



Musical Similarity: More perspectives and compound techniques



CS 275B/Music 254

Musical similarity

- ▶ Similarity studies in general
- ▶ Reductionist approaches
- ▶ Social cognition
- ▶ Timbral confounds
- ▶ **Affective similarity**
- ▶ **Cognitive distance metrics**
- ▶ **Compound search techniques**

Cognitive distance metric (1)

I. Basic Pitch-Accent Structure			Range = 0-4
A.	If meter matches target		Max = 1.00
	and	If subunit (e.g. quarter note) is the same	Score = 1.00
	or	If subunit is different (e.g., 4/8 vs. 2/4)	Score = 0.50
	Else		Score = 0.00
B. Percentage of matched pitches on primary beats*			Max = 2.00
	If matching number of scale degrees=100%		Score = 2.00
	or	If matching number of scale degrees =>90%	Score = 1.33
	or	If matched number of notes/unit =>80% Score=0.67	Score = 0.67
	Else		Score = 0.00
C. Percentage of matched pitches on secondary beats			Max = 1.00
	If matching number of scale degrees=100%		Score = 1.00
	or	If matching number of scale degrees=>90%	Score = 0.67
	or	If matched number of notes/unit =>80% Score=0.33	Score = 0.33
	Else		Score = 0.00

Cognitive distance metric (2)

II. Basic Harmonic-Accent Structure			Range = 0-6
A. Mode of work (major, minor, other)			Max = 1.00
	If modes match		Score = 1.00
	Else		Score = 0.00
B. Percentage of matched chords on downbeat***			Max = 2.50
	If unambiguous matches on primary beats =>90%		Score = 2.50
	or	If unambiguous matches on primary beats =>80%	Score = 2.00
	or	If unambiguous matches on primary beat =>70%	Score = 1.50
	Else		Score = 0.00
C. Percentage of matched chords on secondary beats**			Max = 2.00
	If unambiguous matches =>90%		Score = 2.00
	or	If unambiguous matches =>80%	Score = 1.50
	or	If unambiguous matches =>70%	Score = 1.00
	Else		Score = 0.00
D. Percentage of matched chords on tertiary beats			Max = 0.50
	If unambiguous matches =>90%		Score = 0.50
	Else		Score = 0.0

Cognitive distance metric (3)

Example	Pitch-Accent score		Harmonic-Accent score		Total score (additive)	
	Raw	Ranked	Raw	Ranked	Raw	Ranked
2a	3.67	2	5.5	3	9.17	2
2b	3.67	2	5.0	4	8.67	3
2c	2.67	6	6.0	1	8.67	3
2d	1.17	9	4.5	5	6.67	8
2e	2.67	6	4.0	9	6.67	8
2f	2.33	8	4.5	5	6.83	7
2g	1.00	10	2.0	11	3.00	11
2h	3.50	4	4.5	5	8.00	6
2i	4.00	1	4.5	5	8.50	5
2j	1.00	10	4.0	9	5.00	10
2k	3.33	5	6.0	1	9.33	1

Evaluating search viability and efficiency

- ▶ Krumhansl, 2000 [theory/experiment]
- ▶ Sapp, Liu, Selfridge-Field, 2004 [practical]

“Search effectiveness measures for symbolic music queries in very large databases” ISMIR 2004:


<http://ismir2004.ismir.net/proceedings/p051-page-266-paper135.pdf>

Search Effectiveness (1)

Sapp, Liu, Selfridge-Field (ISMIR 2004)

Data

Dataset	Genre	Orig. Code	# Incipits
• US RISM A/II	Instrumental, Vocal (17th– 18th cents.)	Plaine & Easie	55,490
• Classical*	Instr., Vocal	MIDI	10,718
Total			100,299



Search Effectiveness (2)

	Abbr.	Search type	# states
Pitch features	p1	pch enharmonic pitch class	35
	p2	mi musical interval	(35)
	p3	12p 12-tone pitch class	12
	p4	12i 12-tone pitch interval	(12)
	p5	sd scale-degree (diatonic pitch class)	7
	p6	pge pitch gross contour	3
	p7	pre pitch refined contour	5
Meter features	r1	dur duration	?
	r2	dgc duration gross contour	3
	r3	drc duration refined contour	5
	r4	blv beat level	2
	r5	mlv metric level	?
	r6	mge metric gross contour	3
	r7	mrc metric refined contour	5

Search Effectiveness (3)

Sample search



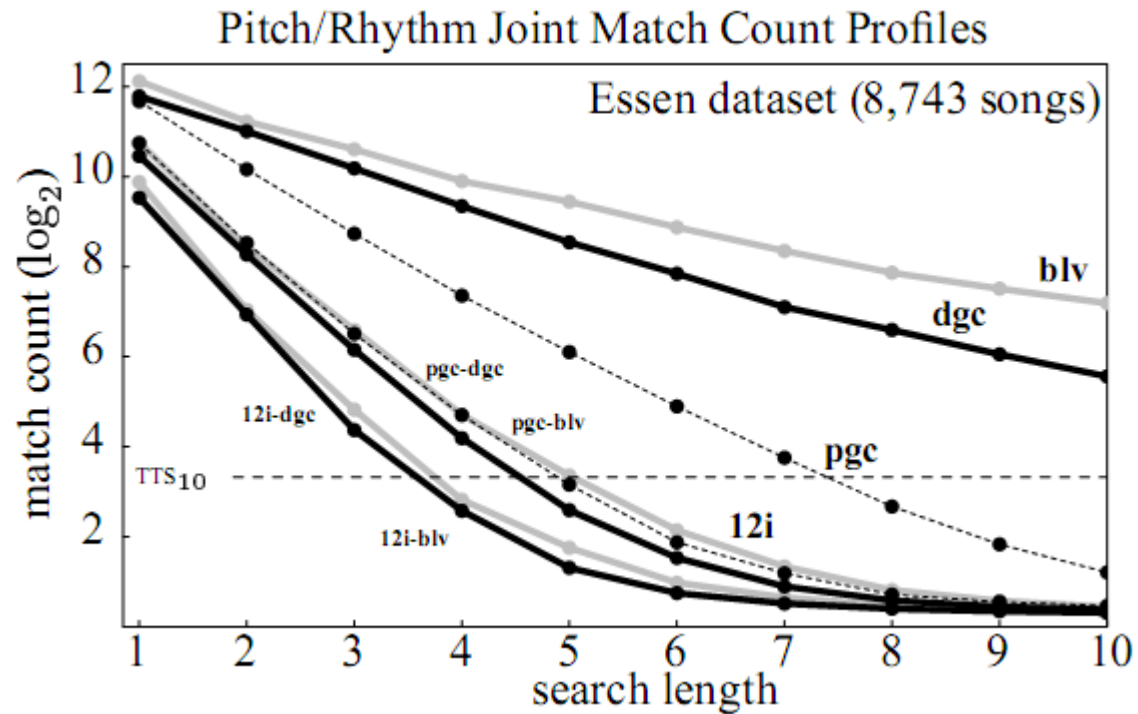
pch	F	A	C	C	C	D	C	A	F	A	G
mi		+M3 +m3	P1	P1	+M2	-M2		-m3	-M3	+M3	-M2
12p	5	9	0	0	0	2	0	9	5	9	7
12i		+4	+3	0	0	+2	-2	-3	-4	+4	-2
sd	1	3	5	5	5	6	5	3	1	3	2

Coupled search

pre	U	U	S	S	u	d		D	D	U	d
dur	E	E	Q	Q	Q	Q	Q.	E	Q	Q	H
dgc	E	L	E	E	E	L		S	L	E	L

blv	1	0	1	1	1	1	1	0	1	1	1
mlv	0	-1	2	0	1	0	2	-1	1	0	2
mge	W	H	W	H	W	H		W	H	W	H
mrc	w	H	w	h	w	H		W	H	w	H

Results




Search Effectiveness (1)

Sapp, Liu, Selfridge-Field (ISMIR 2004)

Data

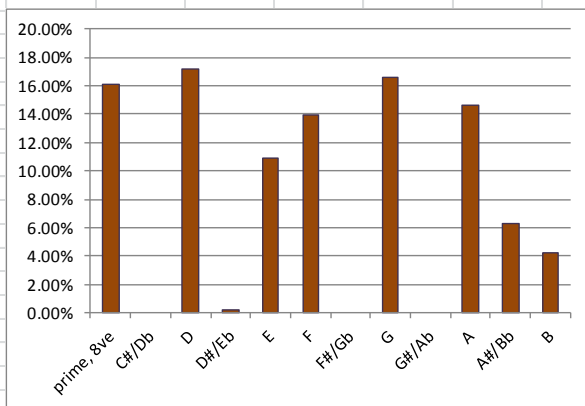
Dataset	Genre	Orig. Code	# Incipits
• US RISM A/II	Instrumental, Vocal (17th– 18th cents.)	Plaine & Easie	55,490
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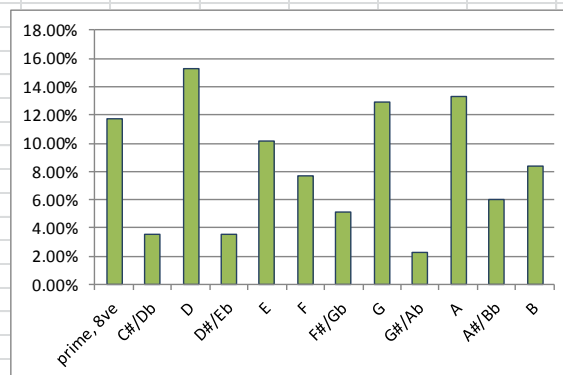
Stats from 2004 study

Repertory	Repeated pitch intervals	Up intervals	Down intervals	Symbols/t heme	Total incipits
15-16th cent/Latin	29,916	66,026	67,151		18,947
15th-16th cent/Polish	57,006	130,030	157,117		6,016
17-18th cent/US RISM	110,478	344,621	399,079		55,491
18th-19 cent/classical	19,241	80,166	86,430		10,722
Essen Europe	43,581	53,033	58,815		6,232
Essen Asia	16,441	54,577	65,684		2,240
Luxembourg	4,355	9,720	11,927		612
	281,018	738,173	846,203		100,260

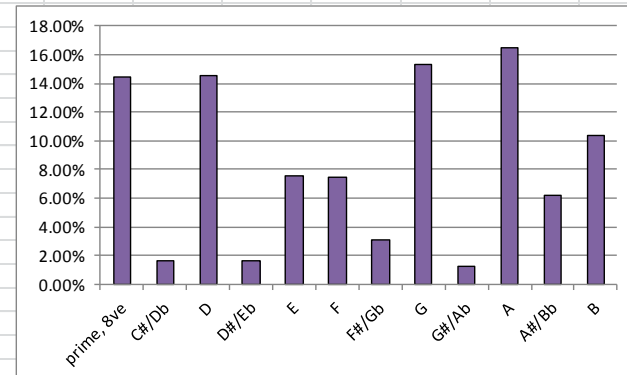
Normalized pitch usage by repertory



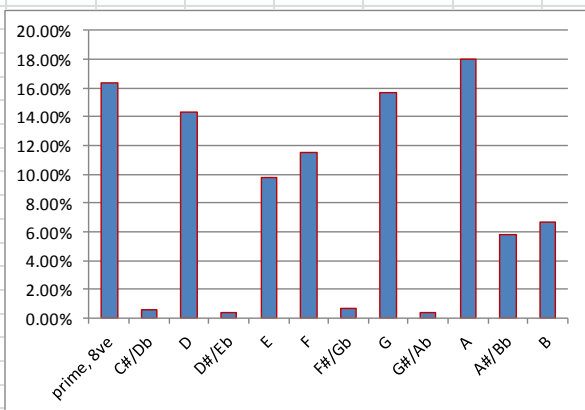
Latin motets



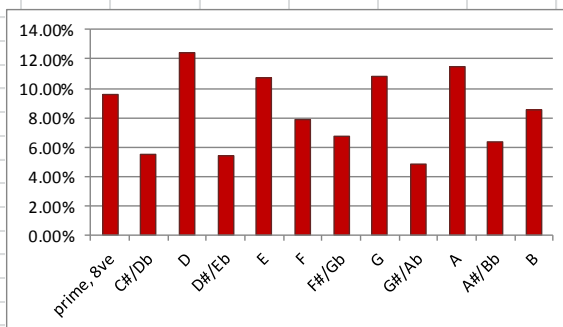
US RISM



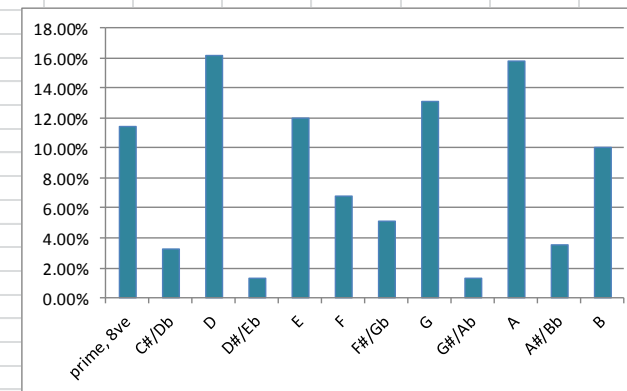
Essen Europe



Polish



Classical



Essen Asia



Form: Boundary strength

- ▶ **Markus Pearce, Daniel Müllensiefen, Geraint Wiggins** (Goldsmith): et al: “Melodic Grouping in Music Information Retrieval”

GPR	Description	n	Boundary Strength
2a	Rest		absolute length of rest (semibreve = 1.0)
2b	Attack-point	length	$\begin{cases} 1.0 - \frac{n_1+n_3}{2 \times n_2} & \text{if } n_2 > n_3 \wedge n_2 > n_1 \\ \perp & \text{otherwise} \end{cases}$
3a	Register change	pitch height	$\begin{cases} 1.0 - \frac{ n_1-n_2 + n_3-n_4 }{2 \times n_2-n_3 } & \text{if } n_2 \neq n_3 \wedge \\ & n_2 - n_3 > n_1 - n_2 \wedge \\ & n_2 - n_3 > n_3 - n_4 \\ \perp & \text{otherwise} \end{cases}$
3d	Length change	length	$1.0 - \begin{cases} n_1/n_3 & \text{if } n_3 \geq n_1 \\ n_3/n_1 & \text{if } n_3 < n_1 \end{cases}$

After Frankland and Cohen (2004)

Binary (?) melodic segmentation

► Melodic segmentation (binary~)

Incorrect 0 0 0 0 1 0 0 0 1 0 0 0 0 1 0 0 0 0 1 1

Correct 0 0 1 0 0 1 0 0 0 1 0 0 1 0 0 1 0 0 0 1



Pearce's "ground truth" tests using Essen ERK data with binary segmentation(s)

Marcus T. Pearce, Daniel Müllensiefen, and Geraint A. Wiggins,

"Melodic Grouping and Music Information Retrieval" (2010)

[same], "A Comparison of Statistical and Rule-Based Models and Melodic Segmentation", ISMIR (2008).

Pearce results (Springer Verlag, 2010)

- GPR2a preference rules work
- Temperley 2001 also useful
- Synthesis of Grouper, LBDM, GPR2a, and Thom better than any one individually
- Unsupervised learned performed better than any statistical method

Pearce et al software and articles

- ▶ Idiom software:

<https://code.soundsoftware.ac.uk/projects/idiom-project>

- ▶ Perceptual segmentation of melodies:

http://www.doc.gold.ac.uk/~mas03dm/papers/icmpc08_PearceMul lensiefenWiggins.pdf

- ▶ Expectation in audio boundaries:

<http://www.ncbi.nlm.nih.gov/pubmed/21180358>