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# Kernel interpolation of acoustic transfer functions with adaptive kernel for directed and residual reverberations

### Background

- Sound wave propagation is unpredictable.  $\Rightarrow$  Physical phenomena such as reflection and d
- Acoustic transfer function (ATF):  $\Rightarrow$  Describes space's influence on sound waves
- **Region-to-region**:  $\Rightarrow$  Interpolate ATF continuously for source/receive
- Sound field analysis:  $\Rightarrow$  Deeply related problem we can draw solutions

### **Problem statement**

Objective: interpolate ATF values from measuren



- ATF components  $h(\mathbf{r}|\mathbf{s}) = h_{\mathrm{R}}(\mathbf{r}|\mathbf{s}) + h_{\mathrm{D}}(\mathbf{r}|\mathbf{s})$ .
- Direct is known:  $h_{\mathrm{D}}(\mathbf{r}|\mathbf{s},k) = \frac{e^{\mathrm{i}k\|\mathbf{r}-\mathbf{s}\|}}{4\pi\|\mathbf{r}-\mathbf{s}\|}.$
- Reverberant:  $h_{\rm R}(\mathbf{r}|\mathbf{s},k)$  is more involved.  $\Rightarrow$  [Ribeiro+, 2020]: physical constraint in <u>kernel</u> regression with ATF kernel.
- $\Rightarrow$  [Ribeiro+, 2022]: directionality improved perfor Generalized representation: Herglotz wave funct

$$\begin{split} h_{\mathrm{R}}(\mathbf{r}|\mathbf{s}) &= \mathcal{T}(\tilde{h}_{\mathrm{R}};\mathbf{r}|\mathbf{s}) \\ \hline \text{Plane wave} \\ \text{superposition} &\coloneqq = \int_{\mathbb{S}^2 \times \mathbb{S}^2} \mathrm{e}^{\mathrm{i}k(\hat{\mathbf{r}}\cdot\mathbf{r}+\hat{\mathbf{s}}\cdot\mathbf{s})} \tilde{h}_{\mathrm{R}}(\hat{\mathbf{r}},\hat{\mathbf{s}}) \mathrm{d}\hat{\mathbf{r}} \mathrm{d}\hat{\mathbf{s}} \end{split}$$

Representation that guarantees physics of the pr  $\Rightarrow$ Kernel function that learns weight function as o model.

$$\kappa(\mathbf{r}|\mathbf{s},\mathbf{r}'|\mathbf{s}') = \mathcal{T}\left(w(\hat{\mathbf{r}},\hat{\mathbf{s}})\frac{\mathrm{e}^{-\mathrm{i}k(\hat{\mathbf{r}}\cdot\mathbf{r}'+\hat{\mathbf{s}}\cdot\mathbf{s}')} + \mathrm{e}^{-\mathrm{i}k(\hat{\mathbf{r}}\cdot\mathbf{s}'+\hat{\mathbf{s}}\cdot\mathbf{r}')}}{2}\right)$$

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diffraction.	Previous kernels can be expressed [Ribeiro+, 2020]: kernel function equ ⇒ Satisfies basic physical propertie [Ribeiro+, 2022]: <u>sunken sphere</u> w ⇒ Introduces directionality to the est ⇒ Physical model limited, assumes gain on the sides. Data models for the weights are infl	equiva rties. <u>e</u> weig estimation
s from	<ul> <li>We propose an <u>adaptive kernel</u> the weight as a more general da</li> </ul>	
ments.	• Weighting represents directed a sum $w = w_{dir} + w_{res}$ .	
	$\begin{split} & \underbrace{\text{Directed weight } \mathcal{W}_{\text{dir}}}_{\bullet} \\ & \bullet \text{ High amplitudes on sparse set of directions} \\ & \bullet \text{ Strong directionality.} \\ & \bullet \text{ Combination of von Mises-Fisher distributions.} \\ & w_{\text{dir}}(\hat{\mathbf{r}}, \hat{\mathbf{s}}) = \varphi_{\text{dir}}(\hat{\mathbf{r}})\varphi_{\text{dir}}(\hat{\mathbf{s}}), \\ & \varphi_{\text{dir}}(\hat{\mathbf{v}}) = \sum_{d=1}^{D} \alpha_d \frac{\mathrm{e}^{\beta_d \hat{\mathbf{v}} \cdot \hat{\mathbf{v}}_d}}{4\pi C(\beta_d)}, \end{split}$	Res • Lo di • Ur • Re w • In
ridge mance. tion.	$\ \boldsymbol{\alpha}\ _{1} = 1,$ $C(\beta_{d}) = \begin{cases} \frac{\sinh(\beta_{d})}{\beta_{d}}, & \beta_{d} \neq 0\\ 1, & \beta_{d} = 0 \end{cases}.$	
	<ul> <li>The resulting adaptive kernel will be the kernels.</li> <li>Kernel optimized as to minimize leave loss.</li> </ul>	
<b>roblem.</b> <b>data</b> $(') = ; \mathbf{r}   \mathbf{s}$	<ul> <li>Model parameters β and θ optimized</li> <li>Parameter α optimized with reduced restrictions are upheld.</li> <li>The adaptive kernel learns particular without compromising model dynamic</li> </ul>	

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ow amplitude on a dense set of irections

npredictable behavior.

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itegral operator approximated umerically.



the superposition of the

e-one-out cross validation

using gradient descent. gradient descent to guarantee

ties of the ATF in question CS.

- -Reverberation time:  $T_{60} = 0.45$  s. -Radius of both regions: 0.2 m. -Centers of  $\Omega_{S,R}$ :  $\pm [0.65, 0.8, 0.48]^T$ .
- kernels.

### Normalized mean square error (NMSE)







x (m)

NMSE = -13.1 dB

### Experiments

Simulations with the image source method. -Room dimensions:  $3.2 \text{ m} \times 4.0 \text{ m} \times 2.7 \text{ m}$ . **Proposed** compared to **uniform** and **sunken sphere** 

#### **Reconstruction of ATF (1150Hz)**

