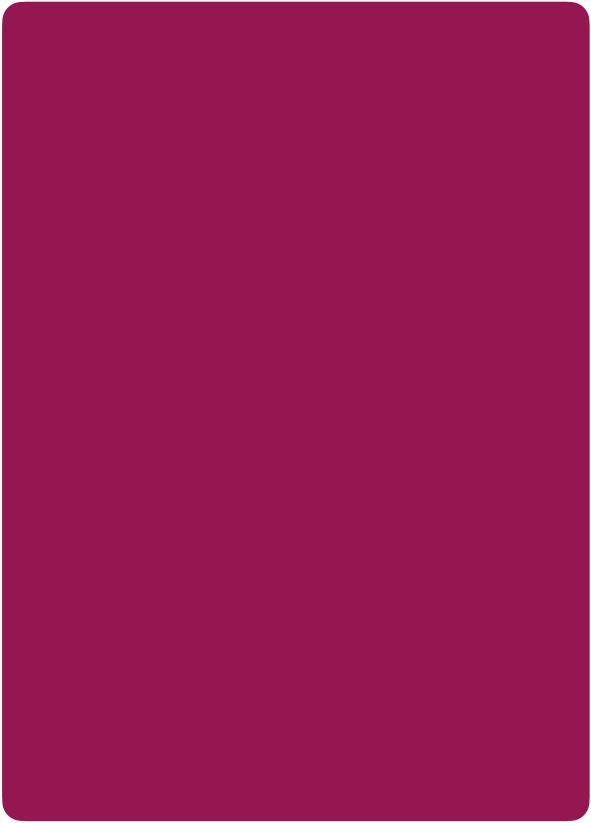






How do Performance Features relate to Expressive Dimensions?

Dimension 1	Dimension 2	Dimension 3	Dimension 4	
PP $(R^2 = 0.24)$	$PP(R^2 = 0.18)$	$PP(R^2 = 0.26)$	$PP(R^2 = 0.24)$	
loudness avg 0.51***	loudness sk 0.45**	loudness std -0.53^{**}	beat period k -0.34^* loudness std -0.44^*	
	$MF (R^2 = 0.00)$ minorness 0.15	$MF (R^2 = 0.00)$ articulation -0.15	$ \text{MF } (R^2 = 0.29) $ $ \text{rhythmic complexity} 0.52^* $ $ \text{tonal stability} \qquad 0.84^{***} $	
HF $(R^2 = 0.22)$ valence sk 0.48^{**}	$HF (R^2 = 0.00)$ valence avg 0.14	$HF (R^2 = 0.36)$ valence k 0.42^{**} arousal avg -1.24^{***} valence std 0.27^* valence avg -0.82^*	HF ($R^2 = 0.09$) valence k -0.33^*	



How do Performance Features relate to Expressive Dimensions?

Dimension 1	Dimension 2	Dimension 3	Dimension 4	
PP ($R^2 = 0.24$) loudness avg 0.51^{***}	$PP (R^2 = 0.18)$ loudness sk 0.45^{**}	$PP (R^2 = 0.26)$ loudness std -0.53^{**}	$PP (R^2 = 0.24)$ beat period k -0.3 loudness std -0.4	$34^* \\ 44^*$
	$MF (R^2 = 0.00)$ minorness 0.15	$MF (R^2 = 0.00)$ articulation -0.15	$MF (R^2 = 0.29)$ rhythmic complexity 0.52 tonal stability 0.84)* 1***
HF $(R^2 = 0.22)$ valence sk 0.48^{**}	$HF (R^2 = 0.00)$ valence avg 0.14	$HF (R^2 = 0.36)$ valence k 0.42^{**} arousal avg -1.24^{***} valence std 0.27^* valence avg -0.82^*	$HF (R^2 = 0.09)$ valence k -0.3	33*



Sorting Musical Expression: Characterization of Descriptions of Expressive Piano Performances

Carlos Cancino-Chacón^{1,4}, Silvan Peter¹, Shreyan Chowdhury¹, Anna Aljanaki² and Gerhard Widmer^{1,3}

¹Institute of Computational Perception, Johannes Kepler University Linz, Austria ²Institute of Computer Science, University of Tartu, Estonia ²LIT AI Lab, Linz Institute of Technology, Linz, Austria

⁴RITMO Centre for Interdisciplinary Studies in Rhythm, Time and Motion, University of Oslo, Norway











The Con Espressione Game

Research Aims: Find the dimensions of musical expression that can be attributed to a performance, as perceived and described in natural language by listeners

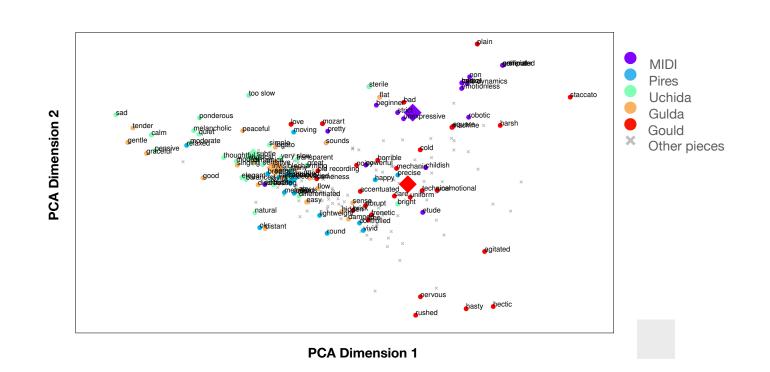
- · Web based questionnaire: verbal descriptors of expressive performance.
- Different performances of 9 classical piano pieces (45 performances)
- Dataset enriched with score-to-performance alignments

Composer	Piece	#	Pianists
Bach	Prelude No.1 in C, BWV 846 (WTC I)	7	Gieseking, Gould, Grimaud, Kempff, Richter, Stadtfeld, MIDI
Mozart	Piano Sonata K.545 C major, 2nd mvt.	5	Gould, Gulda, Pires, Uchida, MIDI deadpan
Beethoven	Piano Sonata Op.27 No.2 C# minor, 1st mvt.	6	Casadesus, Lazić, Lim, Gulda, Schiff, Schirmer
Schumann	Arabeske Op.18 C major (excerpt 1)	4	Rubinstein, Schiff, Vorraber, Horowitz
Schumann	Arabeske Op.18 C major (excerpt 2)	4	Rubinstein, Schiff, Vorraber, Horowitz
Schumann	Kreisleriana Op.16; 3. Sehr aufgeregt (ex 1)	5	Argerich, Brendel, Horowitz, Vogt, Vorraber
Schumann	Kreisleriana Op.16; 3. Sehr aufgeregt (ex 2)	5	Argerich, Brendel, Horowitz, Vogt, Vorraber
Liszt	Bagatelle sans tonalité, S.216a	4	Bavouzet, Brendel, Katsaris, Gardon
Brahms	4 Klavierstücke Op.119, 2. Intermezzo E minor	5	Angelich, Ax, Serkin, Kempff, Vogt

What are the main dimensions for expressive character?

Principal component analysis (PCA) on the occurrence matrix of the terms and find 4 principal dimensions

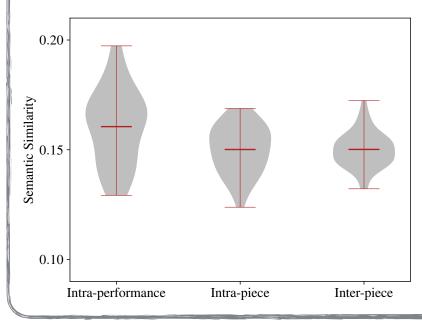
Dimension 1				Dimension 2			
positive corr	elation	negative correlation		positive c	orrelation	negative correlation	
hectic	0.17	sad	-0.20	rushed	0.22	hard	-0.19
staccato	0.15	gentle	-0.18	nervous	0.20	stumbling	-0.18
hasty	0.15	tender	-0.18	too fast	0.17	staccato	-0.17
agitated	0.14	calm	-0.16	bit	0.16	ponderous	-0.14
irregular	0.14	graceful	-0.16	hasty	0.15	monotonous	-0.13
Dimension 3				Dimension 4			
positive correlation		negative correlation		positive correlation		negative correlation	
monotonous	0.22	heavy	-0.14	ok	0.24	cold	-0.15
bad	0.17	graceful	-0.13	happy	0.21	warm	-0.14
warm	0.16	smooth	-0.12	joyful	0.19	floating	-0.14
peaceful	0.16	ponderous	-0.12	free	0.15	blurred	-0.14
beautiful	0.15	soaring	-0.10	breathy	0.14	mysterious	-0.13

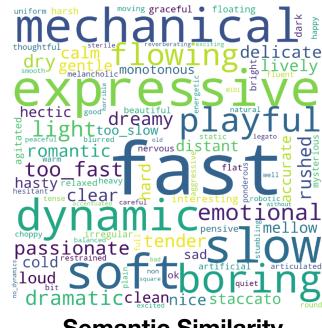


How similarly do listeners describe the performance of a piece?

Distribution of Terms

- 94 participants (on average listened to 4.5 out of 9 pieces)
- 88% had some musical training
- 1,515 individual descriptions, 3,166 terms (45% unique)





Semantic Similarity

- Semantic similarity for short sentences by [Li et al., 2007]
- Intra-performance: same piece, same pianist
- Intra-piece: same piece, other pianists
- inter-piece: other pieces

Pile Sorting Experiment

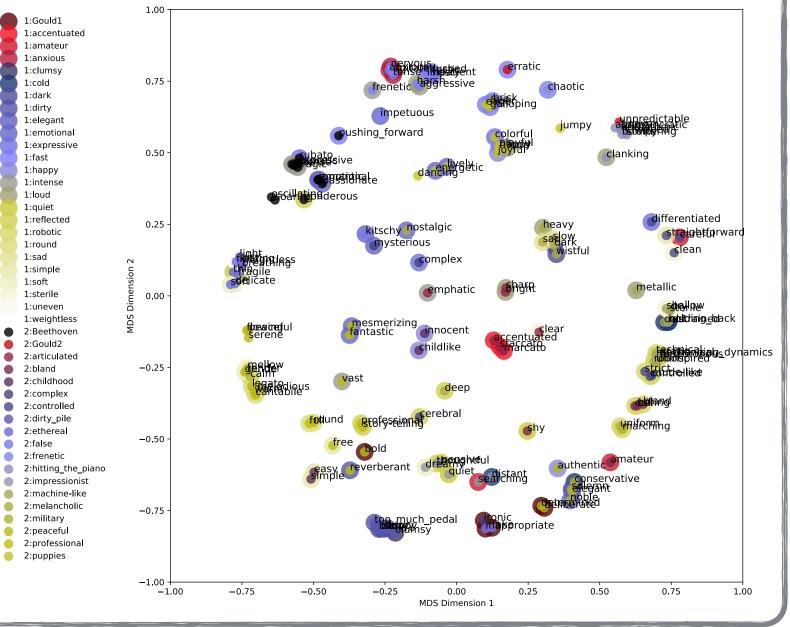
Participants

- Two groups of expert musicians (G1 and G2)
- Each group sorted (independently) 150 of the most frequently used terms that had been collected through the CEG.
- The number of piles as well as the types of similarity within the piles were left open.

Explore the interactive visualization!

Results

- G1 (25 piles), G2 (19 piles)
- Average maximal overlap (Szymkiewicz-Simpson coefficient):
- 62 % piles G1 with piles G2
- 65% piles G2 with piles G1
- Multidimensional Scaling (MDS) to explore the structure of the terms



Get the Dataset!



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number 287152.





Pile Sorting Experiment

- We want to see how expert listeners (i.e., classically trained musicians)
 categorize the descriptions of the terms in the Con Espressione Game
 - We selected 150 of the most representative terms
- Pile Sorting: Participants sort each term into categories (piles) in a collaborative fashion

