



Acoustics as intangible cultural heritage

Measurement, archiving and reproduction

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Context

Early Music Research

Source



Performer

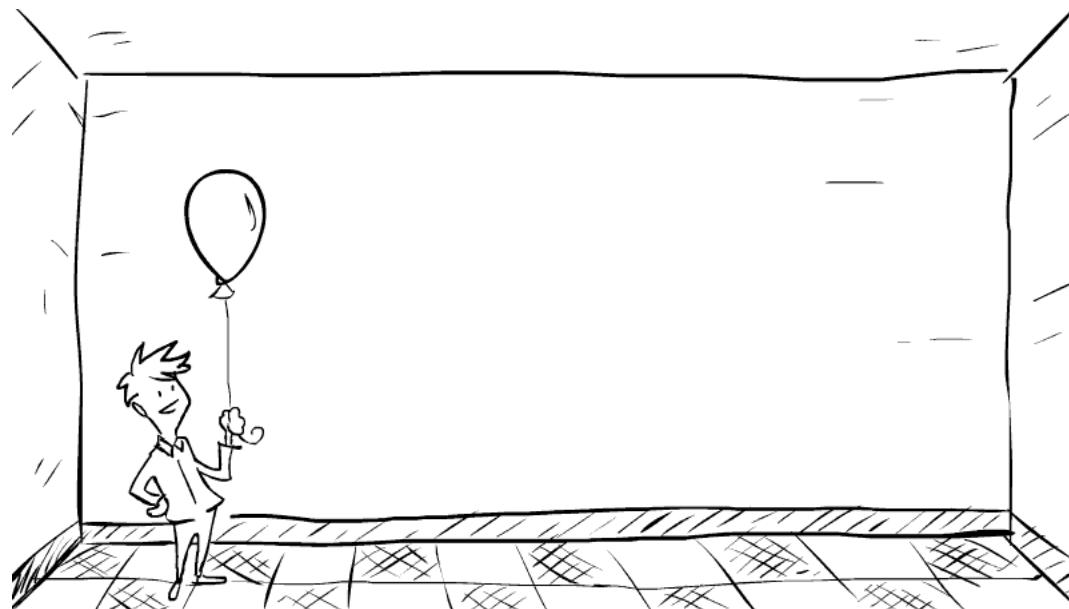


Environment



Problem statement

- **What is “acoustics” of a space?**
 - Acoustics = propagation and scattering of sound waves within a space, independent of source signal



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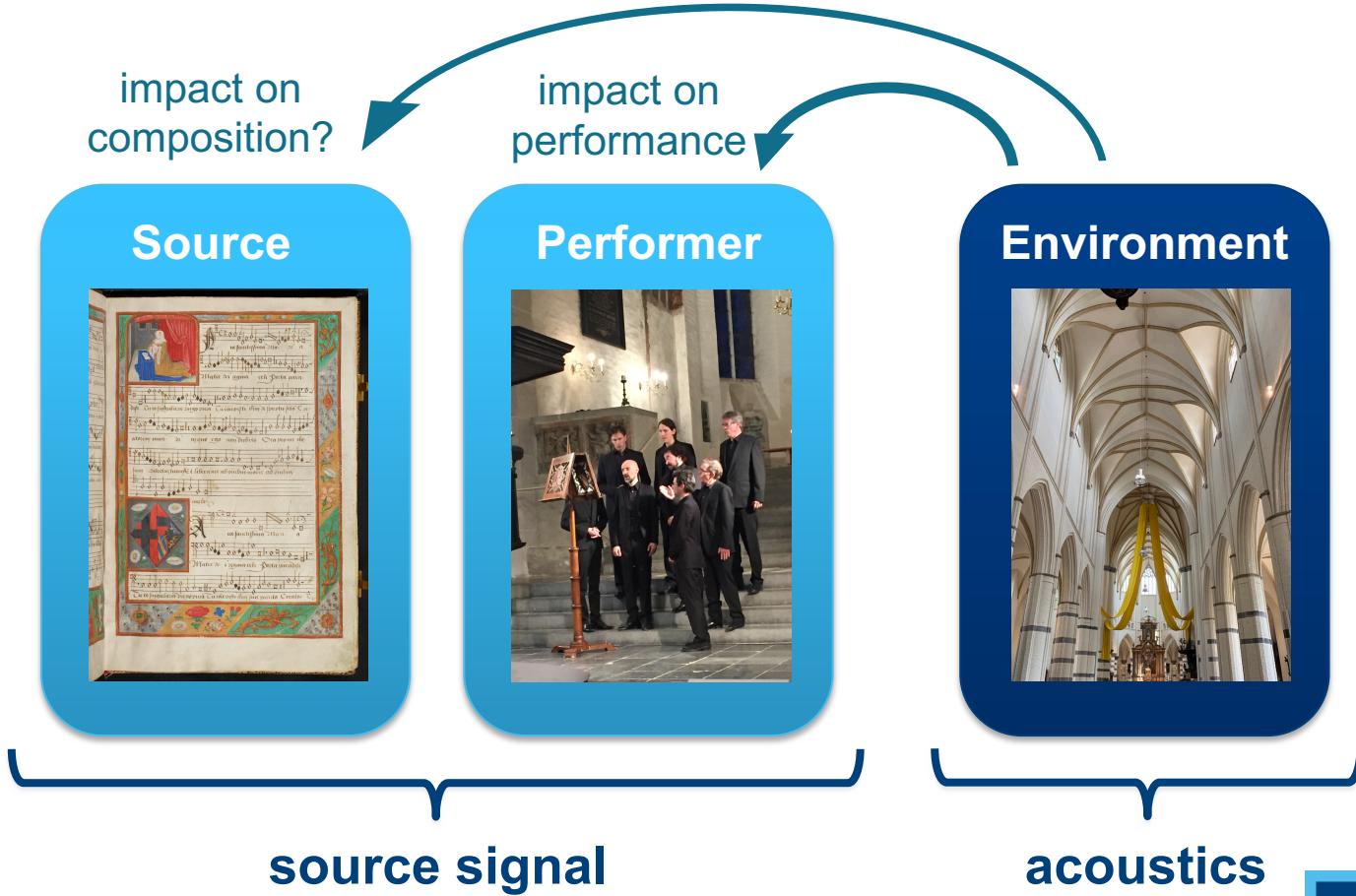


source signal

acoustics

Problem statement

- What is “acoustics” of a space?



Problem statement

- **Research question: how can we archive the acoustics of a culturally valuable space and “reuse” it later?**



- **Reuse** = auditory experience as if one is listening to any performance in the acoustics of the original space

Role of AI

2022



2122



measurement



Data-
driven
modeling

archiving



Structured
data
approximation

reproduction



Perceptual
assessment
data analysis

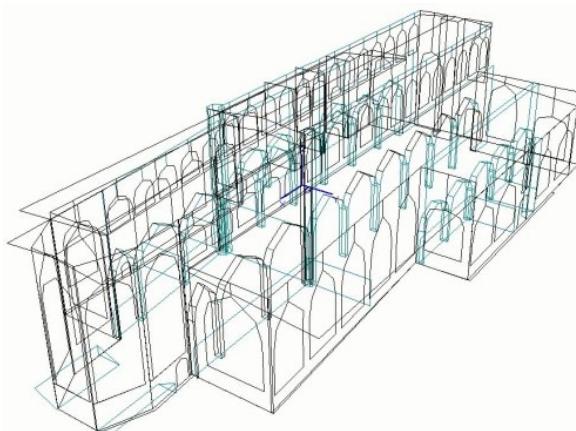
Measurement: Data-driven modeling

- How to measure the acoustics of a space?



- (1) measure geometry of space and objects inside
- (2) determine sound scattering properties of walls and objects

= “descriptive” archaeoacoustics



- tedious & time-consuming
- reproduction?

Measurement: Data-driven modeling

- **Data-driven modeling:** “systems” view on acoustics



source

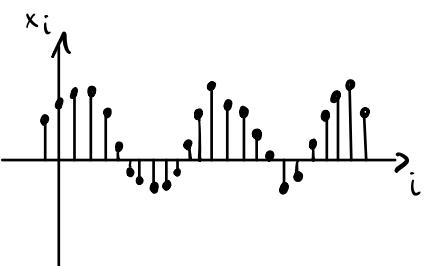
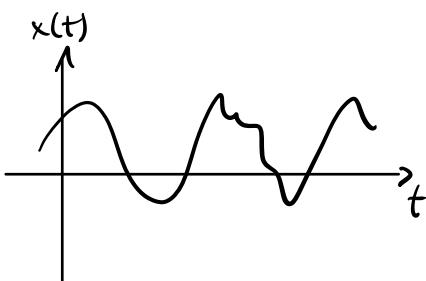
acoustics

observed sound

INPUT

SYSTEM

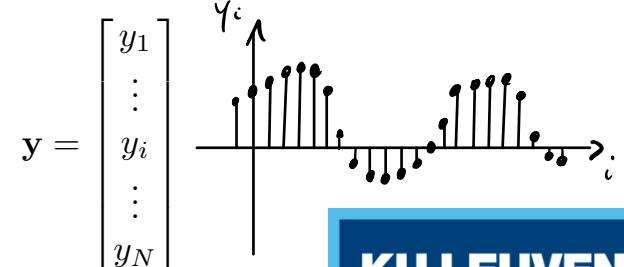
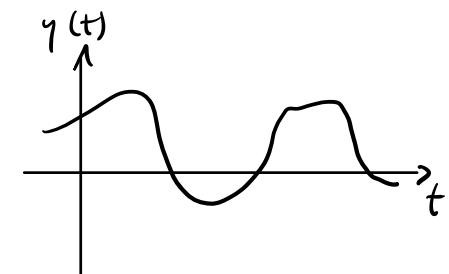
OUTPUT



$$\mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_i \\ \vdots \\ x_N \end{bmatrix}$$

continuous time:
signals = **functions**

discrete time:
signals = **vectors**



$$\mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_i \\ \vdots \\ y_N \end{bmatrix}$$

Measurement: Data-driven modeling

- Fundamental result for linear time-invariant systems:



- impulse response contains all information on system
- output to arbitrary input can be computed by convolution



Measurement: Data-driven modeling

- Fundamental result for linear time-invariant systems:

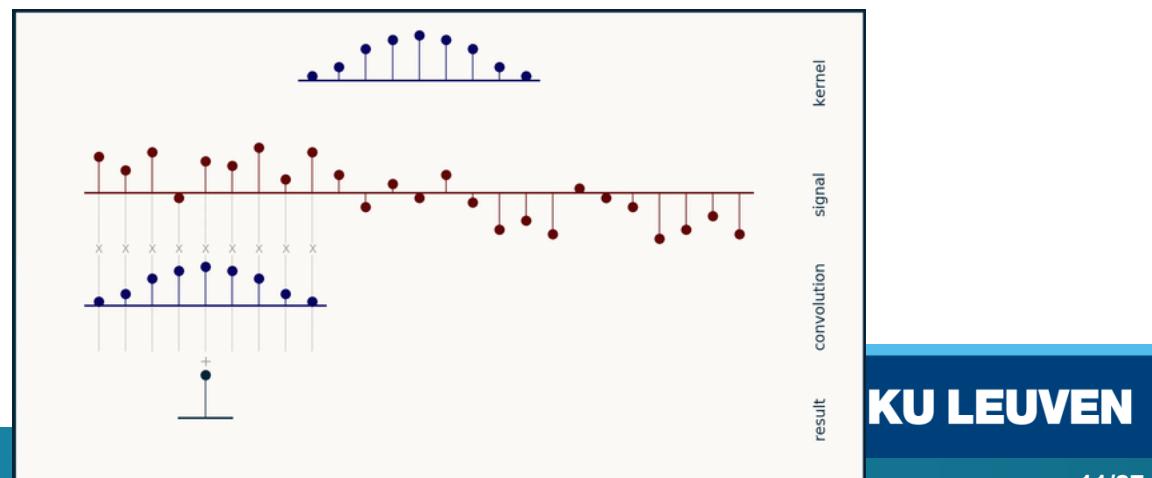


- impulse response contains all information on system
- output to arbitrary input can be computed by convolution

$$\mathbf{h} = [h_1 \dots h_L]^T$$

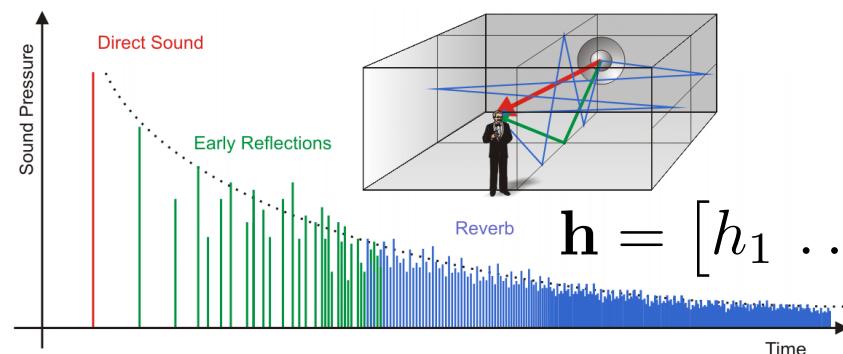
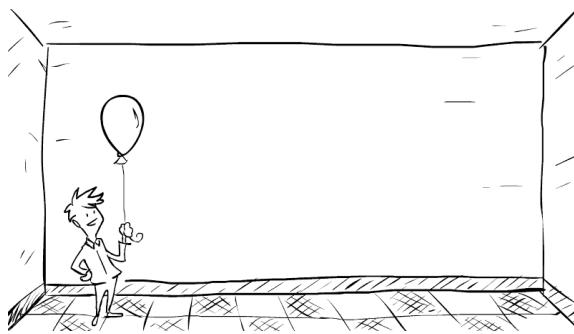
$$\mathbf{x} = [x_1 \dots x_N]^T$$

$$\mathbf{y} = \mathbf{h} * \mathbf{x}$$



Measurement: Data-driven modeling

- **Geometric interpretation of room impulse response**

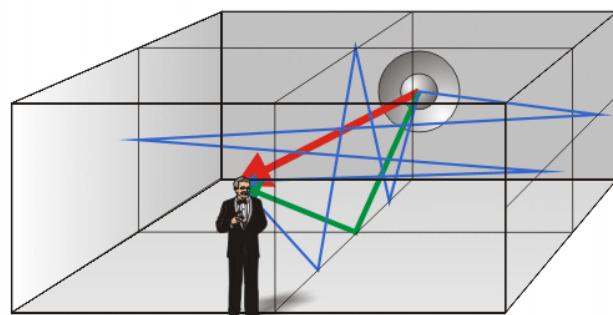


- peaks in impulse response represent **acoustic reflections**
- nice properties:
 - **purely data-driven model** for room acoustics
 - simple measurement protocol
- problems:
 - dependent on source and observer positions
 - **no information on direction of arrival of reflections**
 - very long (vectors with up to 10^4 elements)

Measurement: Data-driven modeling

- **Spatial room impulse response**
 - include **direction of arrival information** (azimuth and elevation angles θ, ϕ) to each peak in impulse response
 - **spatial decomposition method [1]** allows to compute spatial impulse response from set of 6 measured impulse responses

$$\begin{bmatrix} h_1^{(1)} \\ \vdots \\ h_L^{(1)} \end{bmatrix}, \begin{bmatrix} h_1^{(2)} \\ \vdots \\ h_L^{(2)} \end{bmatrix}, \dots, \begin{bmatrix} h_1^{(6)} \\ \vdots \\ h_L^{(6)} \end{bmatrix} \rightarrow \mathbf{h}_S = \begin{bmatrix} h_1 & \theta_1 & \phi_1 \\ \vdots & \vdots & \vdots \\ h_L & \theta_L & \phi_L \end{bmatrix}$$



[1] Tervo *et al.*, “Spatial Decomposition Method for Room Impulse Responses,” *J. Audio Eng. Soc.*, 61(1/2), 2013.

Measurement: Data-driven modeling

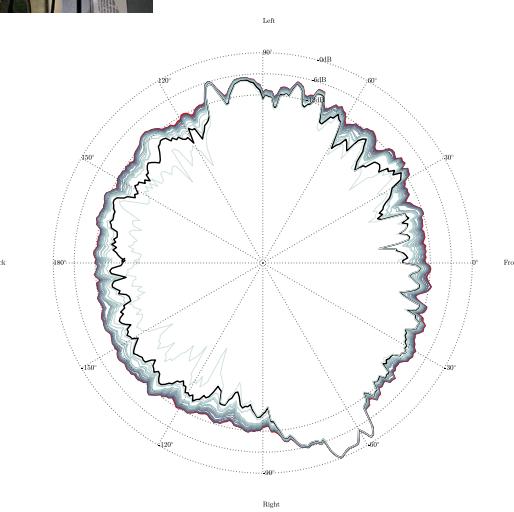
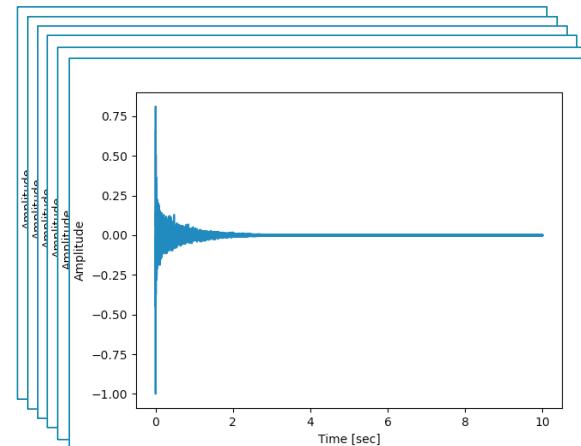
- Case study: Nassau Chapel @ KBR Brussel



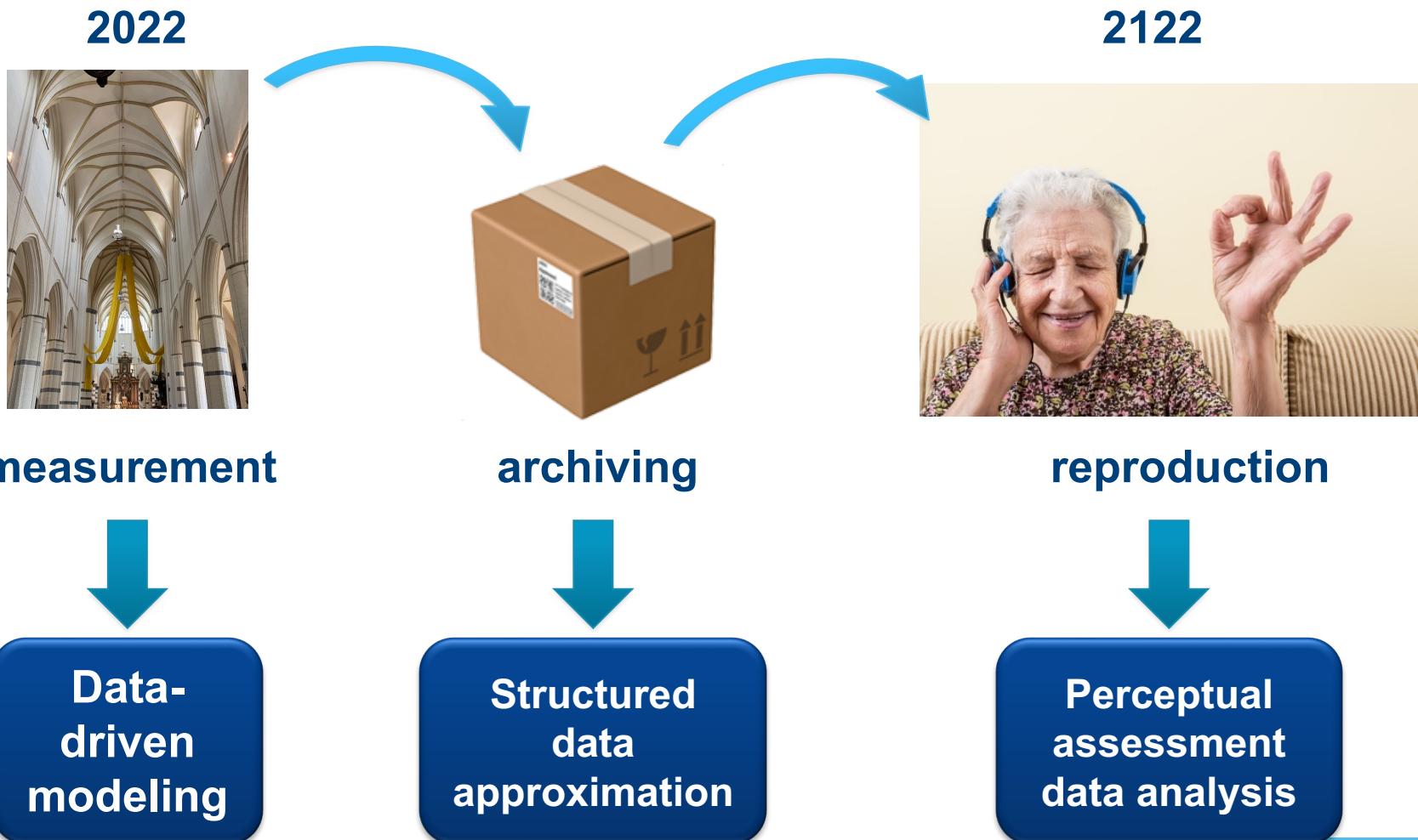
Genelec 8030C



G.R.A.S. VIP 50VI-1



Role of AI



Archiving: Structured data approximation

- **Some simple math...**

- impulse response contains $L \sim 10^4$ samples $\sim 10^5$ bits
- 1 spatial impulse response = 6 impulse responses $\sim 10^6$ bits
- plenacoustic sampling theory [2]: accurate sound field reconstruction requires spatial resolution of ~ 10 cm
 $\sim (100)^3$ source positions x $(100)^3$ observer positions
 $\sim 10^{12}$ spatial impulse responses $\sim 10^{18}$ bits **~ 100 petabyte**



[2] Ajdler *et al.*, “The Plenacoustic Function and Its Sampling,” *IEEE Trans. Sig. Process.*, 54(10), 2006.

Archiving: Structured data approximation

- **Relieving curse of dimensionality using tensorization**
 - impulse response vector (= 1-D array) of length $L = 10^4$ samples can be **reshaped** into N -D array, e.g.
 - matrix (= 2-D array) of dimensions $\sqrt{L} \times \sqrt{L} = 100 \times 100$
 - tensor (= N -D array) of dimensions $\sqrt[N]{L} \times \sqrt[N]{L} \times \dots \times \sqrt[N]{L}$

example: 2-D tensorization

$$\begin{bmatrix} h_1 \\ h_2 \\ h_3 \\ \vdots \\ h_L \end{bmatrix} \xrightarrow{\text{Tensorize}} \text{blue square}$$

3-D tensorization

$$\begin{bmatrix} h_1 \\ h_2 \\ h_3 \\ \vdots \\ h_L \end{bmatrix} \xrightarrow{\text{Tensorize}} \text{blue cube}$$

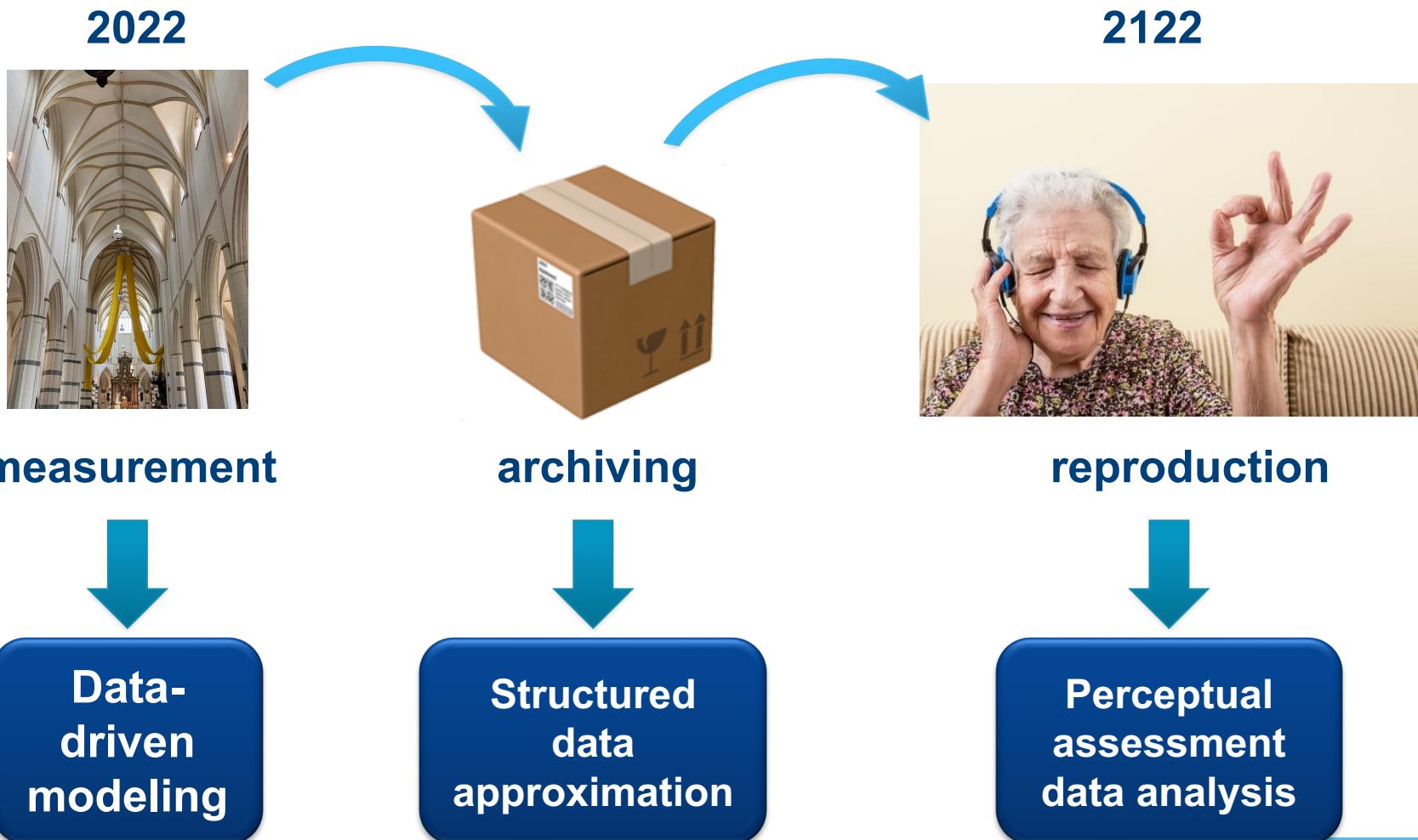
Archiving: Structured data approximation

- **Relieving curse of dimensionality using tensorization**
 - impulse response vector (= 1-D array) of length $L = 10^4$ samples can be **reshaped** into N -D array
 - N -D array can be **approximated** as sum of R rank-1 terms (canonical polyadic decomposition)
 - low-rank approximation makes sense as impulse response is linear combination of damped sinusoids (“**room modes**”) [3]
 - up to **80-90% reduction** of required archiving space [3]



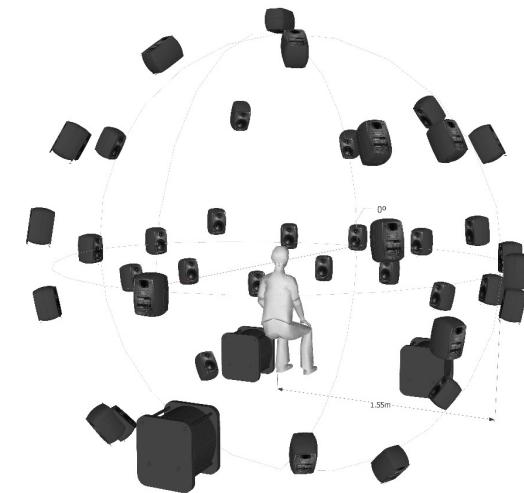
[3] Jälmby, Elvander & vW, “Low-Rank Tensor Modeling of Room Impulse Responses,” *Proc. EUSIPCO*, 2021.

Role of AI



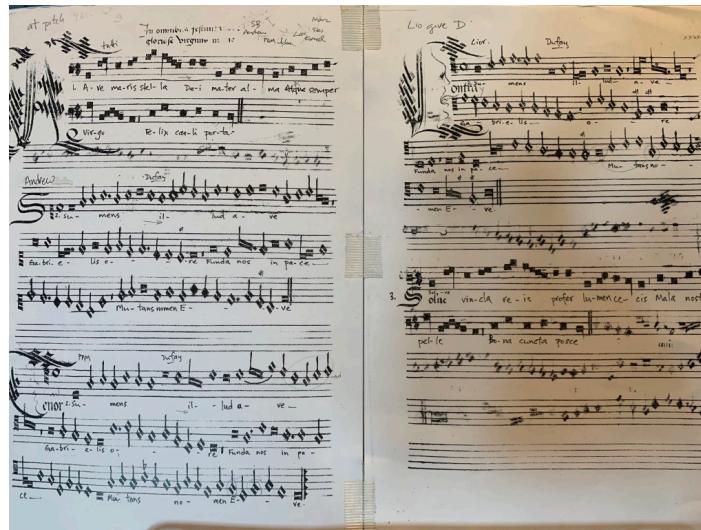
Reproduction: Perceptual assessment data analysis

- **Spatial room impulse response auralization**
 - 3-D loudspeaker array in (semi-)anechoic listening room:
Alamire Interactive Lab @ Library of Voices
 - convolution of source signal with components of spatial impulse response where reflection angle \approx loudspeaker angle



Reproduction: Perceptual assessment data analysis

- **Perceptual assessment of room acoustics**
 - listening experiments with naive or **expert listeners**
- **Materials:** 6 virtual acoustic spaces, 4 singers, 2 pieces
 - 4 male singers from Cappella Pratensis
 - one plainchant + one polyphonic piece



Reproduction: Perceptual assessment data analysis

- **Methodology:** Flash Profile rapid sensory analysis
 - Originally developed in frame of **food tasting** experiments [4]
 - Validated for perceptual modeling of virtual acoustics **by listening** (e.g. auralization of concert halls, car cabins) [5]
 - Evaluated here for perceptual modeling of virtual acoustic **by listening while singing**

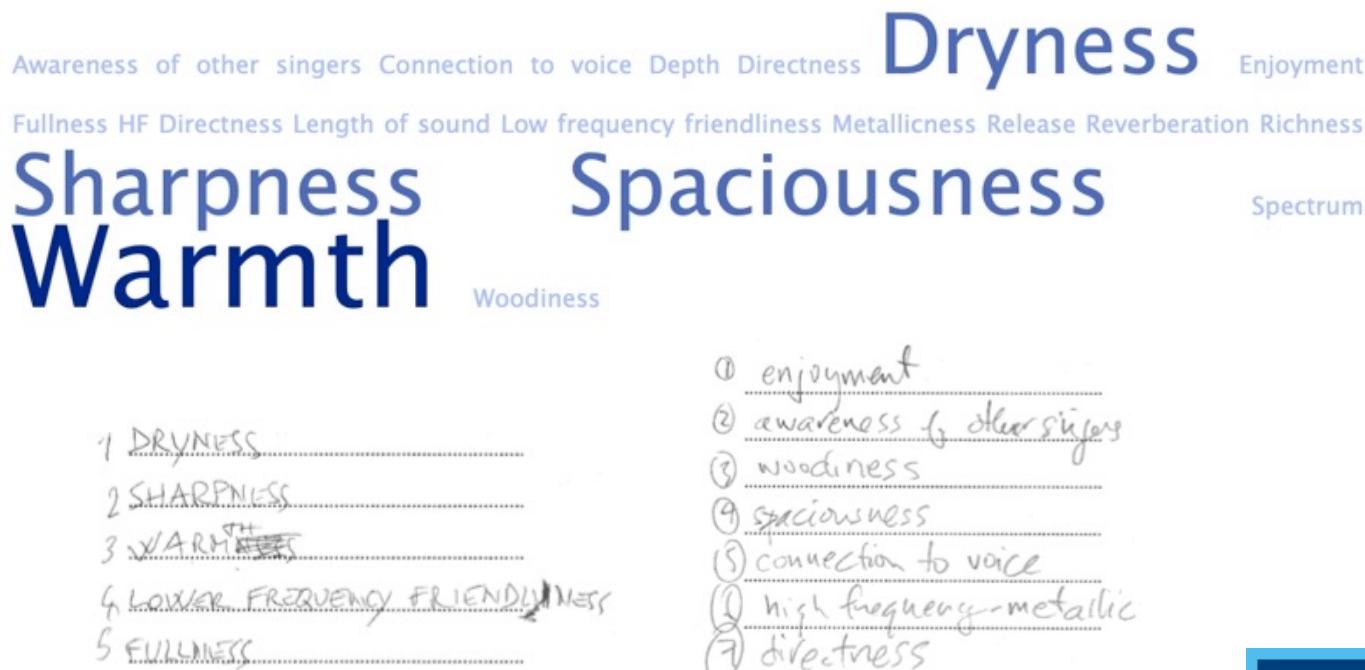


[4] Dairou and Sieffermann, "A comparison of 14 jams characterized by conventional profile and a quick original method, the flash profile", *J. Food Sci.* 67, 2002.

[5] Kaplanis, Bech, Tervo, Pätynen, Lokki, vW, and Jensen, "A rapid sensory analysis method for perceptual assessment of automotive audio," *J. Audio Eng. Soc.*, 65(1/2), 2017.

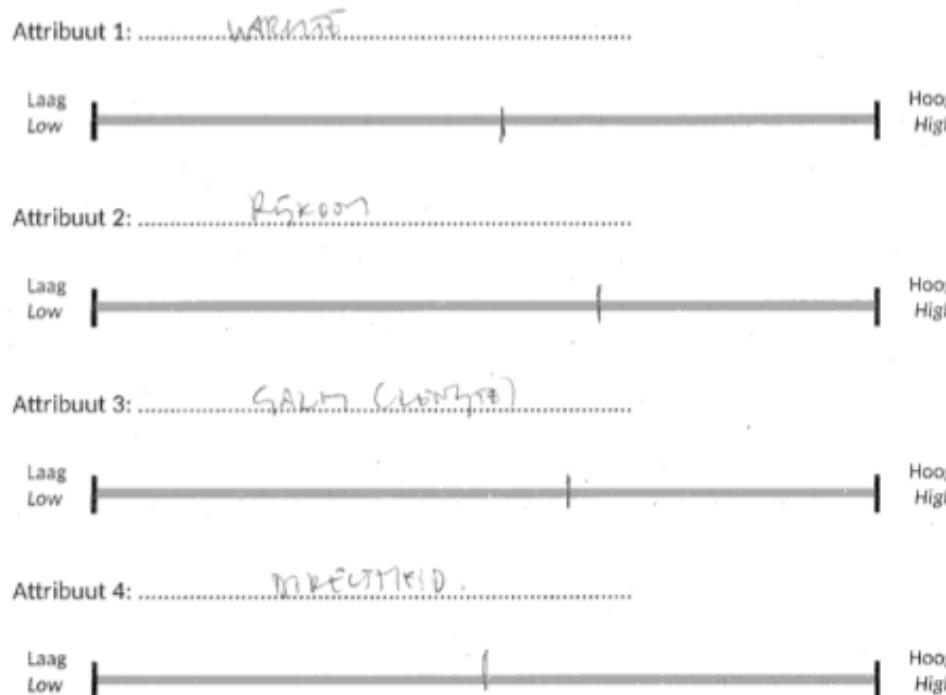
Reproduction: Perceptual assessment data analysis

- **Methodology:** Flash Profile rapid sensory analysis
 - 1) **Elicitation phase:** individual semantic definition of perceptual attributes to characterize differences between virtual spaces



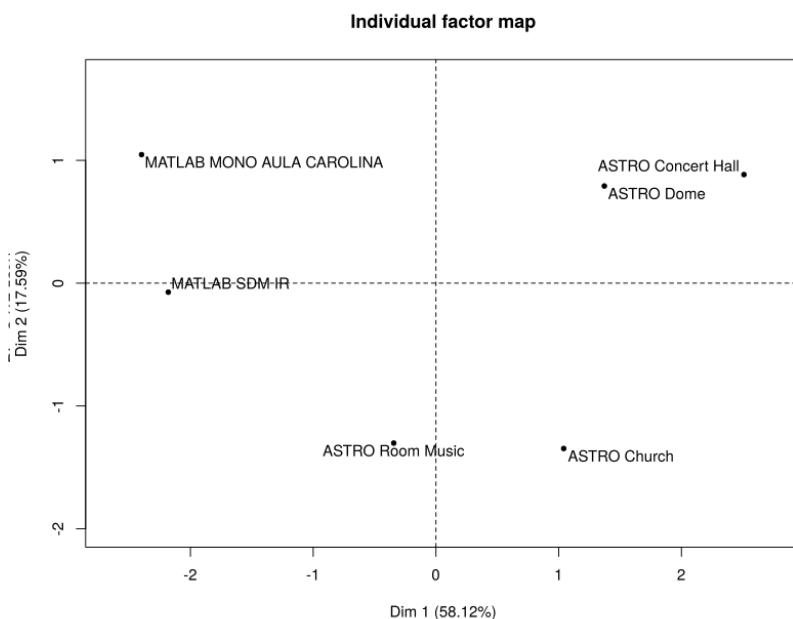
Reproduction: Perceptual assessment data analysis

- **Methodology:** Flash Profile rapid sensory analysis [4,5]
 - **2) Ranking phase:** continuous-scale (low-high) quantification of each perceptual attribute for each virtual space



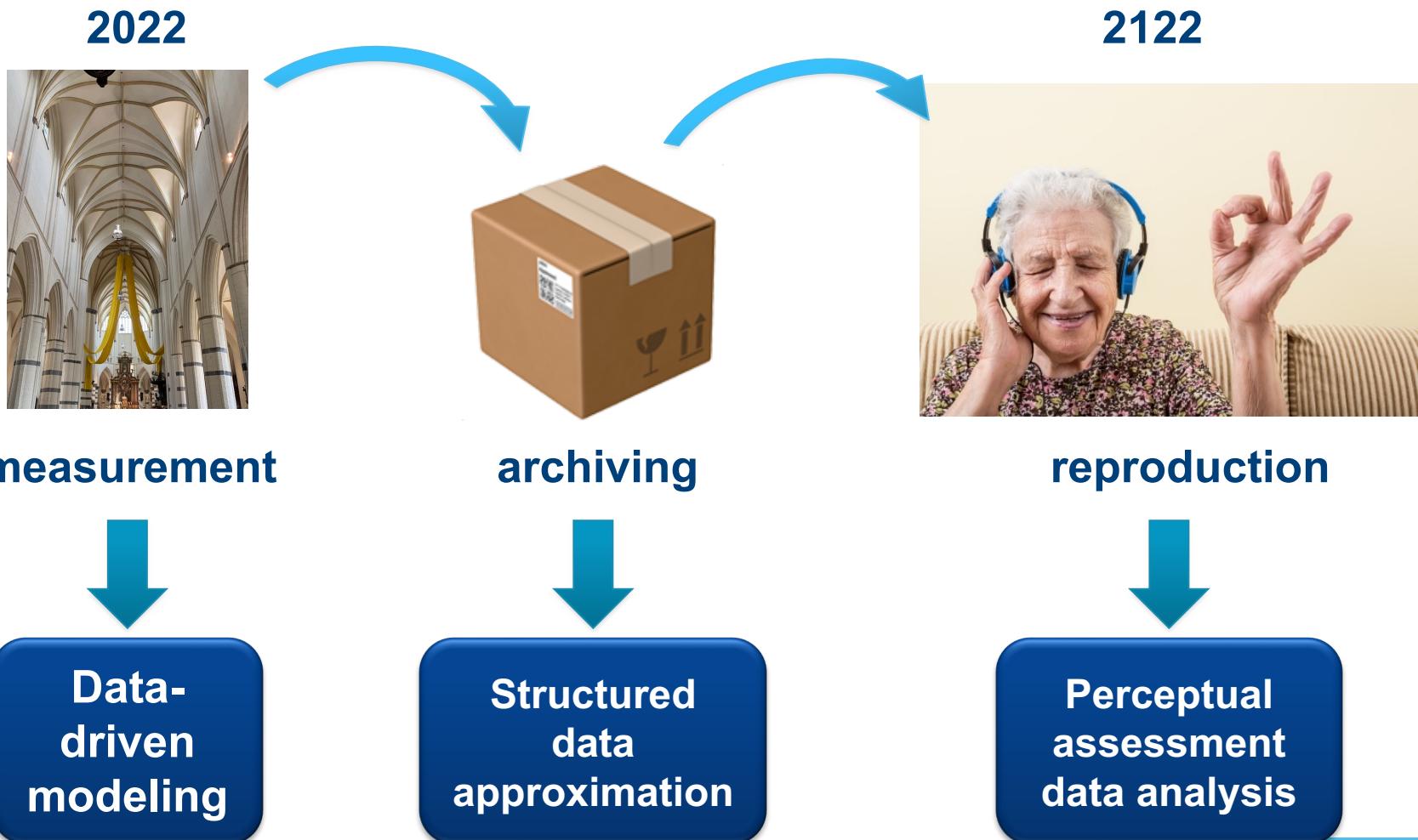
Reproduction: Perceptual assessment data analysis

- **Methodology:** Flash Profile rapid sensory analysis [4,5]
 - **3) Statistical data analysis:** multi-factor principal component analysis + clustering



- 75% of perceived variance among spaces is modeled by two largest principal components
- two largest principal components correlate to perceptual attributes of **spaciousness/reverberance** (Dim 1) and **spectral content** (Dim 2)
- six virtual spaces can then be mapped into this 2-D principal component subspace

Role of AI



Thank you...

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