary and sufficient stability condition of the AR model, which requires all the poles of A(z) to lie within the unit circle $z=e^{j2\pi f\Delta t}$, is simply that all $|K_k|\leq 1$, for $k=1,2,\cdots,p$. The resulting AR model is proven to be stable and minimum-phase, i.e. whose inverse is also stable [23] 1. Notice that the inverse filter A(z) is a Finite Impulse Response (FIR) filter.

IMPLEMENTATION

To demonstrate this method, we measure a voice coil actuator, characterize its frequency response to design its inverse filter, and then investigate the same actuator's response, when the inverse filter is applied. We now present the apparatus and methodology used to perform these measurements.

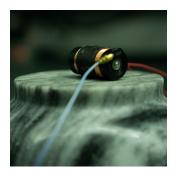
MEASUREMENT APPARATUS

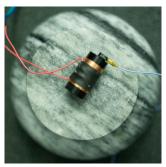
The VCA chosen for this demonstration is a Haptuator Redesigned TL-002-14R 2 by Tactile Labs with impedance = 6 Ω , bandwidth = 50-500 Hz, and weighing 11 g. A lightweight (0.2 g) single-axis accelerometer, PCB Piezotronics 352C23, is affixed to the shell of the Haptuator, using an instant glue³ that is recommended by the manufacturer for adhesive mounting. A PCB Piezotronics 482B11 line conditioner is used to supply the biasing current required for the accelerometer's operation, and for amplifying its output. The accelerometer's signals are recorded in MATLAB by a National Instruments USB-4431 data acquisition unit. The actuator is also fixed to a rigid mass (stone pestle) weighing approx. 897 g, placed on a carpeted floor. The accelerometer cable is tucked under the pestle to create a relief, and the axis of the accelerometer is aligned with the axis of the VCA's vibrating element. This apparatus is illustrated in Figure 1.

¹Link: CCRMA, J. O. Smith: Minimum Phase Definition.

²Link: TactileLabs, HaptuatorRedesignSpec_v1.0.pdf

³Loctite 454, debonded with Acetone





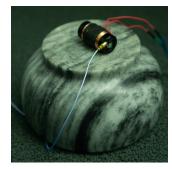


Figure 1: Haptuator Redesigned, mounted onto a stone pestle, with accelerometer affixed to its shell.

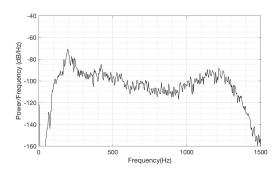
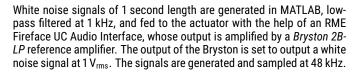


Figure 2: Welch PSD estimate averaged over 5 measurements. Welch estimates are used for smoother visualization.



ACTUATOR MEASUREMENTS

The recordings of the actuator being driven with white noise are repeated 5 times. The Power Spectral Densities (PSD) of each of the 5 recordings are computed and then averaged in order to reduce the variance of the PSD estimator, mitigating the effects of noise on the parameter estimation process. The averaged PSD is shown in Figure 2. The magnitude spectrum is obtained as the square root of the PSD and then multiplied with a random phase between $[0,2\pi]$ to make it possible to perform an Inverse Fourier transform operation. The result is then Inverse Fourier transformed into a time series vector, whose autocorrelation function is used to compute the autoregressive Yule-Walker parametric model.

PARAMETRIC MODEL

The bandwidth requirement of vibrotactile systems has been known to extend up to 1 kHz [24, 6] and sometimes higher among the hearing-impaired demographic [26]. In this case, we aim to assess the actuator up to 1.5 kHz, and as per the data-sheet² of the Haptuator Redesigned, driving the actuator above its rated 500 Hz isn't expected to damage the actuator, but rather, render the signal audible. In order to ensure that the Yule-Walker estimation is focused within this bandwidth, we down-sample the measured data to a sample rate of 3 kHz after applying a 10th order Type-I Chebyshev filter as an anti-aliasing stage.

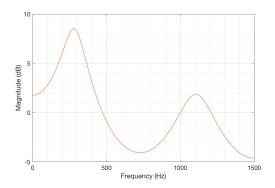


Figure 3: Autoregressive Yule-Walker estimate using 4-poles

One of the noted remarks regarding the use of autoregressive parameter estimation methods is that the number of parameters is user-defined. As a result, while the accuracy of the estimate tends to increase with an increase in the number of parameters, it becomes hard to discern how many parameters are sufficient to estimate the model adequately [11]. In this case, a minimum of 4 poles was required to model the two resonant features, observed in the actuator's PSD estimate, shown in Figure 2.

Figure 3 shows the model obtained with a 4 pole estimate, and Figure 4 shows the accuracy in modelling the overall gain characteristics of the actuator's frequency characteristics, with increasing AR order.

RESULTS

In this section, we present the results of the modelling and inverse filtering process. The model estimated by a 12-pole Yule-Walker analysis is selected heuristically, and its inverse model is derived. The frequency characteristics of the resulting inverse filter are shown in Figure 5.

The measurements are then repeated by applying the inverse filter to pre-emphasize the white noise signal, before feeding it to the actuator. It should be noted that the parametric model was designed for a sampling rate of 3 kHz. For this reason, the white noise signal is down-sampled to 3 kHz before applying the inverse filter, and then up-sampled back to 48 kHz, to be able to operate at audio rate. Figure 6 shows the Welch PSD estimate of the Haptuator, measured with the inverse filter applied, and averaged over 5 measurements.

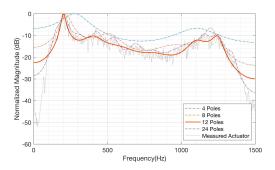


Figure 4: Estimated models show a visible improvement in the accuracy of modelling of the overall gain structure of the response, for increasing pole orders. The magnitude responses of the estimated models are normalized to their peak values. The magnitude response of the actuator, derived from the Welch PSD estimate, shown in Figure 2, and normalized to its peak value. The model estimated with 12 poles is the selected model.

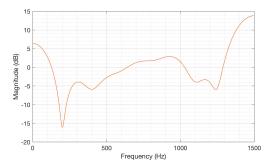


Figure 5: Inverse filter derived from the 12-pole Yule-Walker estimate, chosen heuristically.

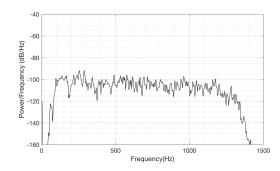


Figure 6: Welch PSD estimate of the actuator measurement, with inverse filter applied.

ESTIMATION OF SPECTRAL FLATNESS

The most elementary assessment of spectral flatness of the inverse filtering process can be done visually, as shown in Figure 7.

The degree of spectral flatness can also be estimated as the ratio of geometric mean to the arithmetic mean of the PSD estimate [7]. Additionally, the reflection coefficients K_p of the estimated model decay to a minimum over a certain number of poles. Increasing the order of the model beyond this point has a minimal effect on the overall estimation and would be regarded as over-fitting poles to the model. The spectral flatness estimate in Figure 8a and the decay of reflection coefficients 8b suggest a minimum of 8 poles and a maximum order of 32 poles, to model this device.

CONCLUSION

The Autoregressive Yule-Walker approach is used as a means of equalizing the response of a VCA-based vibrotactile system. In this kind of equalization method, there is a need to operate at a sampling rate much lower than audio rates, in order to restrict the estimation of parameters, to within the bandwidth of vibrotactile perception. Traditional Digital to Audio Converters (DACs) are designed to operate at audio sampling

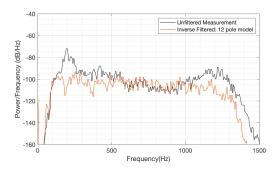
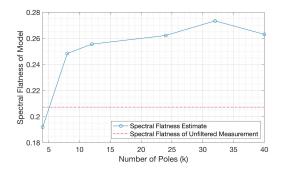
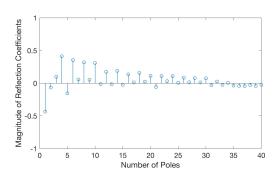


Figure 7: PSD of the inverse filtered result, superimposed on the original PSD estimate. PSD estimates are smoothed with a 3rd order median filter.



(a) Increase in Spectral Flatness Estimate of Inverse Filtered model, increasing up to 32 poles, and then showing a decrease.



(b) Reflection Coefficients decay to zero after 32 poles.

Figure 8: Evaluating the accuracy of fit.

rates (44.1 kHz, 48 kHz, or higher). We are often required to operate at audio rates to generate vibrotactile feedback, where amplitude and frequency can be independently controlled and generated through computer programs. There is a fairly extreme sample rate conversion stage (48 kHz to 3 kHz or less), which begins to play a role in the implementation of parameter estimation methods for equalizing vibrotactile systems.

The use of autoregressive parameter estimation to equalize vibrotactile systems results in the question of how many parameters are required to model the system adequately. In this case, we attempt to use a spectral flattening estimate to analyze the degree of flatness in the inverted frequency response. This is also compared against the decaying magnitudes of reflection coefficients yielded by the Yule-Walker estimation process.

FUTURE WORK

The technique discussed in this paper predominantly accounts for the vibration of the the actuator in its primary axis of vibration. An interesting next step would be to analyze the vibration of actuators using triaxial accelerometers.

The Yule-Walker method is but one among a collection of parameter identification methods, each having pros and cons of their own. In real-

time applications, a higher order model would introduce latency into the process, which may be detrimental to scenarios like musical performance, where latency is ideally kept to a minimum [19], typically not exceeding 25 ms for tactile feedback [10]. Therefore, techniques that can achieve a similar or higher accuracy of equalization with a lower order would be worth investigating.

The method demonstrated in this paper assumes a linear system. Therefore, we only equalize the linear component in the actuator's response. Vibrating transducers can exhibit non-linear operation [12], which might be addressed by methods that can also account for non-linearity [13].

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REFERENCES

- C. Beards. The Vibration of Structures with One Degree of Freedom. In Structural Vibration: Analysis and Damping, chapter 2, pages 10–82. Butterworth-Heinemann, Elsevier Ltd., Oxford, MA, 1996.
- [2] A. T. Bukkapatnam. A Pre-Characterized Vibrotactile Toolkit (In Progress). Master's thesis, McGill University, 2019.
- [3] D. C. Egloff, M. M. Wanderley, and I. Frissen. Haptic Display of Melodic Intervals for Musical Applications. In 2018 IEEE Haptics Symposium (HAPTICS), pages 284–289, San Francisco, CA, 2018. IEEE.
- [4] M. Giordano. Vibrotactile Feedback and Stimulation in Music Performance. Phd thesis, McGill University, 2016.
- [5] M. Giordano, J. Sullivan, and M. M. Wanderley. Design of Vibrotactile Feedback and Stimulation for Music Performance. In S. Papetti and C. Saitis, editors, *Musical Haptics*, chapter 10, pages 193–214. Springer, Cham, Switzerland, 2018.
- [6] M. Giordano and M. M. Wanderley. Perceptual and Technological Issues in the Design of Vibrotactile-augmented Interfaces for Music Technology and Media. In I. Oakley and S. Brewster, editors, Haptic and Audio Interaction Design. 8th International Workshop, HAID 2013, Daejeon Korea., volume LNCS 7989, pages 89–98. Springer, Berlin, 2013.
- [7] A. H. Gray and J. D. Markel. A Spectral-Flatness Measure. IEEE Transactions on Acoustics, Speech, and Signal Processing, ASSP-22(3):207-217, 1974.
- [8] I. Hattwick, I. Franco, and M. M. Wanderley. The Vibropixels: A Scalable Wireless Tactile Display System. In S. Yamamoto, editor, Human Interface and the Management of Information: Information, Knowledge and Interaction Design. HIMI 2017, volume LNCS, vol, pages 517-528, Vancouver, Canada, 2017. Springer, Cham.
- [9] A. Israr, S. Zhao, K. Mcintosh, Z. Schwemler, A. Fritz, J. Mars, J. Bedford, C. Frisson, I. Huerta, and M. Kosek. Stereohaptics: A Haptic Interaction Toolkit for Tangible Virtual Experiences. In ACM SIGGRAPH 2016 Studio, Jul 2016, Anaheim, United States, 2016.
- [10] T. J. Kaaresoja. Latency Guidelines for Touchscreen Virtual Button Feedback. Phd thesis, University of Glasgow, 2015.

- [11] S. M. Kay and S. L. Marple. Spectrum Analysis—A Modern Perspective. Proceedings of the IEEE, 69(11):1380–1419, 1981.
- [12] W. Klippel. Tutorial: Loudspeaker Nonlinearities Causes, Parameters, Symptoms. AES: Journal of the Audio Engineering Society, 54(10):907-939, 2006.
- [13] L. Ljung. System Identification Theory for User. Prentice-Hall Inc., Englewood Cliffs, New Jersey, 1987.
- [14] K. E. MacLean. Foundations of Transparency in Tactile Information Design. IEEE Transactions on Haptics, 1(2), 2008.
- [15] S. Maeda and M. J. Griffin. A Comparison of Vibrotactile Thresholds on the Finger Obtained with Different Equipment. *Ergonomics*, 37(8):1391–1406, 1994.
- [16] M. T. Marshall. Physical Interface Design for Digital Musical Instruments. Phd thesis, McGill University, 2008.
- [17] M. T. Marshall and M. M. Wanderley. Vibrotactile Feedback in Digital Musical Instruments. In *Proceedings of NIME 2006*, pages 226–229, 2006.
- [18] K. Minamizawa, Y. Kakehi, M. Nakatani, S. Mihara, and S. Tachi. TECHTILE Toolkit: A Prototyping Tool for Design and Education of Haptic Media. In *Proceedings of the 2012 Virtual Reality Interna*tional Conference (VRIC '12), Laval, France, 2012. ACM Press.
- [19] S. Papetti, M. Fröhlich, F. Fontana, S. Schiesser, and F. Avanzini. Implementation and Characterization of Vibrotactile Interfaces. In S. Papetti and C. Saitis, editors, *Musical Haptics*, chapter 13, pages 257–282. Springer, Cham, Switzerland, 2018.

- [20] J. Rovan and V. Hayward. Typology of Tactile Sounds and their Synthesis in Gesture-Driven Computer Music Performance. In *Trends in Gestural Control of Music*, pages 297–320. IRCAM, 2000.
- [21] J. Ryu, J. Jung, S. Kim, and S. Choi. Perceptually transparent vibration rendering using a vibration motor for haptic interaction. In RO-MAN2007 The 16th IEEE International Symposium on Robot and Human Interactive Communication, pages 310–315, Jeju, Korea, 2007. IEEE.
- [22] C. Salisbury, R. B. Gillespie, H. Z. Tan, F. Barbagli, and J. K. Salisbury. What You Can't Feel Won't Hurt You: Evaluating Haptic Hardware Using a Haptic Contrast Sensitivity Function. *IEEE Transactions on Haptics*, 4(2):134–146, 2011.
- [23] J. O. Smith. Introduction to Digital Filters with Audio Applications. online book, 2007 edition, 2007.
- [24] R. T. Verrillo. Vibration Sensation in Humans. Music Perception, 9(3):281-302, 1992.
- [25] G. Walker. On Periodicity in Series of Related Terms. Proceedings of the Royal Society. Series A, 131(818):518–532, 1931.
- [26] L. Wyse, S. Nanayakkara, P. Seekings, S. H. Ong, and E. A. Taylor. Palm-area sensitivity to vibrotactile stimuli above 1 kHz. In NIME'12 Proceedings of the International Conference on New Interfaces for Musical Expression, Michigan, 2012.
- [27] G. U. Yule. On a Method of Investigating Periodicities in Disturbed Series, with Special Reference to Wolfer's Sunspot Numbers. *Philo-sophical Transactions of the Royal Society of London. Series A*, 226(642):267–298, 1927.