



音樂搜尋與分析

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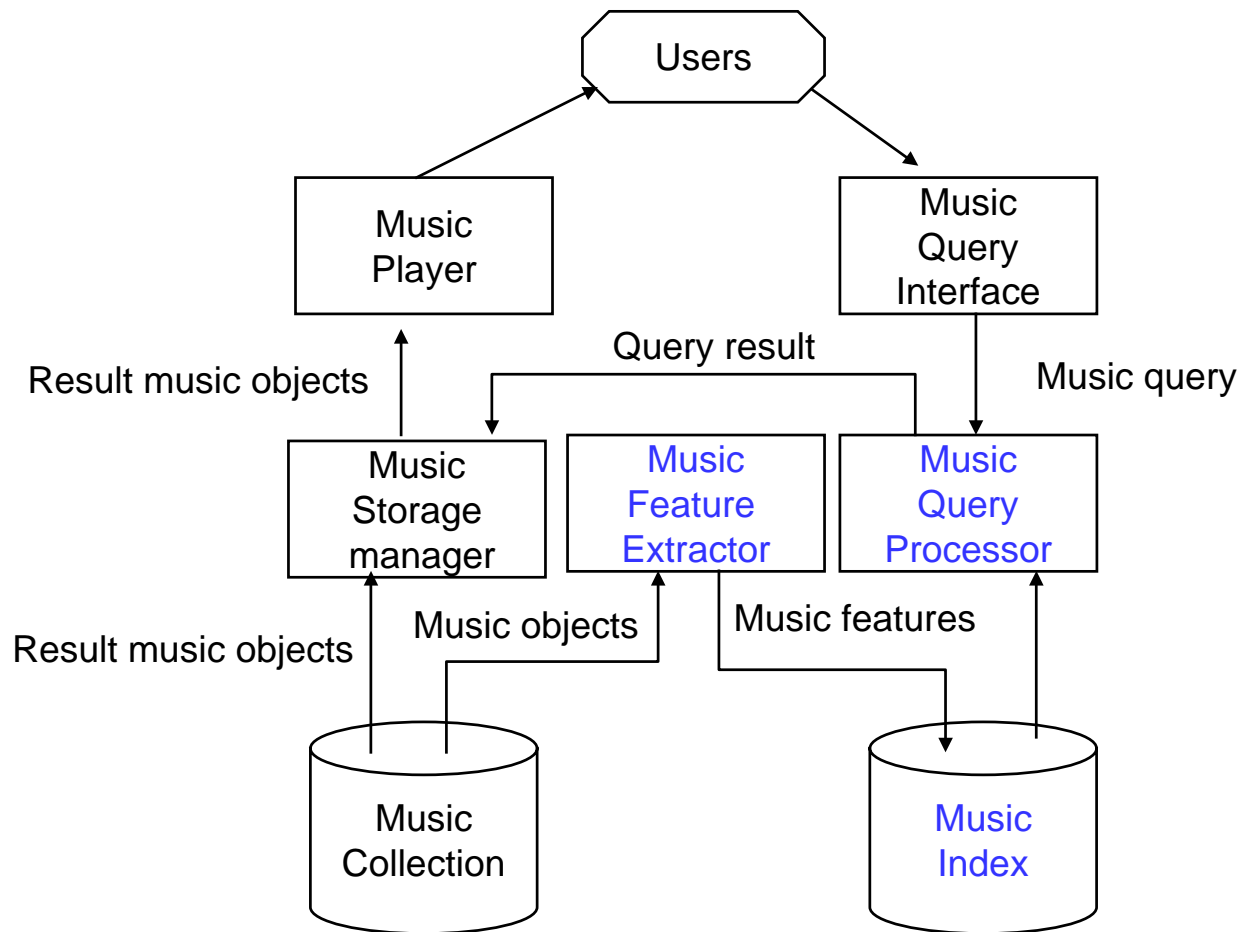
第五屆數位典藏技術研討會

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Outline

- Music Retrieval
 - Architecture for music retrieval
 - Music representations
 - Query processing
 - Indexing
 - System Evaluation
- Music Analysis
 - Repeating pattern discovery
 - Phrase extraction
 - Music classification
 - Music recommendation
- Future Research Directions

Architecture for Music Retrieval



Styles of Music Composition

■ Monophony

- Monophonic music has at most one note playing at any given time; before a new note starts the previous note must have ended

■ Polyphony

- Polyphonic music has no such restrictions. Any note or set of notes may begin before any previous note or set of notes has ended

Monophony Representations

- Absolute measure

- Absolute pitch

- C5 C5 D5 A5 G5 G5 G5 F5 G5

- Absolute duration

- 1 1 1 1 1 0.5 0.5 1 1

- Absolute pitch and duration

- (C5,1)(C5,1)(D5,1)(A5,1)(G5,1)(G5,0.5)(G5,0.5)(F5,1)(G5,1)

- Relative measure

- Contour (in semitones)

- 0 +2 +7 -2 0 0 -2 +2

- IOI (Inter onset interval) ratio

- 1 1 1 1 0.5 1 2 1

- Contour and IOI ratio

- (0,1)(+2,1)(+7,1)(-2,1)(0,0.5)(0,1)(-2,2)(+2,1)



Polyphony Representations



- All information preservation
 - Keep all information of absolute pitch and duration (start_time, pitch, duration)
 - (1,C5,1)(2,C5,1)(3,D5,1)(3,A5,1)(4,F5,4)(5,C6,1)(6,G5,0.5)(6.5,G5,0.5)...
- Relative note representation
 - Record difference of start times and contour (ignore duration)
 - (1,0)(1,+2)(0,+7)(1,-4)...
- Monophonic reduction
 - Select one note at every time step (main melody selection)
 - (C5,1)(C5,1)(A5,1)(F5,1)(C6,1)...

Music Representation - Theme

- Theme
 - A short tune that is repeated or developed in a piece of music
- A small part of a musical work
 - Efficient retrieval
- A highly semantic representation
 - Effective retrieval
- Automatic theme extraction
 - Exact repeating patterns
 - Approximate repeating patterns

Query Processing

- String Matching Algorithms
 - Exact string matching
 - Brute-force method
 - KMP algorithm
 - Boyer-Moore algorithm
 - Shift-Or algorithm
 - Partial string matching
 - Shift-Or algorithm
 - Approximate string matching
 - Edit distance
 - Dynamic programming
 - Candidate pruning [Liu, Wu, Chen, ACM MSJ]



Indexing

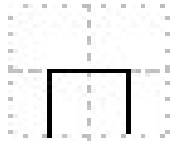
- Tree-based index (Suffix tree)
- List-based index (1D-list)
- N-gram index

Tree-Based Index

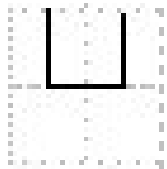
- [Chen, et al., ICME'00]
- Music objects are coded as strings of music segments
 - Four segment types to model music contour
 - Pitch and duration are considered
- Index structures
 - Augmented suffix tree
- Both incipit/partial and exact/approximate matching can be handled

Tree-Based Index (Cont.)

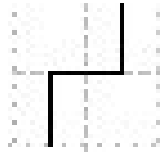
Four segment types



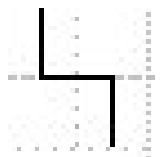
type A



type B



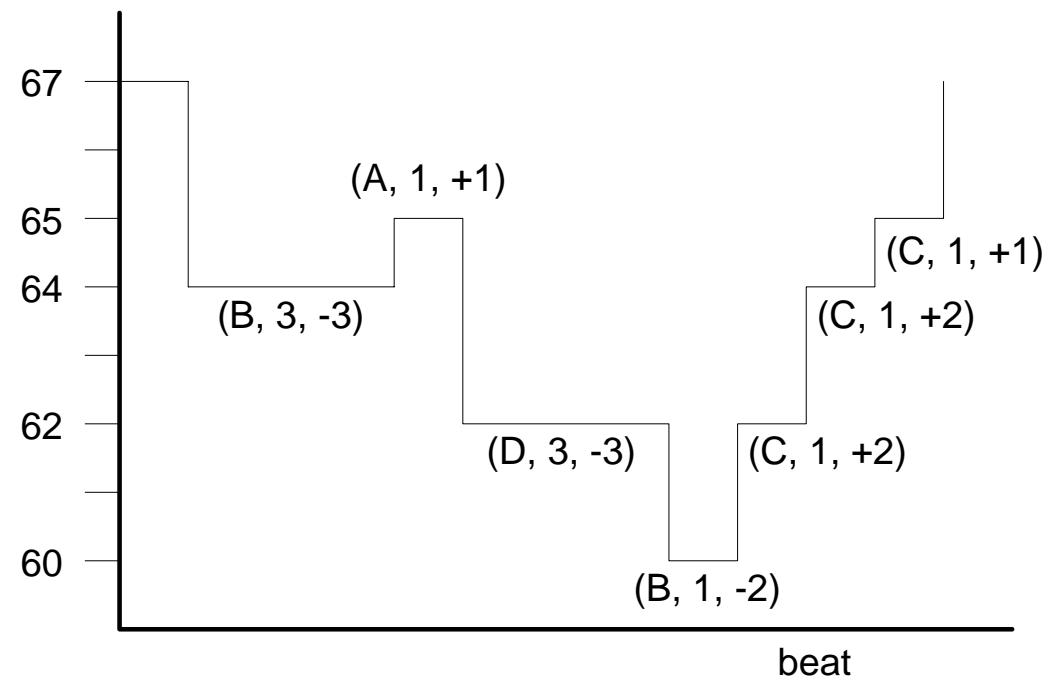
type C



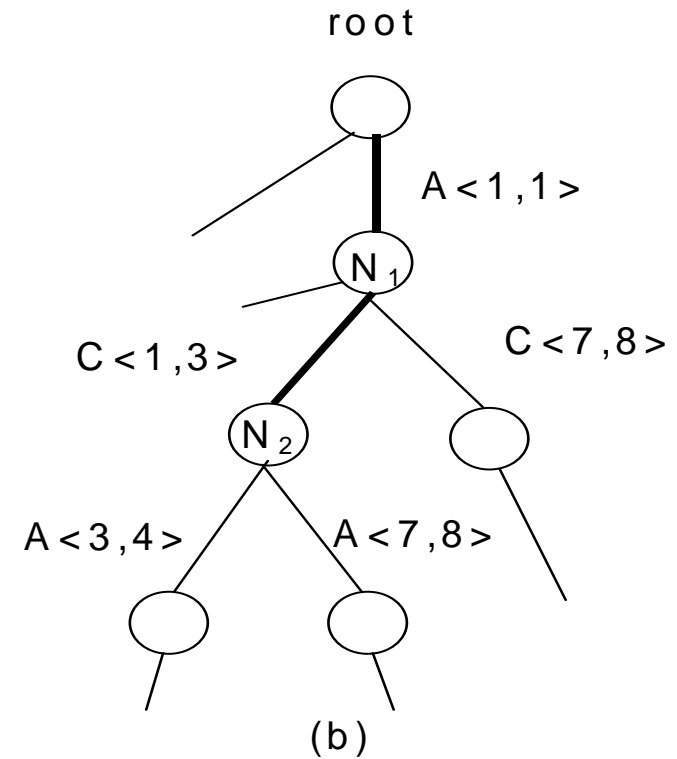
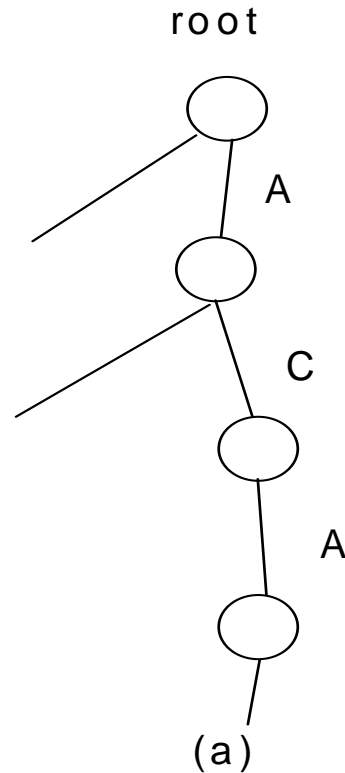
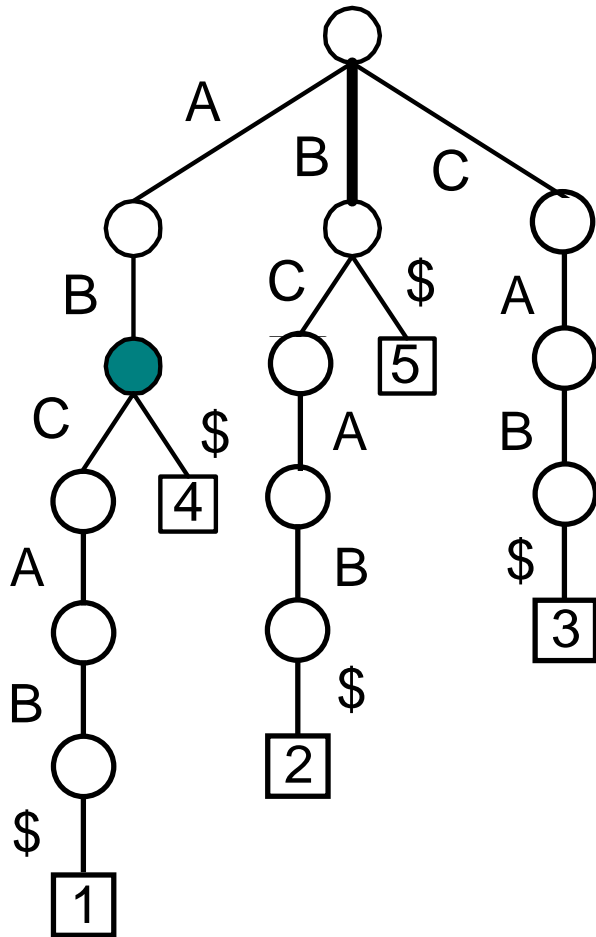
type D



note number



Tree-Based Index (Cont.)



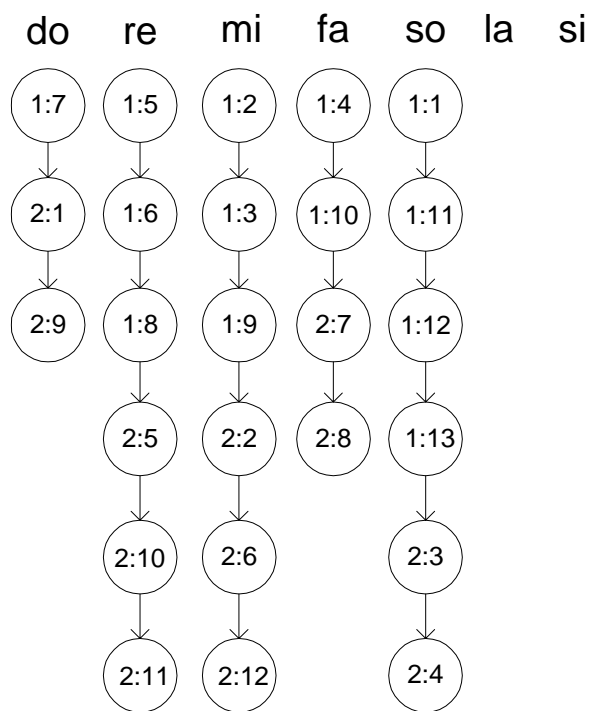
(a) An example suffix tree
 (b) A 1-D augmented suffix tree

The suffix tree of the string $S="ABCAB"$

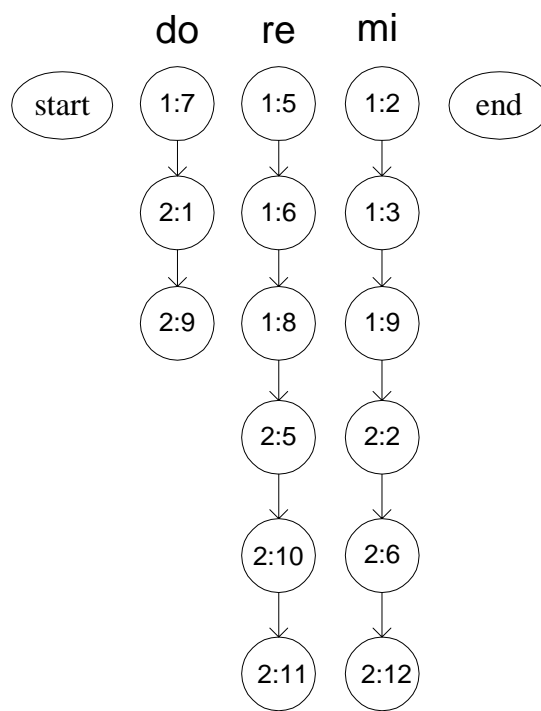
List-Based Index

- [Liu, Hsu and Chen, ICMCS'99]
- Music objects are coded as melody strings
 - “so-mi-mi-fa-re-re-do-re-mi-fa-so-so-so”
- Melody strings are organized as linked lists
- Both incipit/partial and exact/approximate matching can be handled
 - Exact link, insertion link, dropout link, transposition link

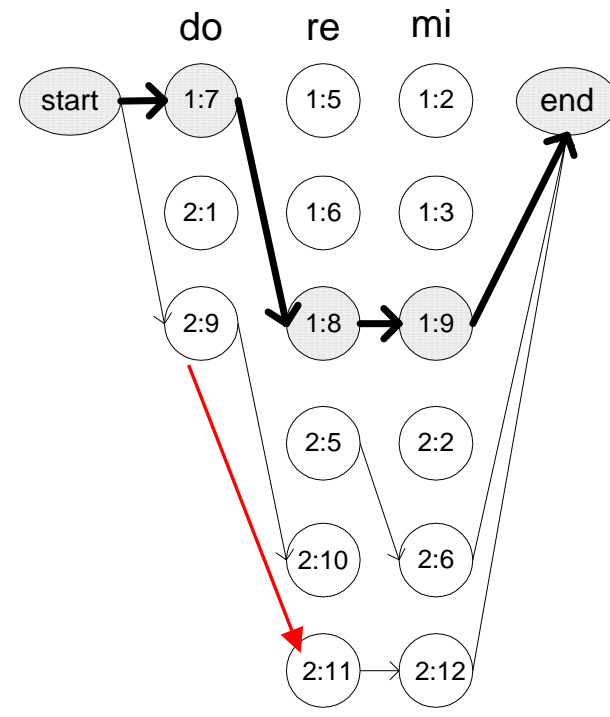
List-Based Index (Cont.)



(a)



(b)



(c)

N-Gram Index

- [Doraisamy and Ruger, ISMIR'02]
- A widely used technique in music databases
- Target strings are cut into index terms by a sliding window with length N
- Index can be implemented by various methods, e.g., inverted file
- Queries are also cut into index terms with length N
- Searching and joining are then performed

N-Gram Index (Cont.)

S=aabbcaab

2-Gram	Position
aa	1,6
ab	2,7
bb	3
bc	4
ca	5

Inverted file

Query=bbca

↓ Cut into 2-grams

bb, ca

Position: 3

Position: 5

Join

The substring is found from position 3 to position 6

System Evaluation

- Traditional measures of effectiveness are precision and recall

$$\textit{precision} = \frac{\textit{number of retrieved references that are relevant}}{\textit{number of references that are retrieved}}$$

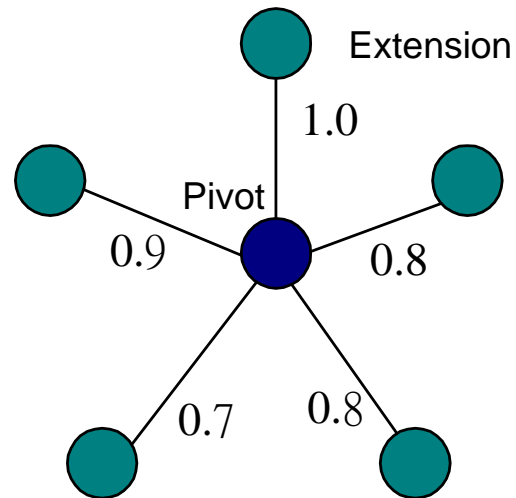
$$\textit{recall} = \frac{\textit{number of retrieved references that are relevant}}{\textit{number of relevant references}}$$

A Platform for Evaluating MIR Systems

- Evaluation of various music retrieval approaches
 - Efficiency
 - Effectiveness
- The Ultima project builds such a platform [Hsu, Chen and Chen, CIKM'01]
 - Same data set and query set for various approaches
 - Compare recall-precision curves

Repeating Pattern Discovery

- A repeating pattern in music data is defined as a sequence of notes which appears more than once in a music object
- The themes or motives are typical kinds of repeating patterns
- Exact repeating patterns [Hsu, Liu and Chen, TMM'01]
 - By the string-join operator
- Approximate repeating patterns [Liu, Wu and Chen, Dasfaa'05]



- Each vertex represents a music segment and the number on each link indicates the similarity degree between the two segments
- The center is called a *pivot*; if the similarity degree to the pivot is larger than the predefined threshold, the associated segment is then called an *extension* of the pivot
- A pivot is an ARP if the number of extensions reaches the support threshold

Phrase Extraction

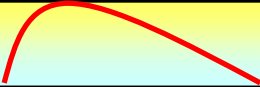
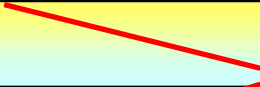
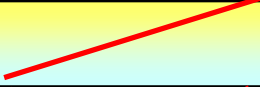
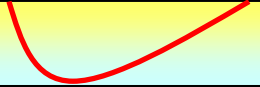
- Two features used for phrase extraction
 - Duration and rest
- Melodic Shapes [Huron, Computing in Musicology'95]
 - Statistics Information in Western Folksongs
 - The most common length of a phrase is 8 notes
 - Half of all phrases are between 7 and 9 notes in length
 - Three-quarters of all phrases are between 6 and 10 notes in length

Phrase Extraction (Cont.)

A: the pitch value of the **first** note in the target phrase

B: the pitch value of the **last** note in the target phrase

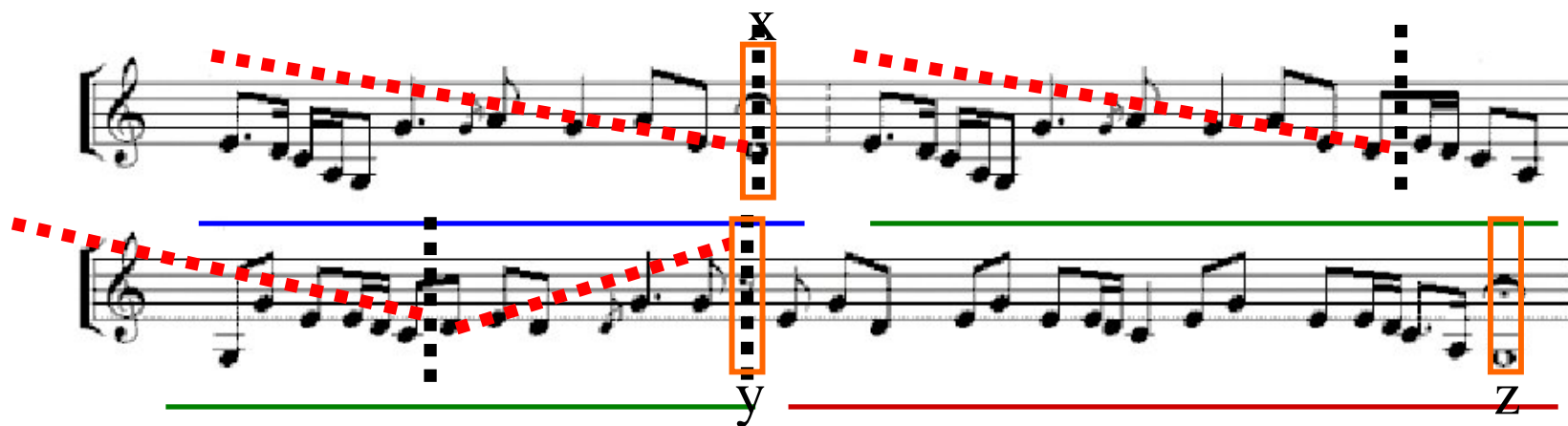
C: the **average** pitch value of the remaining notes in the target phrase

Contour Type	Number of Phrases	Percentage	Arch Shape	Definition
Convex	13926	38.6%		$A < C \wedge B < C$
Descending	10376	28.8%		$A > C > B$
Ascending	6983	19.4%		$A < C < B$
Concave	3496	9.7%		$A > C \wedge B > C$
Others	1294	3.5%		

Phrase Extraction (Cont.)

- Identify the positions of all the **terminative notes**
- Extract the **music pieces** according to the terminative notes
- Select the candidate music pieces for decomposition based on the length information
 - If the length ≤ 12 , the music piece is marked as a phrase
 - If the length > 12 , decompose the music piece into phrases
 - **convex > descending > ascending > concave**

Phrase Extraction (Cont.)



64 62 60 57 55 67 67 69 67 69 64 62

64 62 60 57 55 | 67 67 69 67 69 64 62 | 64 62 60 57 55 67 64 64 62 60 62 64 62 62 67 67

	Order	The Length of the Prefix Fragment	The Pitch of the First Note	The Pitches of the Remaining Notes	The Pitch of the Last Note
Convex?			4	62, 60, 57, 55	67
No			4	62, 60, 57, 55, 67	67
			4	62, 60, 57, 55, 67, 67	69
Descending?			4	62, 60, 57, 55, 67, 67, 69	67
			4	62, 60, 57, 55, 67, 67, 69, 67	69
Length = 12, A = 64, B = 62, C = 63.7			4	62, 60, 57, 55, 67, 67, 69, 67, 69	64
	7	12	64	62, 60, 57, 55, 67, 67, 69, 67, 69, 64	62

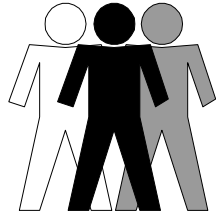


Music Classification

- Different kinds of music units can be extracted from music objects, such as repeating patterns and phrases
- Different kinds of music units may have different semantics in musicology
- These extracted music units can be used in music classification, retrieval, and analysis

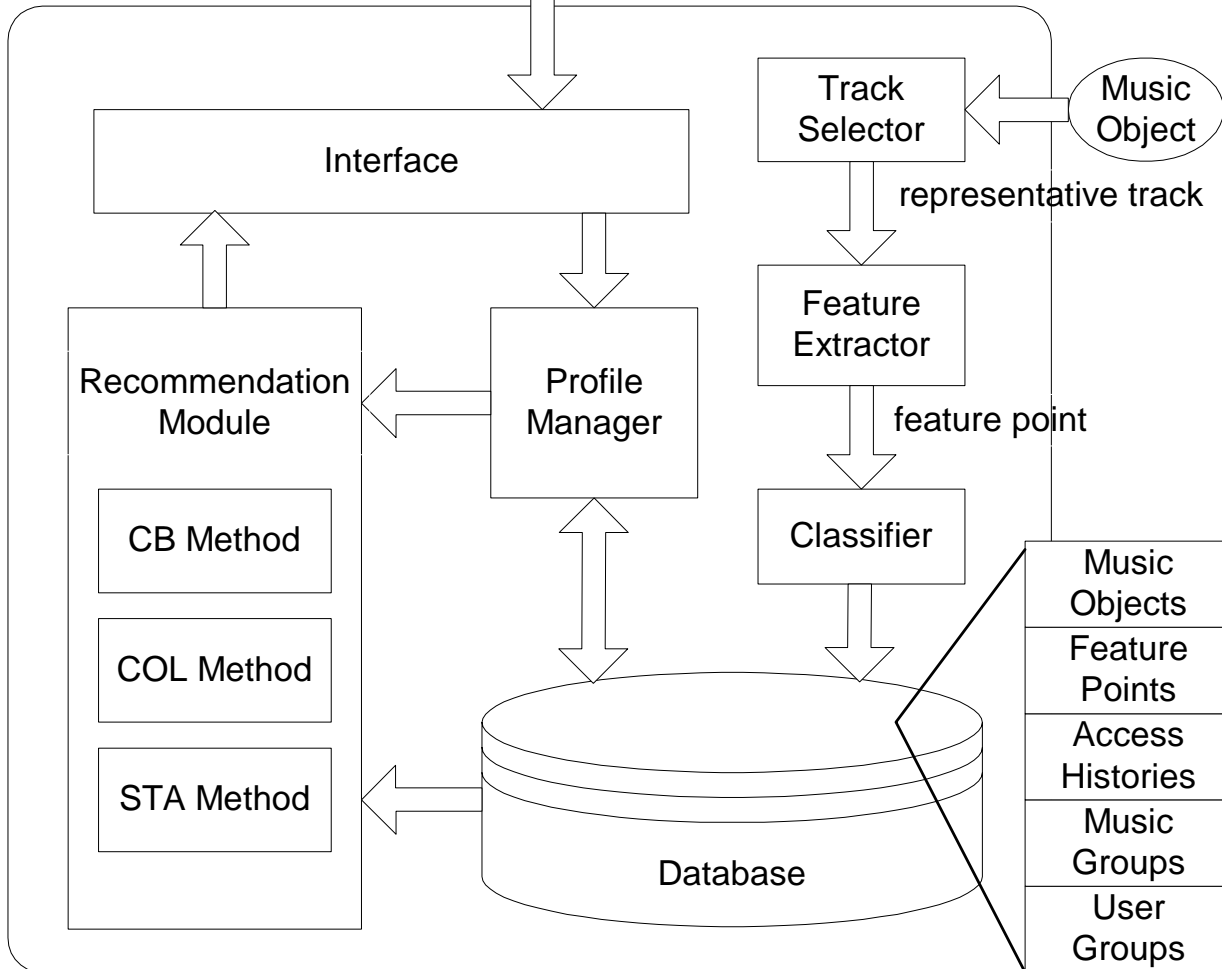
Music Recommendation Systems

- [Chen and Chen, CIKM'01]
- The results of music classification can be used for music-related services
- By analyzing the user access histories, we can discover which music classes the users may be interested in and which users belong to the same group
- By using different kinds of recommendation mechanisms, we can recommend the users suitable music objects



Users

Architecture



A polyphonic music object

- one melody track
- other accompaniment tracks

Recommendation Mechanisms

- **Content-based filtering approach**
 - Similarity between music objects and user profiles
 - Recommend the music objects that belong to the music groups the user is recently interested in
- **Collaborative filtering approach**
 - Similarity between user profiles
 - Provide collaborative recommendations to the users in the same user group
- **Statistical approach**
 - Recommend “hot” music objects



Future Research Directions

- Similarity measures, index structures and segmentation for polyphonic music
- Cross-media retrieval
- Music retrieval based on user's emotion/situation
 - Music mood analysis
 - music therapy
 - Creating a High Quality Learning, Relaxing and Sleeping Environment through Affective and Attentive Computing