

# Exploring variations through computational analysis

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## **Possibilities of using computation**



Using computers changes or even challenges the practices of music analysis.

- Musical data can be analysed with greater precision.
- Greater quantities of music can be analysed.
- Using computers changes the questions asked in analysis.

Alan Marsden, "What was the question?": Music analysis and the computer', in Tim Crawford & Lorna Gibson, *Modern Methods for Musicology: Prospects, Proposals and Realities* (Ashgate, 2009).

## **Necessities of using computation**



Some music theory and analysis makes general claims about music.

- General claims require evidence and arguments of general validity.
- Empirically verifiable claims should be empirically verified.
- Cook & Clarke call for musicology to become a 'data rich' discipline (*Empirical Musicology* (OUP, 2004)).

## Validity requires

- No bias (objectivity)
- Sufficient evidence
- Precision of argument

Computers, suitably programmed and with suitable databases, deliver these. They are difficult to obtain by purely human means.

## How is a variation related to a theme?

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A general question about a kind of music.

Needs first to be framed more precisely:

• What properties does a variation share (or share more) with the theme of which it is a variation, but not share (or share less) with a different theme?

No bias

Selection of material on objective criteria

Sufficient evidence?

10 themes, 76 variations, but only four bars of each

Precise argument

- Computational comparison
- Mathematical analysis of results

## **Automatic Schenkerian reduction**



Previous work (Kassler, etc.) has shown the theoretical possibility of Schenkerian reduction by computer, but implementation is a complex problem.

AHRC-sponsored project to investigate Schenkerian reduction by computer.

- System capable of deriving a reduction from small extracts of keyboard music (c. 4-8 bars).
- For short themes with Ursatz, matches human-produced reductions moderately well.

Essential problem is that there are a vast number of possible reductions of even short extracts. Identifying 'the best' reduction is difficult.

## **Example (hand-made) reduction**





## **Formalisation of theory**



Structure

- binary trees
  - parallel congruent trees for different voices
  - can share parts of their structure
  - so properly a directed acyclic graph (DAG, digraph)
- Nodes = notes and rests
  - pitch, duration & tie
  - harmony (global key & metre)
  - no explicit voices
- Arcs = **`atomic reductions**'
  - one parent; two children
  - constraints on immediately preceding and following context
  - harmonic constraints
  - inheritance of harmony



## **Example atomic reductions**



Appoggiatura

- First child: no tie
- Second child: no tie; pitch one step above or below first child
- Parent: no tie; pitch equal to second child; harmony equal to second child's; pitch of second child consonant
- **Required pre-context: none**
- **Required post-context: none**

Neighbour Note

- First child: [no constraint]
- Second child: no tie
- Parent: tied if first child tied; pitch equal to first child; harmony equal to first child's; pitch of first child consonant
- **Required pre-context: none**
- **Required post-context: note one step above or below second child**

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## **Atomic reductions**

No context constraints:

- hold (tied)
- repetition
- shortening (followed by rest)
- delay (preceded by rest)
- appoggiatura
- consonant skip 1 (first pitch = parent)
- consonant skip 2 (second pitch = parent)
- interruption (I-V)

Constraint on following context:

- anticipation
- neighbour note (incomplete; resolves to following context note)

Constraint on preceding context:

suspension

Other reductions can be constructed from combinations of these

Discussion and detail of formalisation in

Alan Marsden, 'Generative Structural Representation of Tonal Music', *Journal* of New Music Research, 34 (2005), 409-428

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## **Computational process**



Basic process:

- **1.** Divide the score into a sequence of `segments'.
  - each segment covers a span where no note begins or ends
- 2. For each pair of segments, compute the possible reductions, deriving new segments.
  - do this recursively for pairs involving derived segments also
- **3.** Select only analyses which contain an Ursatz.
- 4. Select the best alternative.

The number of alternatives is far too great for a naive process.

- number of possibilities related to n!
  - (*n* = number of segments in the piece)
  - $n! = n \times (n-1) \times (n-2) \times ... \times 1$

## **Chart parser; CYK algorithm**



Instead of making a set of analyses, make a chart of possible reductions at each point, from which complete analyses can be extracted.

- Triangular matrix of cells
  - bottom row contains segments of the 'surface'
  - higher rows contain derived segments spanning 2, 3 ... surface segments
  - top row has a single cell spanning the entire piece



- Lower computational complexity
  - in principle, cubic (n<sup>3</sup>) instead of factorial

## **Up-Down Process**

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`Up':

- Derive segments
- Record best score for each possibility
- Record possible Ursatz membership for each segment

'Down':

- Prune segments which have no parent
- Prune segments which cannot be part of an Ursatz or be reduced to a member of an Ursatz
- Select best-scoring analysis
  - best-first search









Step 2 3 new segments: 67% D5 G3 B4, G3 D5, 67% В4 G3 B4 D5 100% G3 E5 2 C5 4 D5 1 B4 1 FЗ G3 С4 С4

Step 3a No new segments

	67% D5		
	67% B4		
	100% G3		
E5 2	D5 1	B4 1	C5 4
C4	F3	G3	C4



Step 3b 2 new segments: C4 C5, G3 C4 C5

	100% C5		
	100% C4		
	50% G3		
	67% D5		
	67% B4		
	100% G3		
E5 2	D5 1	B4 1	C5 4
C4	F3	G3	C4

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Full table

63%	E5						
75%	С5						
75%	C4						
63%	G3						
100%	E5	100%	С5				
100%	C3	100%	C4				
50%	G3	50%	G3				
		67%	D5				
		67%	В4				
		100%	G3				
E5 2		D5 1		В4	1	C5	4
C4		F3		G3		C4	

## **Selection of Best Analysis**



Prune and select best scoring



## Weights

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### To find a good analysis

- **1.** Select higher level pitches which are more often present in the surface.
- **2.** Avoid splitting and joining of voices.
- **3.** Select reductions with small intervals between notes reduced together.
- 4. Reduce segments of approximately equal duration together.
- **5.** Avoid reductions which create syncopations at higher levels.
- 6. Avoid reducing a shorter segment with a following longer segment.
- **7.** Prefer reductions with more tonic and dominant harmony.
- 8. Avoid reductions where a note is followed by a rest.
- 9. Prefer reductions where higher level harmonies are more often consonant with the surface.

## Automatically derived best-scoring analysis $LIC\Lambda$



## **Exploring variations**



**Hypothesis:** Variations and themes share a common structure.

- The reduction of a variation will match the reduction of the theme, at least at higher levels.
- The match will be greater than a match based on the surface alone.

**Method:** Compare how much variations match their theme with how much they match unrelated themes.

- Take corresponding extracts of variations and themes.
  - First four bars of all Mozart piano variations in simple triple and duple metres, avoiding variations in a different key or metre, and two juvenile pieces.
- Match each variation with each theme.
  - match surface with surface and best reduction of theme with reduction matrix of variation
- Test for a greater degree of match with the correct theme.





Variations

Theme



## **Examples of materials**

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## **Matching methods**



All combinations of

- Pitch matching: pitches/pitch classes
- Pitches from: full texture/melody+bass/melody/bass
- Voice must match (melody, middle or bass): yes/no
- Match tied notes: yes/no
- Weight by metre/reduction level: yes/no
- Limit by parent match (reduction only): yes/no
- Value recorded:
  - surface: proportion of span/present in span/present in bar
  - reduction (from multiple possible segment): maximum/simple average/score-weighted average

384 different combinations for surface matches.1024 different combinations for reduction matches.

## **Surface-matching example**



#### K.265 theme with K.265 variation 3



Blue: portions of theme notes matched with variation notes Red: portions of theme notes not matched with variation notes Yellow: variation notes matched with theme notes Green: variation notes not matched with theme notes

## **Reduction-matching example**



### K.265 theme and K.265 variation 3



Blue: portions of theme notes matched with variation notes in *some* corresponding segment

Red: portions of theme notes not matched with variation notes

• Matches are also made at higher levels of reduction. In this case higher levels match perfectly.

## Method



- **1.** All themes and variations were transposed to F major.
- 2. Every theme was compared to every variation in the same metre with each method and the degree of match recorded.
- **3.** For each theme and each method, a maximum possible F-measure was calculated.
  - Select a threshold of match.
  - Count how many variations of this theme have a degree of match to the theme greater than this threshold (tp), and how many less (fn).
  - Count how many variations of other themes have a degree of match to the theme greater than this threshold (fp).
  - F-measure is 2 \* tp/(2 \* tp + fn + fp).
  - Test for all possible thresholds.
- 4. High F-measure indicates a method which tests what a theme and its variations have in common

## **Example results**



## For theme of K.265 Reduction-based result



#### Surface-based result





Surface methods	Average F-measure	Reduction methods	Average F-measure	
Best	0.867	Best	0.842	
Average	0.776	Average	0.748	
Worst	0.540	Worst	0.671	

- **1.** Contrary to the hypothesis, variations and themes do not appear to be more similar in their reductions than at the surface.
- 2. Best surface-based method matches pitch classes rather than pitches, matches notes in their respective voices, includes tied notes, weights by duration, and measures the proportion of span which matches.
- 3. Best reduction-based method matches pitch classes in melody and bass in their voices, ignores tied notes, weights by duration, and measures the maximum match in alternative segments.

## **Going about computational analysis**

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- Write your own software.
  - requires expertise
  - very time-consuming
- Use an existing package
  - Sonic Visualiser for analysis from audio
  - Humdrum for score-based analysis
  - not many packages
  - still require some expertise
- Use general-purpose software
  - Excel or similar
  - Matlab or similar
- Get someone else to write the software for you
  - computer-science student as project
  - collaborate with a computer scientist
  - software service such as centre shortly to be established at QMUL Centre for Digital Music

## **The computational approach**



- Precise definition of data
- Unambiguous and tractable analysis processes
  - what is to be found out
  - how to find it out
- Rigorous assessment of results
  - mathematical analysis
  - tests for significance

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