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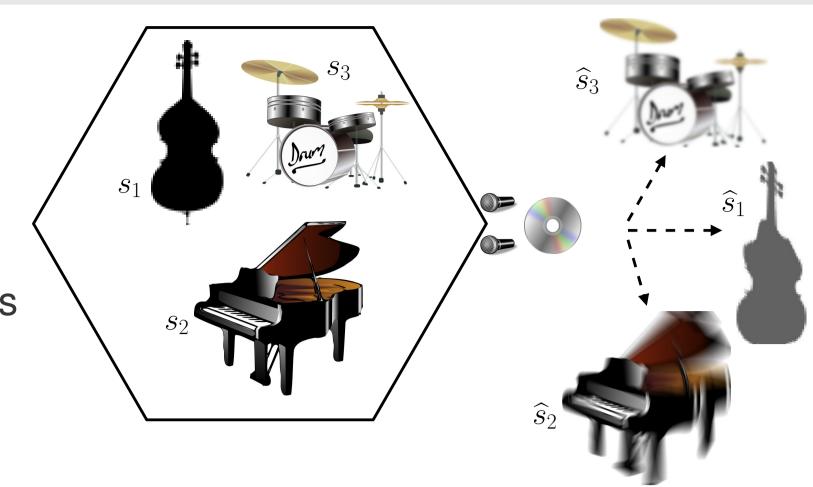
The PEASS Toolkit - Perceptual Evaluation methods for Audio Source Separation

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THE PEASS TOOLKIT: overview

A toolkit for the perceptual evaluation of audio source separation

- The PEASS Software: a set of objective measures to predict the perceptual quality of the source/image estimates
- The PEASS Listening Test GUI: a Matlab MUSHRA GUI realized for the proposed test protocol
- ullet The PEASS Subjective Database: a set of subjective measures resulting from listening tests (20 subjects imes 80 sounds \times 4 rating criteria)



The PEASS Toolkit is freely available at http://bass-db.gforge.inria.fr/peass/

- [1] Subjective and objective quality assessment of audio source separation, V. Emiya, E. Vincent, N. Harlander, V. Hohmann, IEEE Trans. on Audio, Speech and Language Processing, submitted, 2010.
- [2] Multi-criteria subjective and objective evaluation of audio source separation, V. Emiya, E. Vincent, N. Harlander, V. Hohmann, AES 38th Int. Conf. on Sound Quality Evaluation, Pitea, Sweden, June 2010.

MOTIVATION: the need for a multi-criteria perceptuallybased evaluation

Existing model for distortion decomposition [3]:

$$\widehat{s}_j(t) - s_j(t) = e_j^{\text{target}}(t) + e_j^{\text{interf}}(t) + e_j^{\text{artif}}(t)$$
 (1)

- \bullet e_i^{target} denotes the error component related to the target distortion,
- e_i^{interf} denotes the interference from concurrent sources,
- e_i^{artif} is the remaining distortion component (artifacts and noise).

Defining and estimating the distortion components e_i^{target} , e_i^{interf} , e_i^{artif} is not trivial. Due to the allowed distortions in use today (time-invariant spatial and filtering distortions), the decomposition is not satisfying.

Existing quality measures: energy ratios SDR, ISR, SIR, SAR are poorly correlated with subjective scores.

Proposed multi-criteria listening test protocol

A series of 4 MUSHRA tests including several dedicated anchors:

- Rate the *global quality* compared to the reference.
- Rate the quality in terms of *preservation of the target source*.
- Rate the quality in terms of *suppression of other sources*.
- Rate the quality in terms of absence of additional artificial noise.

EVALUATION RESULTS: prediction performance and evaluation at SiSEC 2010

PROPOSED OBJECTIVE MEASURES: what's new?

A better distortion decomposition is achieved by:

- splitting the signals into subbands using gammatone filters;
- segmenting each subband signal into overlapping frames;
- decomposing each frame into distortion components using a matched FIR filter
- reconstructing the full distortion components

Some auditory-motivated features are derived using PEMO-Q/PSM [4]:

$$q_j^{\text{overall}} \triangleq \text{PSM}(\widehat{\mathbf{s}}_j, \mathbf{s}_j)$$
 (2)

$$q_i^{\text{target}} \triangleq \text{PSM}(\widehat{\mathbf{s}}_i, \widehat{\mathbf{s}}_i - \mathbf{e}_i^{\text{target}})$$
 (3)

$$q_j^{\text{interf}} \triangleq \text{PSM}(\widehat{\mathbf{s}}_j, \widehat{\mathbf{s}}_j - \mathbf{e}_j^{\text{interf}})$$
 (4)

$$q_i^{\mathsf{artif}} \triangleq \mathrm{PSM}(\widehat{\mathbf{s}}_j, \widehat{\mathbf{s}}_j - \mathbf{e}_i^{\mathsf{artif}}) \tag{5}$$

By combining the 4 features in a non-linear way to predict subjective scores $(T_1) - (T_4)$, a set of objective measures is finally output:

Accuracy = 0.23; Monotonicity = 0.21;

IPS (0-100)

- OPS: the Overall Perceptual Score,
- TPS: the Target-related Perceptual Score,
- IPS: the Interference-related Perceptual Score,
- APS: the Artifacts-related Perceptual Score.

---Old+SxR ---New+SxR Old+PSM — New+PSM bal quality Monotonicity Consistency Accuracy - Ex.1 1: prediction of glo DEMO SiSSE 50 100 100 OPS (0-100) TPS (0-100) Accuracy = 0.64; Monotonicity = 0.64; Accuracy = 0.16; Monotonicity = 0.34; 3 4 SIR (dB) Task Ex.5 — Ex.1

1 2 3 4

Prediction results (cross-validation on the PEASS database) for the 4 tasks: curves are various combinations of the old/new decompositions with the energy ratio/PEMO-Q measures, as a function of the number of features.

Feature vector size

BSS eval vs. PEASS: scatter plots of the SiSEC 2010 results for the set of *Professionally* produced music recordings.

[3] E. Vincent, R. Gribonval, C. Févotte, Performance measurement in blind audio source separation, IEEE Trans. on Acoustics, Speech and Signal Proc., 14 (4), 2006. [4] R. Huber, B. Kollmeier, PEMO-Q – A New Method for Objective Audio Quality Assessment Using a Model of Auditory Perception, IEEE Trans. on Acoustics, Speech and Signal Proc., 14 (6), 2006.

1 2 3 4

Feature vector size





Accuracy = 0.29; Monotonicity = 0.31;

50

APS (0-100)

1 2 3 4