

# **FANTASTIC: A Feature Analysis Toolbox for corpus-based cognitive research on the perception of popular music**

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# Summary of a Research Project

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*M4S: Modelling Music Memory and the Perception of Melodic Similarity (2006-2009)*

**Question:** How do Western listeners perceive melody?

**Domain:** Western commercial pop music

**Method:** Computational modelling

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# Outline

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## 1. Results

- Music Cognition
- Popular Music Research

## 2. Methods

- Computing Features with FANTASTIC
- Modelling Music Knowledge from a Corpus

## 3. Background

- Similar Approaches/Systems
  - Questions to be addressed
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# Results: Music Cognition I

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Memory for Melodies:

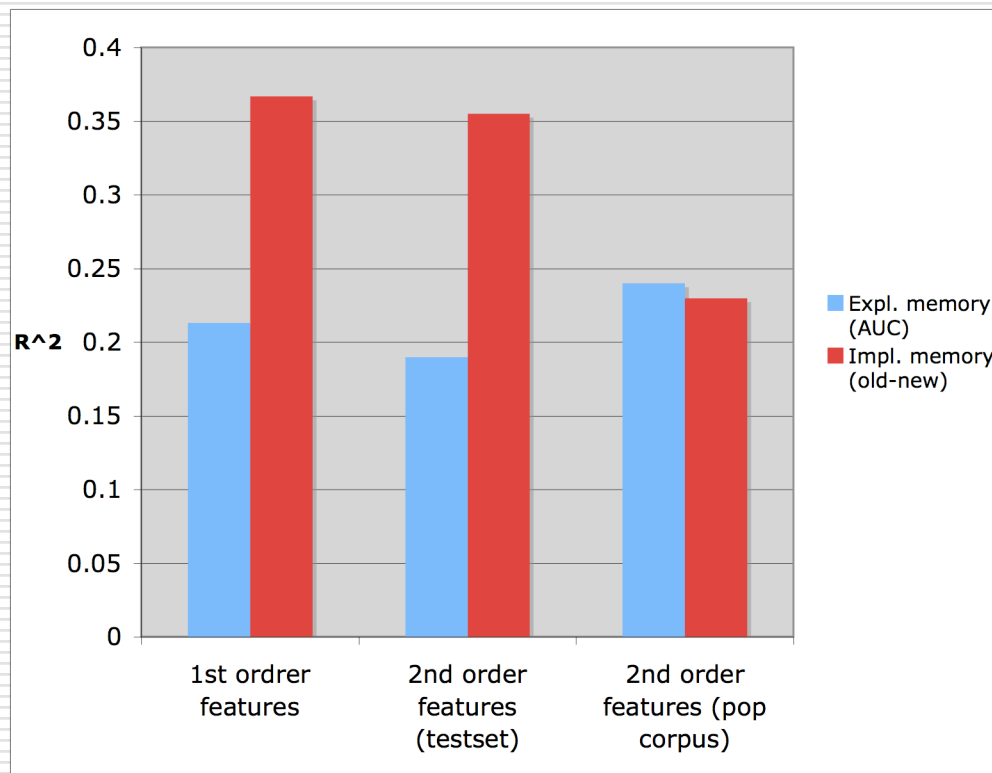
*Are there structural features that make melodies more memorable?*

*How are listeners using musical knowledge to perform implicit and explicit memory tasks?*

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# Results: Music Cognition I

Modelling explicit and implicit memory performance in a recognition paradigm (Müllensiefen, Halpern & Wiggins, in prep.)



## Results:

- Memory performance is partially explained by musical features
- Implicit memory is better explained by raw features or local context
- Explicit memory draws on domain knowledge and features that are distinctive wrt corpus

# Results: Music Cognition II

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Montreal Battery of Amusia, MBEA, (Peretz et al., 2003):

*What makes some test items more difficult than others?*

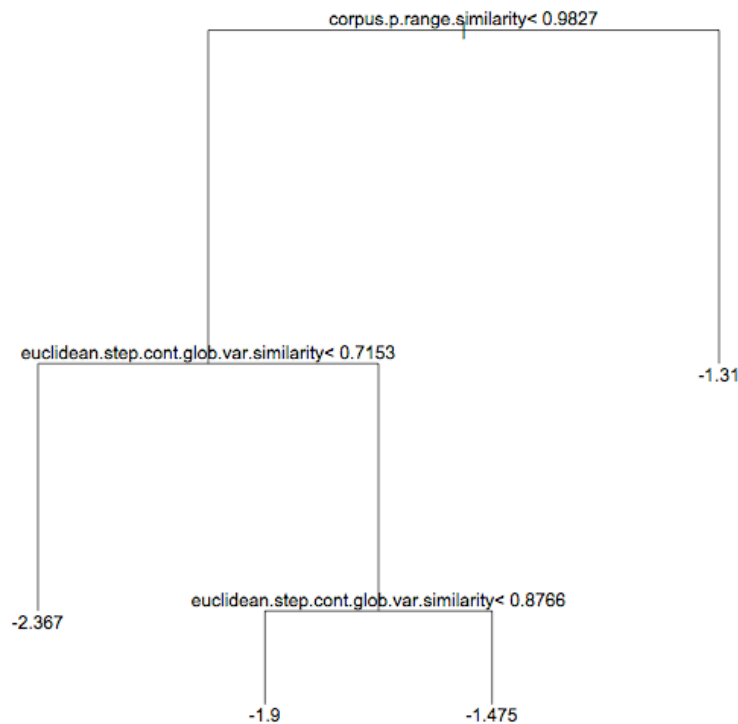
*What information do subjects actually use to process tasks?*

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# Results: Music Cognition II

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Modelling item difficulty in MBEA (Stewart, Müllensiefen & Cooper, in prep)



## Results:

- 70-80% of item difficulty can be explained with as few as three musical features
  - Relation between item difficulty and features is often non-linear
  - Some subtests don't measure what they are believed to measure (e.g. scale)
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# Results: Pop Music Research I

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Court cases of music plagiarism:

*Are court decisions predictable from melodic structures?*

*What musical information is used in court decisions?*

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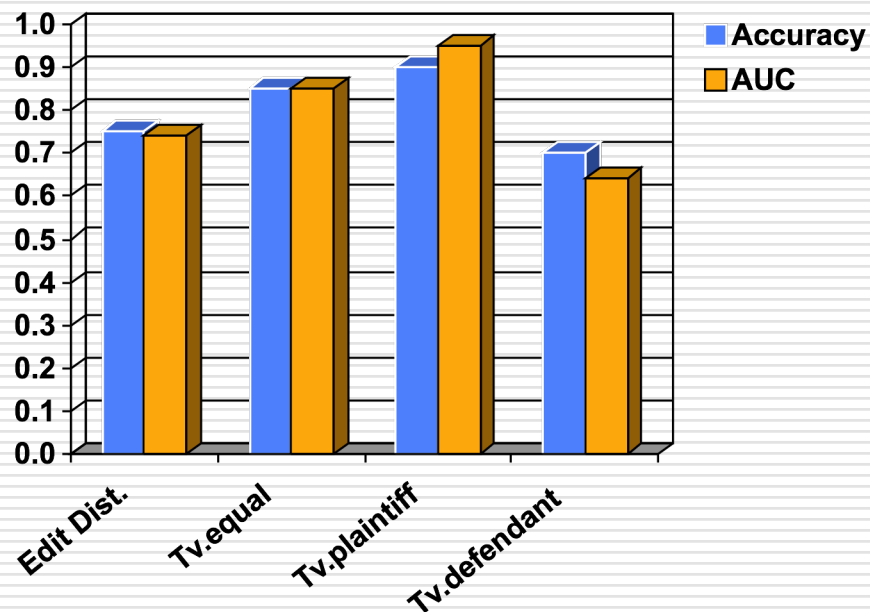


# Results: Pop Music Research I

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## Model court decisions on melody plagiarism

(Müllensiefen & Pendzich, 2009)



### Results:

- Court decisions can be closely related to melodic similarity
  - Plaintiff's song is often frame of reference
  - Statistical information about commonness of melodic elements is important
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# Results: Pop Music Research II

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Melodic structure and popularity:

*Does popularity correlate with certain structural features of a tune?*

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# Results: Pop Music Research II

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Identify features of commercially successful songs on *Revolver*  
(Kopiez & Müllensiefen, 2008)

Criterion for commercial success: Entered charts as cover version (yes/no)

$$p(\text{chart\_entry} = 1) = \frac{1}{1 + e^{-(772.4 + 141.2 \cdot \text{pitch\_range} - 4731.3 \cdot \text{pitch\_entropy})}}$$

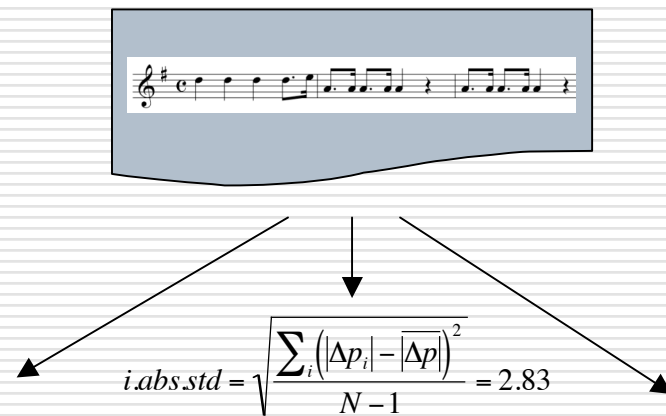
Results:

- 2 features (pitch range and entropy) are sufficient for fully accurate classification into successful / unsuccessful songs
  - Plausible interpretation as compositional exercise: *Invent a chorus melody such that it has a large range and uses only few pitches much more frequently than the majority of its pitches*
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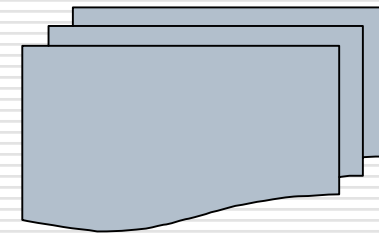
# Method

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## Two Components



Feature Computation



Knowledge from a large  
corpus of music

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# Method: Feature Computation

Pre-requisite: Transformation from notes to numbers



L	Typ	Anfang	Ende	Länge	Wert 1	Wert 2
	♪ Note	1. 1. 1. 5	1. 1. 4.102	0. 3. 97	D4	69
	♪ Note	1. 2. 1. 14	1. 2. 4. 98	0. 3. 84	D4	65
	♪ Note	1. 3. 1. 10	1. 3. 4. 94	0. 3. 84	D4	67
	♪ Note	1. 4. 1. 6	1. 4. 3. 87	0. 2. 81	D4	65
	♪ Note	1. 4. 3.119	1. 4. 4. 93	0. 0. 94	E4	64
	♪ Note	2. 1. 1. 2	2. 1. 3.101	0. 2. 99	A3	69
	♪ Note	2. 1. 4. 13	2. 1. 4.107	0. 0. 94	A3	64
	♪ Note	2. 2. 1. 14	2. 2. 3. 95	0. 2. 81	A3	65
	♪ Note	2. 2. 4. 7	2. 2. 4.103	0. 0. 96	A3	64
	♪ Note	2. 3. 1. 10	2. 3. 4. 94	0. 3. 84	A3	67
	♪ Note	3. 1. 1. 2	3. 1. 3.101	0. 2. 99	A3	69
	♪ Note	3. 1. 4. 13	3. 1. 4.107	0. 0. 94	A3	64
	♪ Note	3. 2. 1. 14	3. 2. 3. 95	0. 2. 81	A3	65
	♪ Note	3. 2. 4. 7	3. 2. 4.103	0. 0. 96	A3	64
	♪ Note	3. 3. 1. 10	3. 3. 4. 94	0. 3. 84	A3	67
	♪ Note	4. 1. 1. 2	4. 1. 3.101	0. 2. 99	G3	69
	♪ Note	4. 1. 4. 13	4. 1. 4.107	0. 0. 94	G3	64
	♪ Note	4. 2. 1. 14	4. 2. 3. 95	0. 2. 81	G3	65
	♪ Note	4. 2. 4. 7	4. 2. 4.103	0. 0. 96	G3	64
	♪ Note	4. 3. 1. 10	4. 3. 4. 94	0. 3. 84	G3	67

# Method: Summary Features

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Cognitive Hypothesis: Listeners abstract summary representation of short melodies during listening

Format: Value that represents particular aspect of melody

Ex. 1: Pitch range (*p.range*):

$$p.range = \max(p) - \min(p)$$

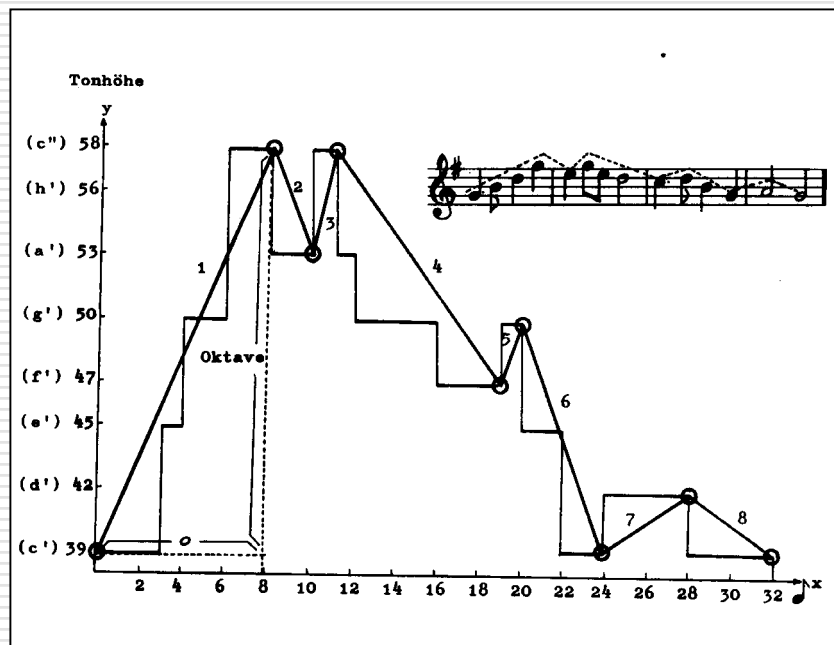
Ex. 2: Standard deviation of absolute intervals (*i.abs.std*):

$$i.abs.std = \sqrt{\frac{\sum_i (|\Delta p_i| - \overline{|\Delta p|})^2}{N - 1}}$$

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# Method: Summary Features

Ex. 3: Relative number of direction changes in interpolated contour representation (int.cont.dir.changes)



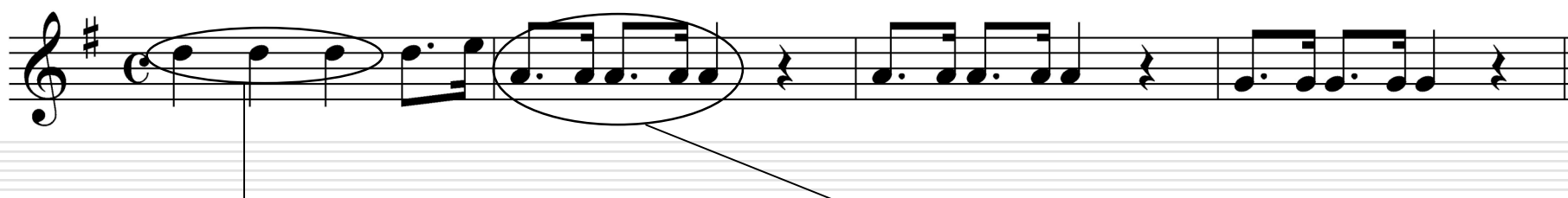
$$\text{int. cont. dir. changes} = \frac{\sum_{\text{sgn}(x_i) \neq \text{sgn}(x_{i+1})} 1}{\sum_{x_i \neq x_{i+1}} 1}$$

# Method: m-type Features

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Cognitive Hypothesis: Listeners use literal representation of short subsequences of melody for processing

Format of m-type: String of digits (similar to “word type” in linguistics)



The image shows a musical staff in treble clef with a key signature of one sharp (F#). The melody consists of several measures. Two specific subsequences are circled in red. The first circle highlights a two-note sequence: a quarter note on G4 followed by a quarter note on A4. The second circle highlights a four-note sequence: a quarter note on G4, a quarter note on A4, a quarter note on B4, and a quarter note on C5. Below the staff, two boxes provide the corresponding m-type feature strings for these subsequences.

m-type of length 2:  
“s1e\_s1e”

m-type of length 4:  
“s1q\_s1l\_s1q\_s1l”

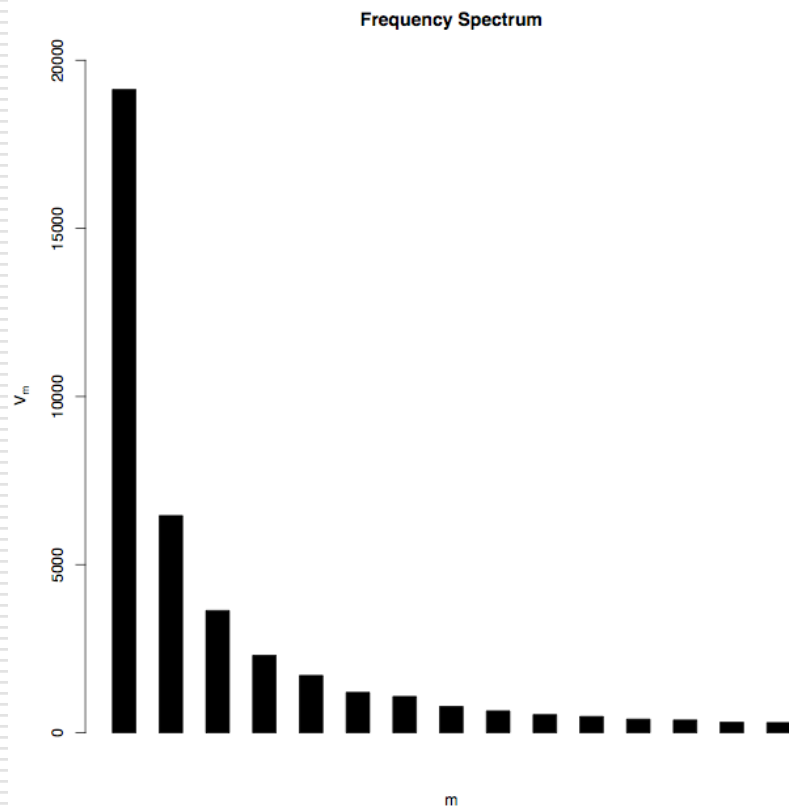
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# Method: m-type Features

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Format of m-type feature: Number that represents distribution of m-types in melody



$$\text{mean Honores } H = \frac{1}{|n|} \cdot 100 \cdot \frac{\log N}{1.01 - \frac{V(1,N)}{V(N)}}$$

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# Method: M4S publications on features

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- Melodic Contour (Müllensiefen, Bonometti, Stewart & Wiggins, 2009; Frieler, Müllensiefen & Riedemann, in press; Müllensiefen & Wiggins, under review)
  - Phrase segmentation (Pearce, Müllensiefen & Wiggins, 2008; accepted)
  - Harmonic content (Mauch, Müllensiefen, Dixon & Wiggins, 2008; Rhodes, Lewis & Müllensiefen, 2007)
  - Melodic accent structure (Pfleiderer & Müllensiefen, 2006; Müllensiefen, Pfleiderer & Frieler, 2009)
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# Method: Using a music corpus

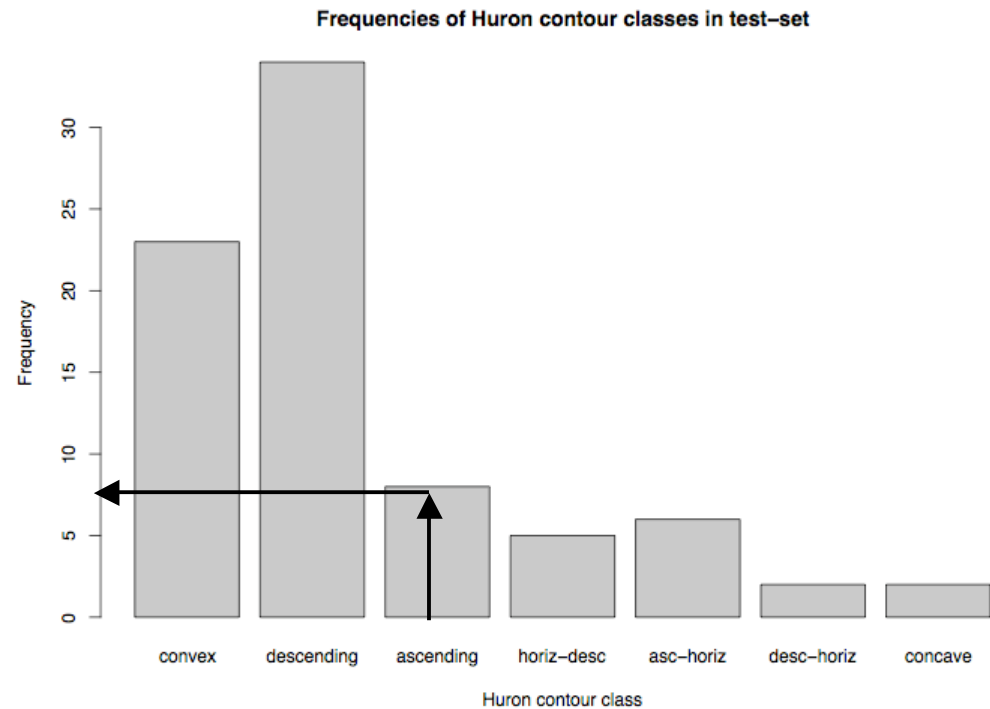
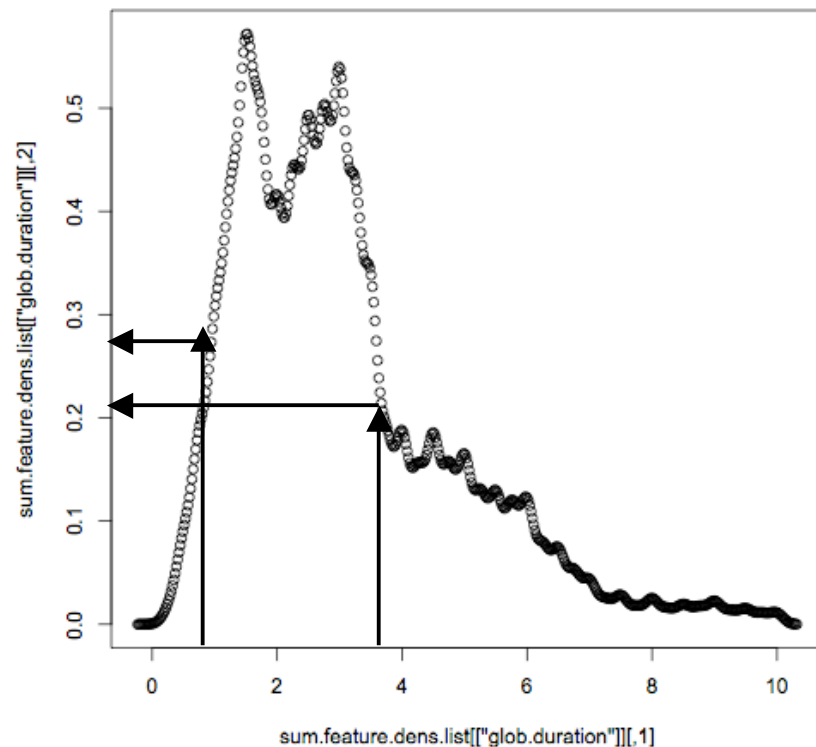
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The M4S Corpus of Popular Music (Müllensiefen, Wiggins & Lewis, 2008):

- ❑ 14,067 high-quality MIDI transcriptions
  - ❑ Representative sample of commercial pop songs from 1950 - 2006
  - ❑ Complete compositional structure (all melodies, harmonies, rhythms, instrumental parts, lyrics)
  - ❑ Some performance information (MIDI patches, some expressive timing)
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# Using a music corpus: 2nd order summary features

Cognitive Hypothesis: Listeners encode commonness of feature value  
Method: Replacing feature values by their relative frequencies



# Using a music corpus: 2nd order m-type features

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Cognitive Hypothesis: Listeners are sensitive to commonness of m-types  
Method: Use frequency information on m-types from large corpus

Example: Normalised distance of m-type frequencies in melody and corpus (*mtcf.norm.log.dist*)

=> measures whether uncommon m-types are used rather frequently in melody

$$mtcf \ .norm.log \ .dist = \frac{\sum_{\tau_i \in m} |TF'_{\tau_i} - DF'_{\tau_i}|}{|TF_{\tau \in m}|}$$

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# Method: Summary

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Feature ANalysis Technology Accessing SStatistics In a Corpus:

## *FANTASTIC*

- ❑ Open source tool box for computational analysis of melodies\*
- ❑ 58 features currently implemented
- ❑ Ideas from: Descriptive statistics, music theory, music cognition, computational linguistics, music information retrieval
- ❑ 2 feature categories: Summary features and m-type features
- ❑ Context modelling via integration of corpus: 2nd order features

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\* <http://www.doc.gold.ac.uk/isms/m4s/?page=Software%20and%20Documentation>

# Background: Similar approaches

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## Folk Song Research / Ethnomusicology

- Bartók (1936), Bartók & Lord (1951)
- Lomax (1977)
- Steinbeck (1982)
- Jesser (1992)
- Sagrillo (1999)

## Popular Music Research

- Moore (2006)
- Kramarz (2006)
- Furnes (2006)
- Riedemann (in prep.)

## Computational / Cognitive Musicology

- Eerola et al. (2001, 2007)
  - McCay (2005)
  - Huron (2006)
  - Frieler (2008)
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# Background: Questions to be addressed

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## **Popular Music Research**

Questions: How does melodic structure relate to

- Popularity and selection processes
- Style
- Transmission processes
- Specific types of behaviour (e.g. singalongability)
- Value attribution (originality, creativity)

## **Music Cognition Research**

Questions: How does melodic structure relate to

- Memory performance and memory errors
  - Similarity judgements
  - Expectancy
  - Preference / aesthetic judgements
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