

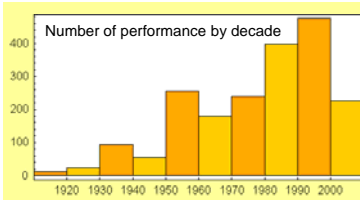
Similarity Measurements in Chopin Mazurka Performances

Craig Stuart Sapp
C4DM Seminar, 11 July 2007
Queen Mary, University of London

Mazurka Project

- 2,210 recordings of 49 mazurkas
= 45 performances/mazurka on average
least: 31 performances of 41/3
most: 64 performances of 63/3

- 105 performers on 160 CDs, 98 hours of music
- Earliest is 1907 Pachmann performance of 50/2



Performers of mazurka 63/3:

Albanowicz (2003)	Friedman (1938)	Pachmann (1927)
Andersen (2003)	Gieseler (1938)	Radwin (1938)
Aukema (1981)	Gonostava (1994)	Reizenstein (1992)
Bart (1996)	Hazarenska (1955)	Pollock (1999)
Ber (2003)	Hazy (1988)	Radwin (1938)
Blask (1995)	Hovren (1949)	Radwin (1938)
Blaumail (1952)	Jalje (1938)	Lang (2001)
Bohacik (1969)	Kapel (1951)	Koren (1989)
Brakowicz (1960)	Kern (1992)	Kozminski (1935)
Braz (1987)	Kubler (1989)	Radwin (1938)
Chiu (1999)	Laiada (1991)	Radwin (1938)
Chiles (1997)	Liszt (2004)	Radwin (1938)
Corst (1951)	Magid (1978)	Shubert (1960)
Czerny-Schubert (1944)	Magn (1975)	Shubert (1960)
Egala (2006)	Michalowski (1933)	Smith (1972)
Fabry (1985)	Mikasa (1970)	Ts'ing (1984)
Ferency (1958)	Milovich (1999)	Ulanov (1952)
Fiere (1977)	Morawa (1969)	Ulanov (1952)
Franca (1956)	Mugaux (1958)	Wawrzyn (1980)
Friedman (1938)	Cziska (1989)	Zak (1997)

<http://mazurka.org.uk/info/discography>

Expressive Audio Features (Piano)

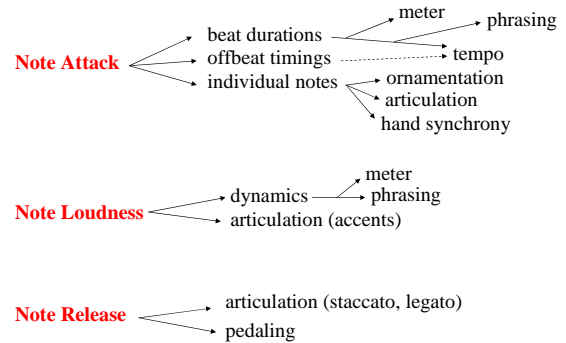
Note Attack

Note Loudness

Note Release

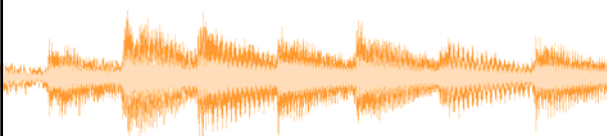
- Not much else a pianist can control
- String instruments have more control variables
- Voice has even more...

Expressive Performance Features



Data Extraction (1)

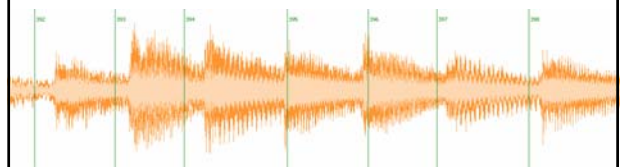
- Extracting beat times from audio files



- Using Sonic Visualiser for data entry processing
<http://www.sonicvisualiser.org>

Data Extraction (2)

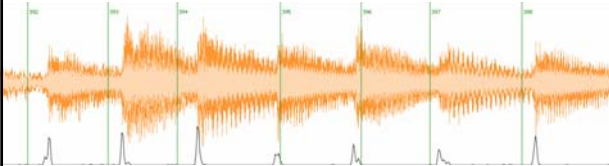
- Step 1: Listen to music and tap to beats (; key)



- Notice taps do not fall on the audio attacks:
 - 23.22 ms hardware granularity built into program
 - Human: ~30 ms SD for constant tempo; ~80 ms SD for mazurkas

Data Extraction (3)

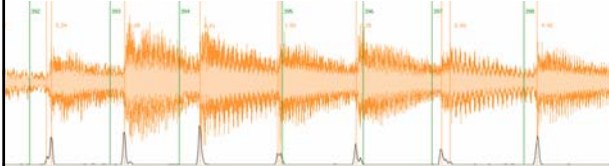
- Step 2: Add onset detection function to the display



- M_z SpectralReflux plugin for Sonic Visualiser:
<http://sv.mazurka.org.uk/download> (Linux & Windows)

Data Extraction (4)

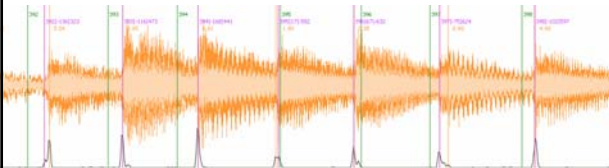
- Step 3: Estimate onset times from function peaks



- Send taps (green) and onsets (orange) to external program:
<http://mazurka.org.uk/cgi-bin/snaptap>
 (no interlayer processing plugins for Sonic Visualiser yet...)

Data Extraction (5)

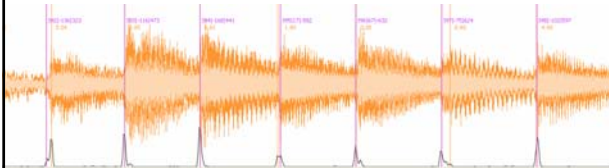
- Step 4: Load snaptap results into SV (purple):



- M_z SpectralReflux currently sensitive to noise (old recordings)
 so snaptap only works on clean recordings.

Data Extraction (6)

- Step 5: Correct errors

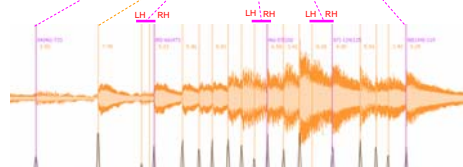


- Hardest part of data entry.
- Down to ~30 min/mazurka for clean recordings.
- 278 performances (of ~6 mazurkas) at this stage.

Well-Behaved

(Mohovich 1999)

- pickup longer than 16th
- LH before RH
- triplets played same speed as sextuplets
- first sextuplet note longest



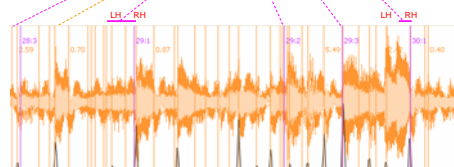
M_z HarmonicSpectrogram

poor low freq resolution...

Misbehaved

(Risler 1920)

- pickup longer than 16th
- LH before RH
- triplets played same speed as sextuplets
- first sextuplet note longest



lots of false positives

lots of noise/clicks

cc:url;cyc

Extracted Feature Data

• Further feature extraction by Andrew Earis (note timings/dynamics)

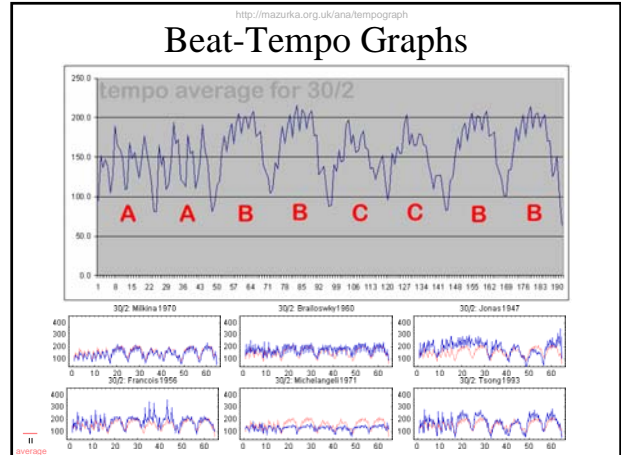
- Beat times durations or tempos
- Audio amplitude at beat locations

0.150	0.604	99	61.6
0.754	0.530	113	65.0
1.284	0.500	120	68.9
1.784	0.512	117	69.9
2.297	0.567	106	66.3
2.864	0.641	94	66.9
3.432	0.567	106	66.1
4.073	0.958	60	63.4
5.072	1.870	32	61.4
6.342	1.870	32	62.5
8.498	1.555	39	62.5
10.320	1.821	33	60.7
11.497	1.177	51	70.4
12.309	0.811	74	65.3
13.145	0.835	72	71.2
13.804	0.658	91	65.5
14.415	0.610	98	66.9
14.415	0.619	97	77.9
15.034	0.622	96	72.7
15.657	0.637	94	70.7
16.294			

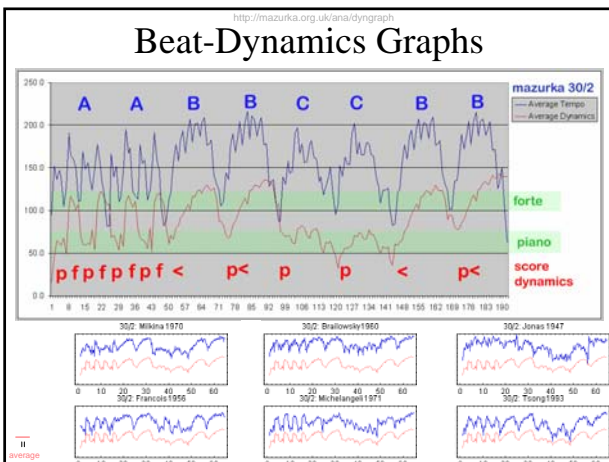
<http://mazurka.org.uk/info/excel/beat>

<http://mazurka.org.uk/info/excel/dyn/gbdyn>

Beat-Tempo Graphs



Beat-Dynamics Graphs



Pearson Correlation

$$r = \frac{\sum_i (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_i (x_i - \bar{x})^2 \sum_i (y_i - \bar{y})^2}}$$

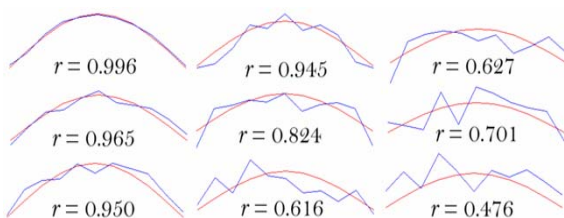
Example:

$$x = (1, 3, 4, 2, 7, 4, 2, 3, 1) \quad y = (4, 4, 3, 5, 5, 6, 4, 3, 2)$$

$$\bar{x} = 3 \quad (\text{average of } x) \quad \bar{y} = 4 \quad (\text{average of } y)$$

$$r = 0.436436$$

Shape Matching



Correlation & Fourier Analysis

correlation: multiply & sum

$$X(k) = \sum_n x(n) e^{-2\pi jnk}$$

spectrum signal sinusoids

$X(k)$ = spectrum, indexed by k (frequency)

$x(n)$ = signal, indexed by n (time)

$e^{-2\pi jnk}$ = set of k complex sinusoids indexed by n

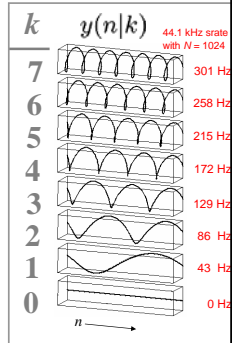
Correlation & Fourier Analysis (2)

Let $y(n|k) = e^{-2\pi jnk}$

Then the DFT can then be written as:

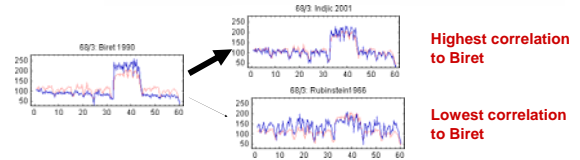
$$X(k) = \sum_n x(n) y(n|k)$$

$$X \left\{ \begin{array}{l} \dots \\ X_3 = \sum x(n) y_3(n) \\ X_2 = \sum x(n) y_2(n) \\ X_1 = \sum x(n) y_1(n) \\ X_0 = \sum x(n) y_0(n) \end{array} \right.$$



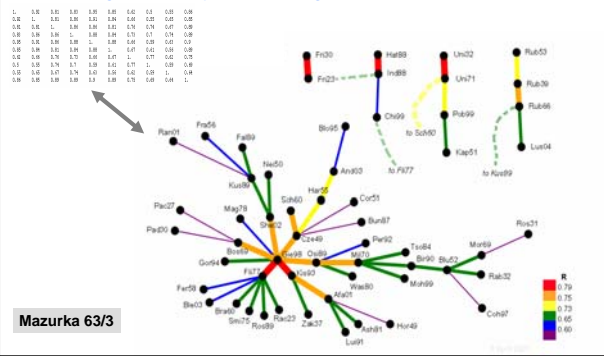
Performance Tempo Correlations

	Bi	Br	Ch	Fl	In	Lu	R8	R6	Sm	Un
Biret	1.	0.92	0.81	0.83	0.95	0.85	0.62	0.5	0.55	0.86
Brailowsky	0.92	1.	0.81	0.86	0.91	0.84	0.66	0.55	0.65	0.85
Chiu	0.81	0.81	1.	0.86	0.86	0.81	0.76	0.74	0.67	0.89
Friere	0.83	0.86	0.86	1.	0.88	0.84	0.73	0.7	0.74	0.89
Indjic	0.95	0.91	0.86	0.88	1.	0.88	0.66	0.59	0.63	0.9
Luisada	0.85	0.84	0.81	0.84	0.88	1.	0.67	0.61	0.56	0.89
Rubinstein 1938	0.62	0.66	0.76	0.73	0.66	0.67	1.	0.77	0.62	0.75
Rubinstein 1966	0.5	0.55	0.74	0.7	0.59	0.61	0.77	1.	0.59	0.69
Smith	0.55	0.65	0.67	0.74	0.63	0.56	0.62	0.59	1.	0.64
Uninsky	0.86	0.85	0.89	0.89	0.9	0.89	0.75	0.69	0.64	1.

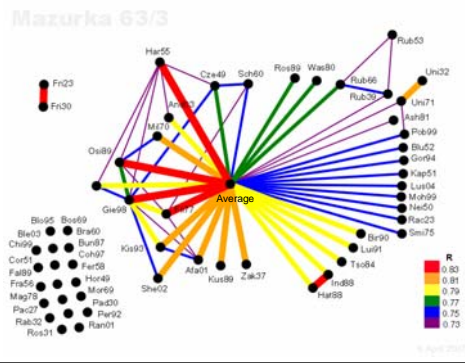


Correlation Maps – Nearest Neighbor

- Draw one line connecting each performance to its closest correlation match
- Correlating to the entire performance length.

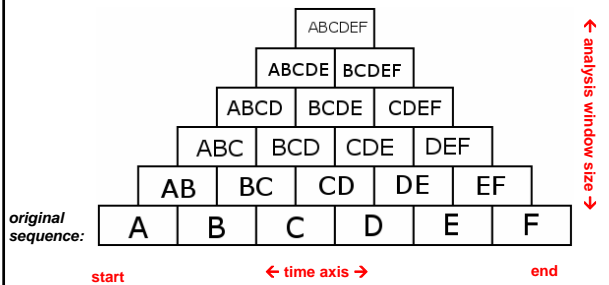


Absolute Correlation Map



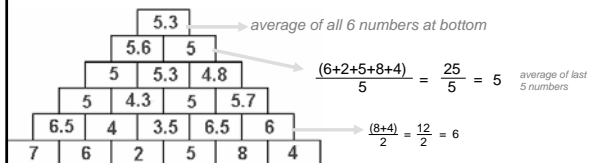
Scape Plotting Domain

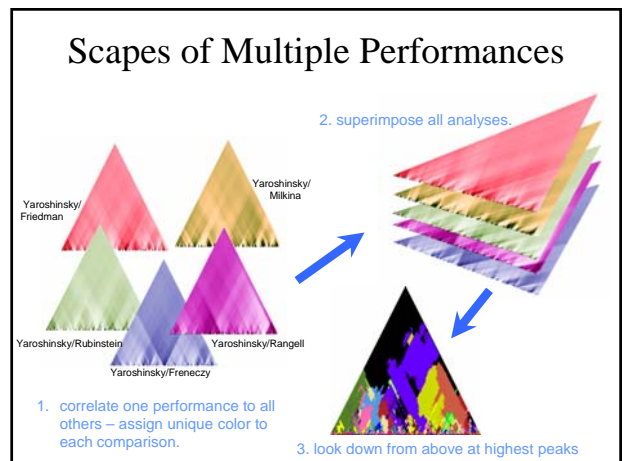
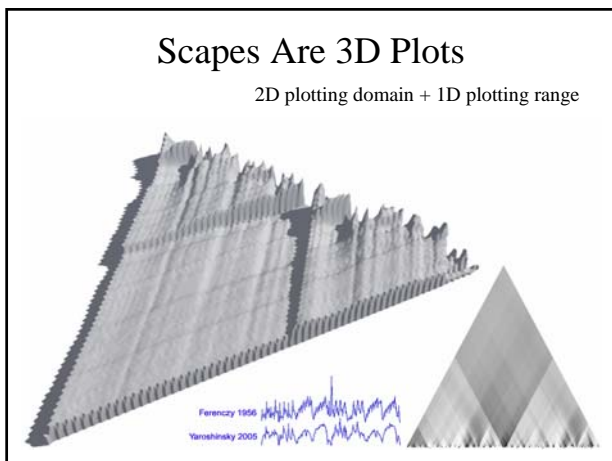
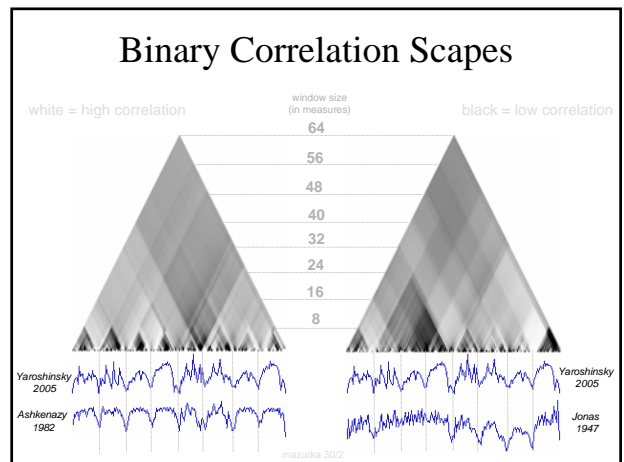
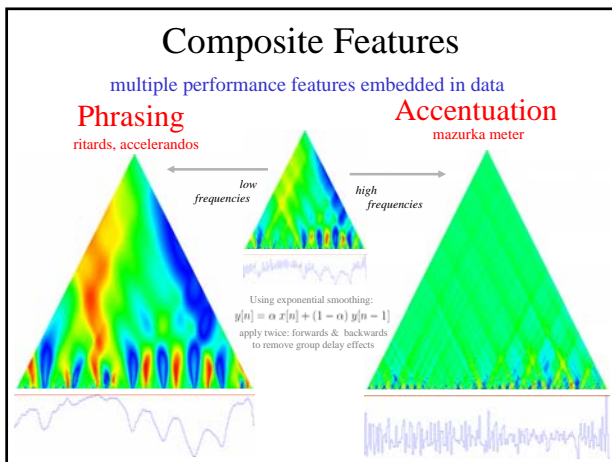
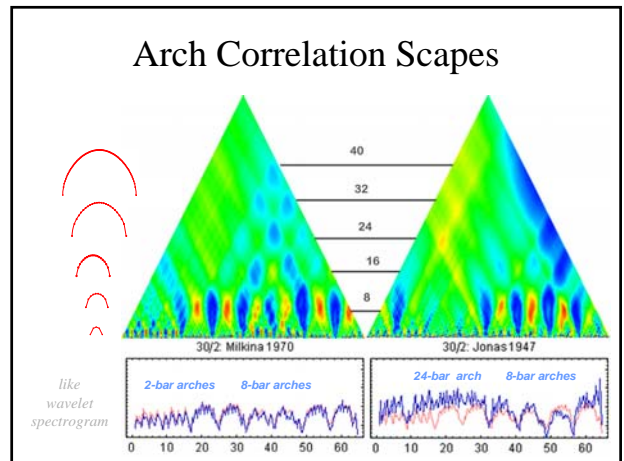
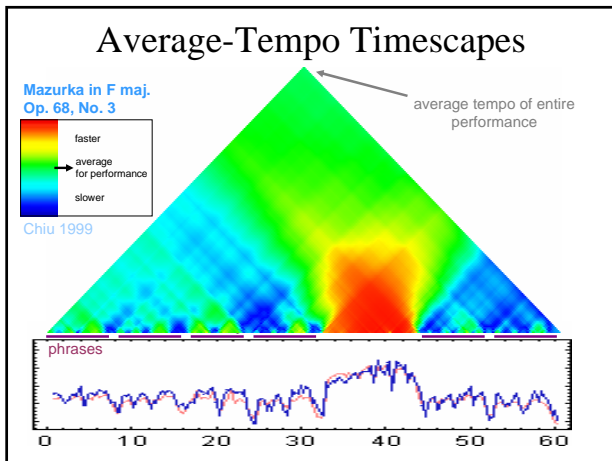
- 1-D data sequences chopped up to form a 2-D plot
- Example of a composition with 6 beats at tempos A, B, C, D, E, and F:



Scape Plotting Example

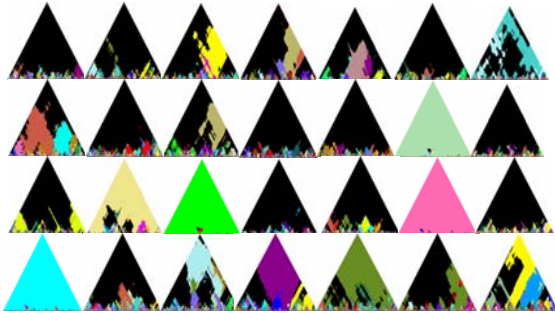
- Averaging in each cell with base sequence (7,8,2,5,8,4):





Perormance Correlation Scapes

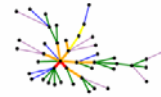
• Who is most similar to a particular performer at any given region in the music?



mazurka.org.uk/ana/pcor

mazurka.org.uk/ana/pcor-gbdyn

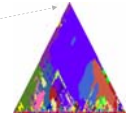
Maps and Scapes



Correlation maps give gross detail, like a real map:



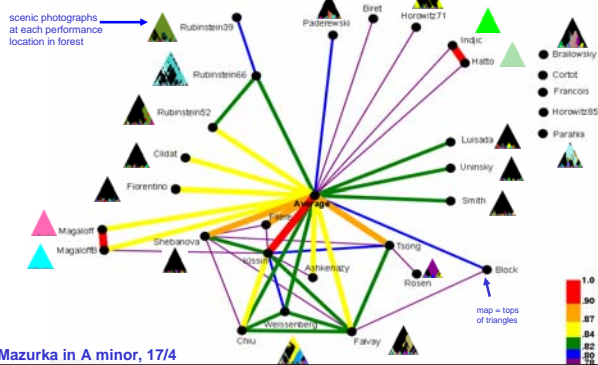
map points are tops of scapes



Correlation scapes give local details, like a photograph:



Map and Photos of the Forest



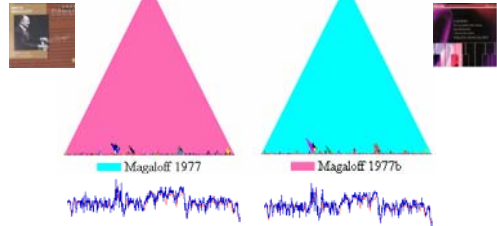
Mazurka in A minor, 17/4

Boring Timescape Pictures

Occasionally we get over-exposed photographs back from the store, and we usually have to throw them in the waste bin.

The same performance by Magaloff on two different CD re-releases:

Philips 456 898-2 mazurka 17/4 in A minor Philips 426 817/29-2

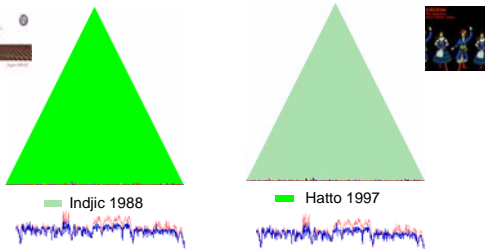


• Structures at bottoms due to errors in beat extraction, measuring limits in beat extraction, and correlation graininess.

Boring Timescape Pictures?

Two difference performances from two different performers on two different record labels from two different countries.

Calliope 3321 mazurka 17/4 in A minor Concert Artist 20012

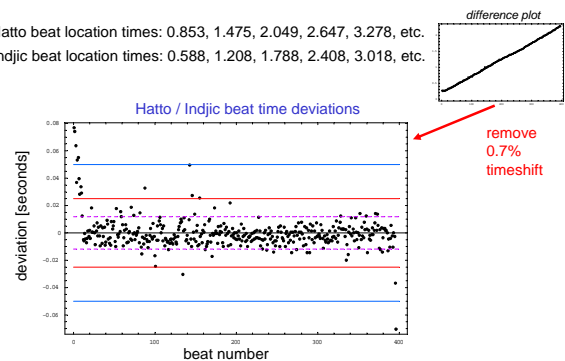


see: http://www.charm.rhul.ac.uk/content/contact/hatto_article.html

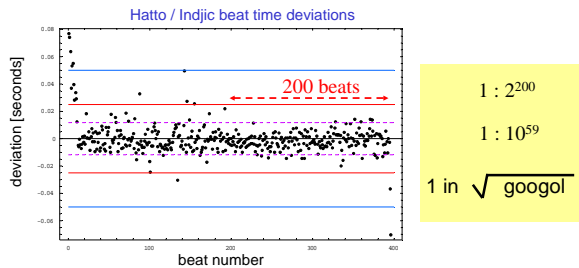
Beat-Event Timing Differences

Hatto beat location times: 0.853, 1.475, 2.049, 2.647, 3.278, etc.

Indjic beat location times: 0.588, 1.208, 1.788, 2.408, 3.018, etc.

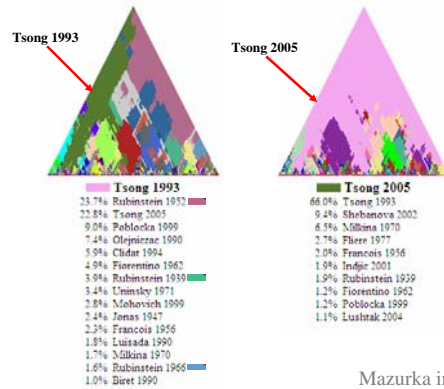


Timing Difference Probability



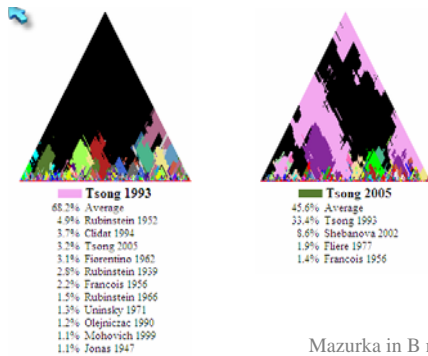
probability that same performer can produce a second performance so closely is equivalent to one atom out of an entire star.

Same Performer Over Time



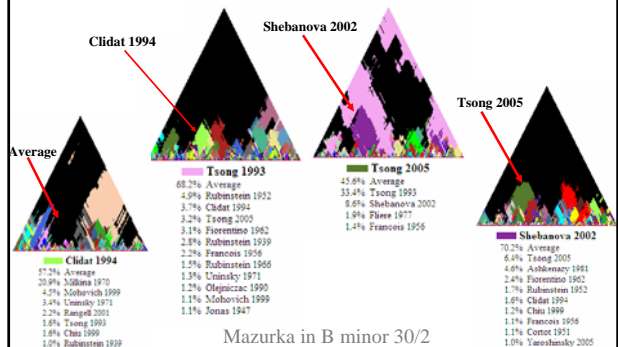
Mazurka in B minor 30/2

Including the Average



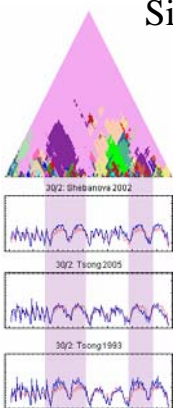
Mazurka in B minor 30/2

Mutual Best Matches



Mazurka in B minor 30/2

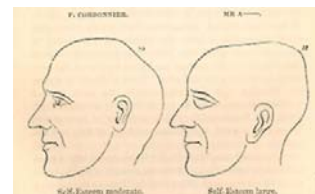
Significance of Similarity



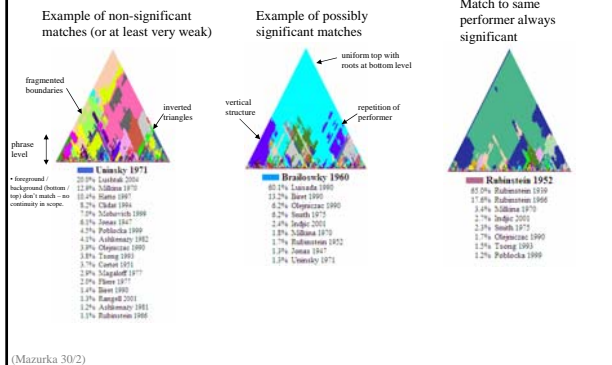
- Ts'ong 2005 performance matches best to Shebanova 2002 in 3 phrases when comparing 36 performances of mazurka 30/2.
- Is this a coincidence or not?
- Could ask the pianist (but might be problem in suggesting an answer beforehand). Also they might not remember or be totally conscious of the borrowing (such as accents in language). Or there could be a third performer between them.
- Ideally a model would be used to calculate a probability of significance.

Mazurka in B minor 30/2

Phrenology

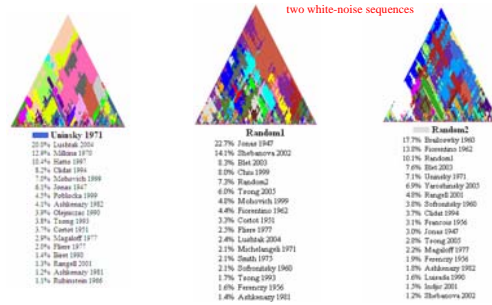


Significant or Not?



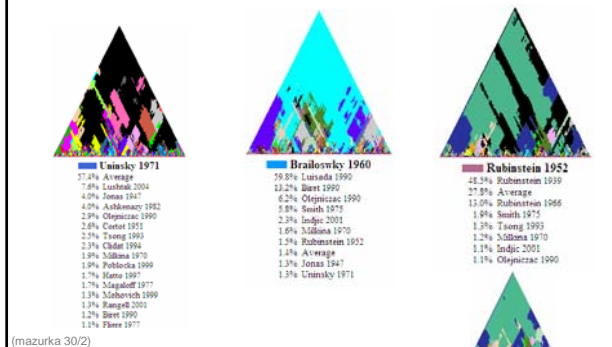
Purely Random Matching

- Plot has to show some match at all points...



Including Average Performance

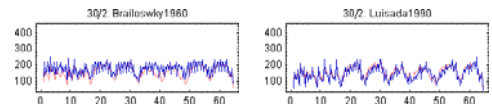
helps but does not solve significance question



What Is Significant?

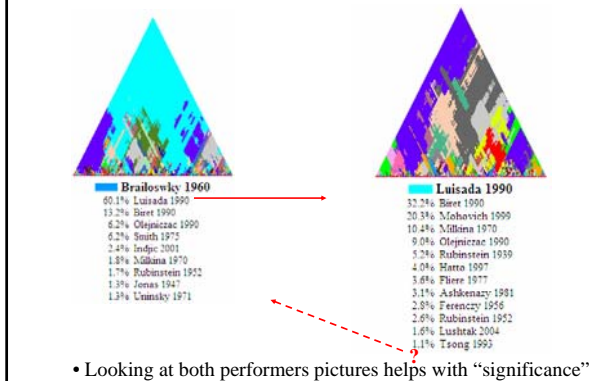


- No direct link found on web between Brailowsky and Luisada (such as teacher/student).
- Strong match in Brailowsky to Luisada probably due to large amount of mazurka metric pattern:



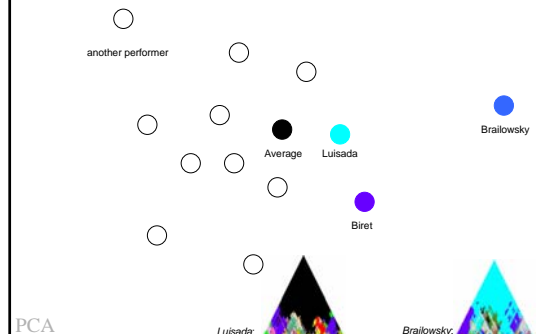
- Phrasing shapes very different, so match is both significant (high frequencies match) and not significant (low frequencies don't match).

Best Matching Not Mutual

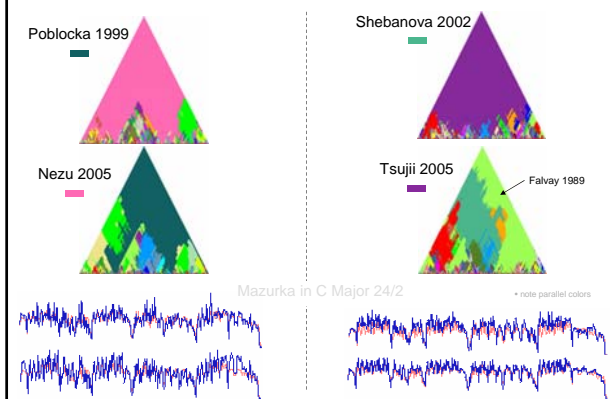


Performance Map Schematic

- Brailowsky has the strongest mazurka meter pattern
- Luisada has the second strongest mazurka meter pattern



Strong Interpretive Influences



Performance Ranking

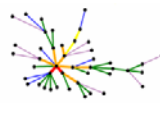
- Given a reference performance,
 - which other performance matches best
 - which other performance matches second best
 - ...
 - which other performance matches worst

0th-Order Similarity Rank

- Use large-scale correlation to order performances similarity to a target

performance	rank	correlation
Rubinstein 1952	0	1.00
Average	1	0.995
Rubinstein 1966	2	0.939
Milkins 1978	3	0.876
Poblacka 1999	4	0.836
Tsong 1993	5	0.814
Bilet 1998	6	0.807
Mohovich 1999	7	0.805
Hatto 1997	8	0.799
Indjic 2001	9	0.798
Rubinstein 1939	10	0.784
Shebanova 2002	11	0.783
Liszt 1978	12	0.767
Magaloff 1977	13	0.751
Oleynikov 1998	14	0.748
Bilet 2003	15	0.721
Cidat 1974	16	0.716
Rangell 2001	17	0.716
Liszt 2004	18	0.715
Chiu 1979	19	0.709
Tsong 2005	20	0.706
Smith 1975	21	0.695
Fliere 1977	22	0.678
Brailosky 1968	23	0.644
Ashkenazy 1982	24	0.629
Ashkenazy 1981	25	0.637
Coret 1951	26	0.625
Ferenczy 1956	27	0.628
Florentino 1962	28	0.615
Uninsky 1971	29	0.597
Francois 1956	30	0.577
Yaroshinsky 2005	31	0.565
Safranitsky 1968	32	0.558
Nicholas 1971	33	0.558
Jonas 1947	34	0.504
Random	35	0.182
Random	36	0.054
Random	37	-0.015

Performance maps used rank #1 data:

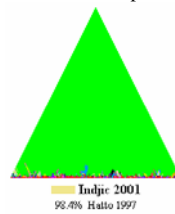


1st-Order Scape Rank

- Area represented by best choice in scape plot.



- Hatto effect causes problems:



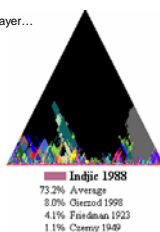
- Who is #2 for Indjic?

Scape Layers

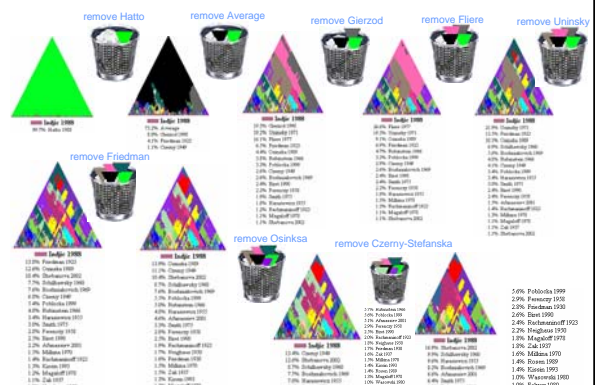
- Hatto takes up nearly 100% of Indjic's scape



- So remove Hatto layer...



Peeling the Layers



2nd-Order Scape Rank

- Rank score = area represented by best choice in scape plot, but peel away previous best choices.

performance	0-rank	1-rank	2-rank
Indic 2001	0	1.000	0
Hatto 1997	1	0.989	1
Ruseae	2	0.861	2
Publiska 1999	3	0.796	3
Rubinstein 1952	4	0.778	4
Milina 1970	5	0.791	5
Rubinstein 1977	12	0.714	13
Mohovich 1977	5	0.709	17
Lisada 1998	0	0.709	27
Biet 1980	0	0.702	4
Rubinstein 1966	9	0.700	21
Olejniczak 1998	11	0.700	7
Rubinstein 1982	26	0.623	23
Rubinstein 1981	2	0.620	9
Francis 1956	10	0.721	5
Hinsky 1971	17	0.697	28
Fliera 1977	24	0.646	26
Smith 1975	28	0.627	22
Clidat 1994	18	0.677	6
Isoog 1993	14	0.706	14
Shebanova 2002	13	0.707	32
Biet 2003	15	0.706	24
Lushak 2004	19	0.676	16
Chiou 1999	20	0.672	12
Magaloff 1977	16	0.698	20
Isoog 2005	20	0.698	20
Brailousky 1968	23	0.649	18
Ferenczy 1956	22	0.654	8
Rangel 2001	25	0.635	31
Flawitsky 1968	33	0.579	49
Corot 1951	29	0.685	25
Florentino 1962	31	0.615	21
Voronchinsky 2005	38	0.575	34
Michelangeoli 1971	32	0.584	39
Jonas 1947	34	0.415	34
Random1	35	0.182	35
Random2	35	0.047	35
Random3	37	0.008	37

• Still slightly sensitive to the Hatto effect

(mazurka 30/2)

3rd-Order Scape Rank

- Start with the 2nd-order rankings, selecting a cutoff point in the rankings to define a background noise level in the scape.
- Then one-by-one add each of the non-noise performances into the scape along with the background non-noise performances. Measure the area covered by the non-noise performance (only one non-noise performance at a time).

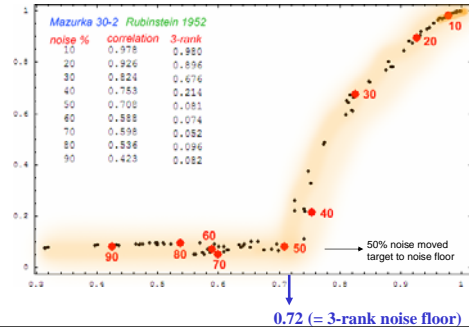
3rd-Order Rankings

performance	0-rank	1-rank	2-rank
Rubinstein 1952	0	1.000	0
Average	1	0.984	1
Rubinstein 1939	10	0.794	2
Rubinstein 1966	2	0.699	3
Milina 1970	3	0.875	4
Publiska 1999	4	0.835	5
Mohovich 1977	7	0.885	6
Isoog 1993	5	0.814	7
Biet 1980	6	0.886	8
Indic 2001	0	0.797	9
Hatto 1997	0	0.798	10
Olejniczak 1998	14	0.739	11
Shebanova 2002	11	0.702	12
Lisada 1998	12	0.766	13
Magaloff 1977	13	0.758	14
Lushak 2004	18	0.715	15
Isoog 2005	20	0.785	16
Biet 2003	15	0.729	17
Ferenczy 1956	27	0.627	18
Rubinstein 1981	25	0.637	19
Fliera 1977	22	0.677	20
Florentino 1962	28	0.615	21
Clidat 1994	16	0.716	22
Smith 1975	23	0.694	23
Rangel 2001	17	0.715	24
Rubinstein 1982	24	0.641	25
Chiou 1999	19	0.788	26
Voronchinsky 2005	31	0.584	27
Corot 1951	29	0.637	28
Brailousky 1968	23	0.643	29
Hinsky 1971	29	0.596	30
Michelangeoli 1971	32	0.558	31
Sofronitsky 1982	32	0.552	32
Francis 1956	30	0.577	33
Jonas 1947	34	0.415	34
Random1	35	0.182	35
Random2	36	0.054	36
Random3	37	0.016	37

→ 50% noise floor
(2-rankings below "noise floor")

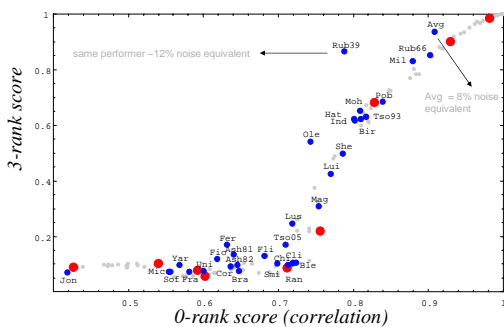
Proportional Noise

- Gradually add noise to target



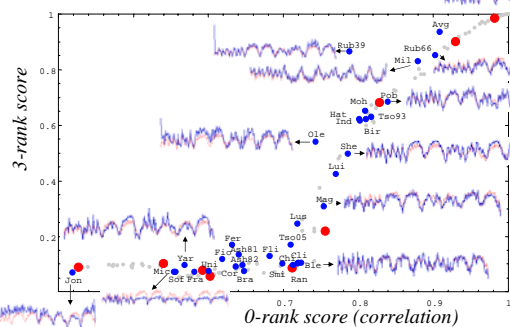
Real Data (1)

Mazurka 30/2
Target: Rubinstein 1952

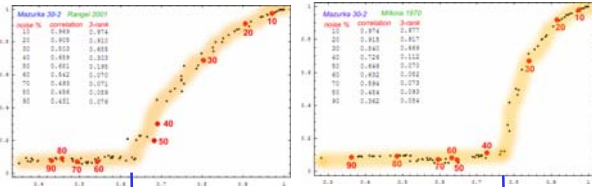


Real Data (2)

Mazurka 30/2
Target: Rubinstein 1952



Distance to the Noise Floor



0.62, 55% noise

38% noise, 0.77

- Metric for measuring individuality of interpretation?

- 3-Rank scores more absolute than correlation values
 - * noise floor is always about a 3-rank score of 10%
- 3-Rank scores less sensitive to local extrema

Cortot Performance Ranking

- Master class recording contains 48 out of 64 measures (75%)

Con. Artists Rankings



- 0-Rank:**
1. Average
 2. Rangell 01
 3. Milkina 70
 4. Mohovich 99
 5. Shebanova 02
 - ...
 32. Masterclass

- 3-Rank**
1. Average
 2. Rangell 01
 3. Mohovich 99
 4. Rubinstein 39
 5. Milkina 70
 - ...
 31. Masterclass

Masterclass Rankings



- 0-Rank:**
1. Poblocka 99
 2. Average
 3. Rubinstein 52
 4. Tsong 93
 5. Tsong 05
 - ...
 33. Con. Artist

- 3-Rank**
1. Average
 2. Rubinstein 52
 3. Luisada 90
 4. Poblocka 99
 5. Hatto 94
 - ...
 35. Con. Artist

Match to other Cortot
near bottom of rankings.

Match to other Cortot
near bottom of rankings.

(comparing 35 performances + average)