Acoustic-similarity-based Musical Instrument Hierarchy and Its Application to Musical Instrument Identification

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Motivation and Aim

Timbre:

One of the basic parameters of sounds, but it is not known well

What physical features are corresponded to timbres? How can we feel the (un)similarity of timbres?

Taxonomy:

Study on making a hierarchy of something for understanding it

Making a hierarchy of timbres will help us to understand timbres

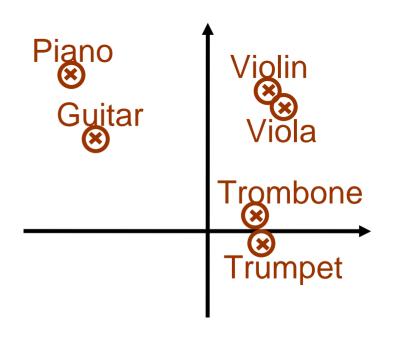
Two Strategies for Making a Hierarchy of Timbres

- Human-perception-based
 - Studied in the field of psychoacoustics
 - but few reports of large-scale experiments
 the burden on human subjects
- Acoustical-similarity-based
 - Facilitates a large-scale one
 - but no reports

⇐ the lack of large-scale DB

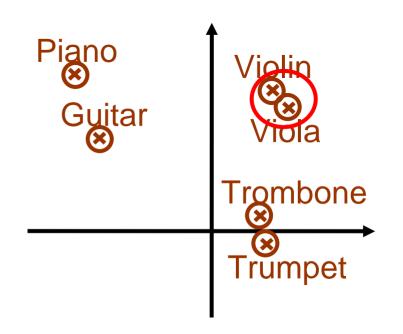
We adopt the second strategy We report a large-scale one using RWC-MDB

- 1. Let each instrument be a cluster
- 2. Merge the closet pair of clusters into a single cluster
- 3. Repeat step 2 until all of the instruments are merged into a single cluster



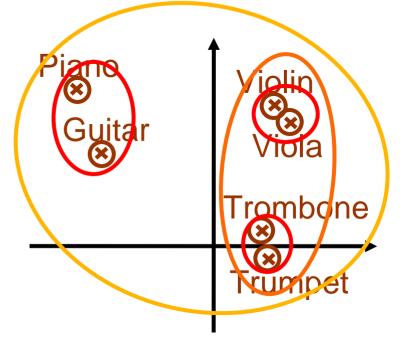
feature space

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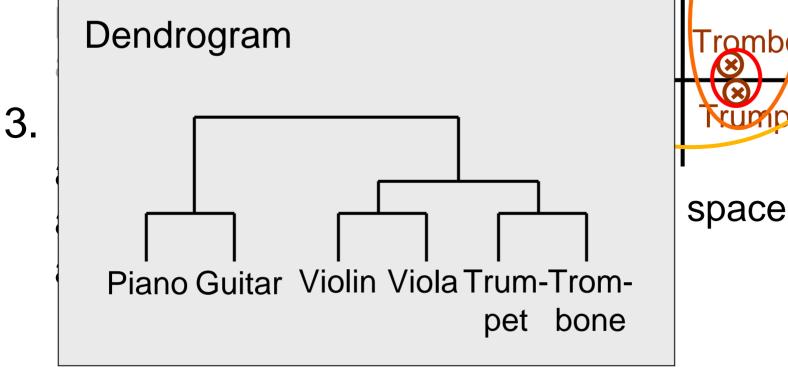
feature space

Guitar

rombone

rupide

- 1. Let each instrument be a cluster
- 2. Merge the closet



What is the Issue?

• What feature space is used?

How to obtain positions in the feature space (They vary according to pitch, etc.)

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 How to obtain positions in the feature space (They vary according to pitch, etc.)
 Approximate the distribution of each instr from a large number of sounds
 ➡ More robust than using single sound

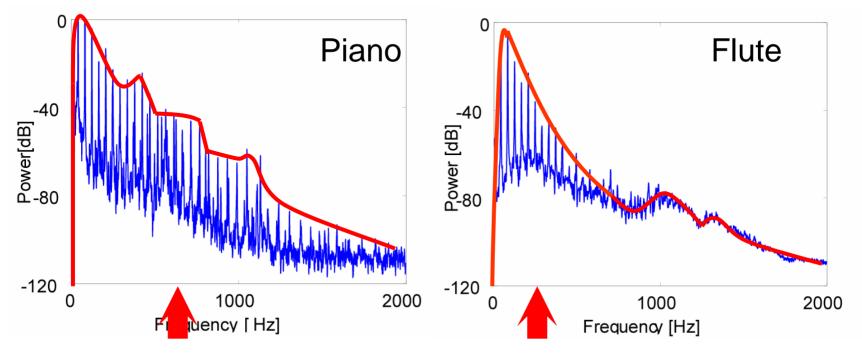
Details of the Method of Making a Hierarchy

- 1. Feature Extraction
- 2. Dimensionality Reduction
- 3. Calculation of the Mahalanobis Distances
- 4. Hierarchical Clustering

Details of the Method

Feature Extraction The 129 features we previously proposed are extracted

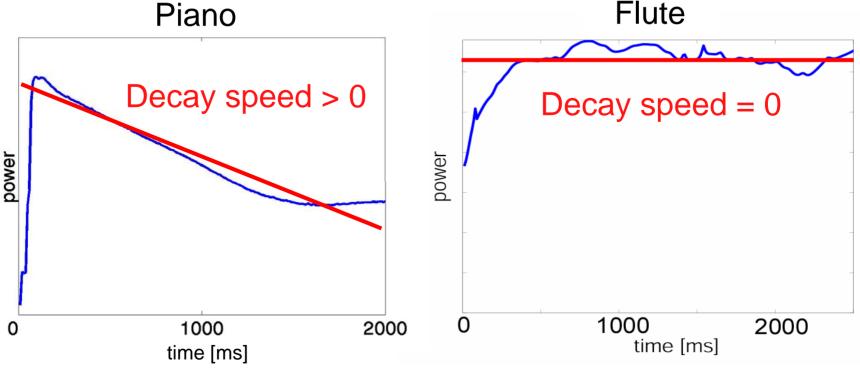
e.g. Spectral centroid



Details of the Method

Feature Extraction The 129 features we previously proposed are extracted

e.g. Decay speed of power



Details of the Method

2. Dimensionality Reduction

129 dim.
79 dim.
18 dim.
Principal Component Analysis
Linear Discriminant Analysis

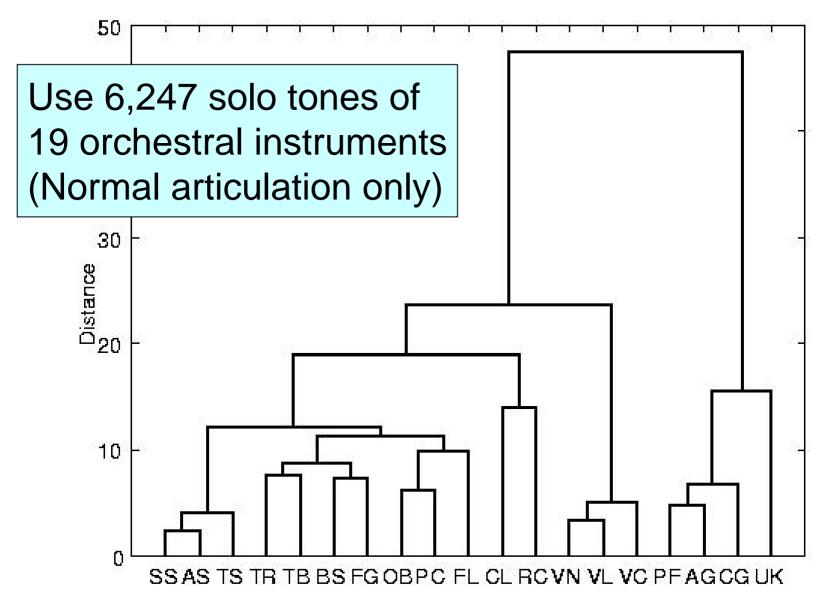
3. Calculation of Mahalanobis Distances

- (1) Calculate mean μ_i and covariance Σ_i of each instrument
- (2) Calculate the following equation:

$$D(i, j) = (\mu_j - \mu_j)' \Sigma^{-1} (\mu_j - \mu_j)$$
$$(\Sigma = (\Sigma_i + \Sigma_j) / 2)$$

4. Hierarchical Clustering

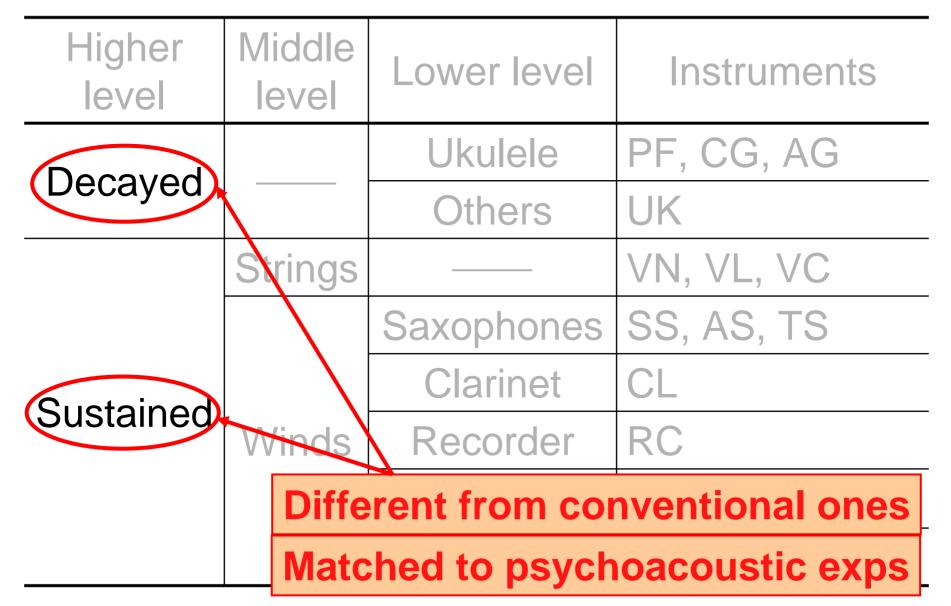
Results of Making a Hierarchy



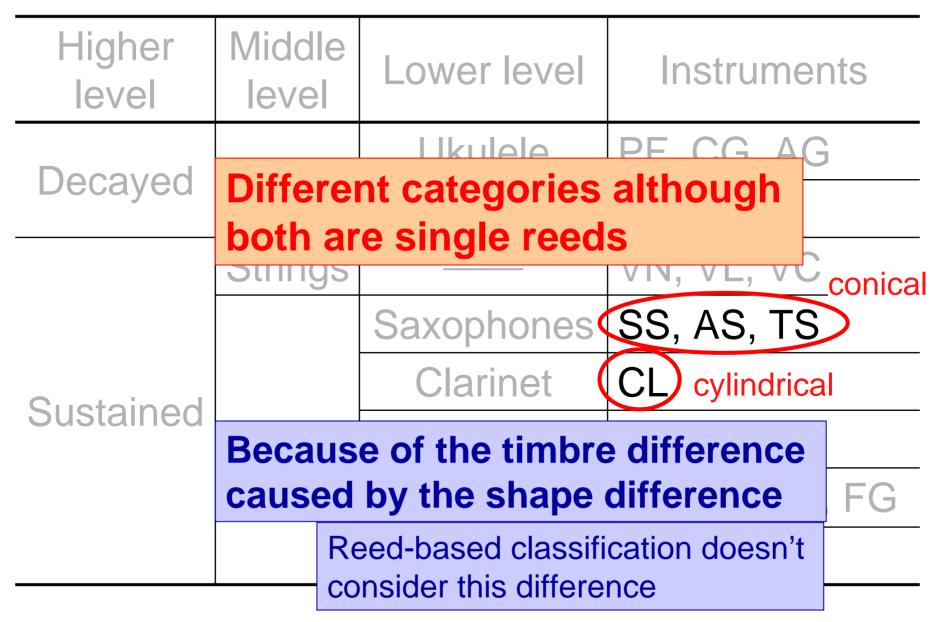
Categories based on Our Hierarchy

Higher level	Middle level	Lower level	Instruments
Decayed		Ukulele	PF, CG, AG
		Others	UK
Sustained	Strings		VN, VL, VC
	Winds	Saxophones	SS, AS, TS
		Clarinet	CL
		Recorder	RC
		Brasses + α	TR, TB, BS, FG
		Others	OB, PC, FL

Categories based on Our Hierarchy



Categories based on Our Hierarchy



Application: Category-level Identification of Non-registered Instruments

Non-registered Instruments: instrs that are *not included* in training data

- Numerous kinds of instruments
- Recent technology of making novel sounds
 Impossible to prepare all of the sounds

Treating non-registered instrs is needed

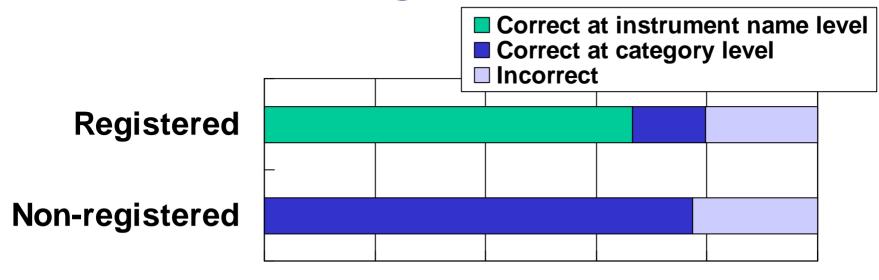
Our solution: Identify them at category level

Application: Category-level Identification of Non-registered Instruments

- If a sound is registered
 ⇒ "It's a violin"
- If a sound is not registered
 ⇒ "I don't know this, but it's a kind of strings"

For this identification, our instrument categorization is applied

Application: Category-level Identification of Non-registered Instruments



0% 20% 40% 60% 80% 100%

• Training & registered test sets: Real Instrs (6,247 solo tones of 19 instrs from RWC-MDB-I-2001)

 Non-registered test set: Electric sounds (from MU2000 (Yamaha))

Conclusions

- Acoustical-similarity-based Musical Instrument Hierarchy:
- Made from a large-scale musical sound DB
- Compared with psychoacoustic exps
 It refrects timbres better
- **Non-registered Musical Instruments:**
- Essential in musical instrument identification
- We proposed identifying categories of them
 Our instr hierarchy was applied

Piano	Piano	
Guitars	Classical Guitar	Acoustic Guitar
	Ukulele	
Strings	Violin	Cello
	Viola	
Brass	Trumpet	Trombone
Saxophones	Soprano Sax	Tenor Sax
	Alto Sax	Baritone Sax
Double Reeds	Oboe	Faggoto
Clarinet	Clarinet	
Air Reeds	Piccolo	Recorder
	Flute	