

# Effectiveness of HMM-Based Retrieval on Large Databases

Jonah Shifrin

William Birmingham

University of Michigan, Ann Arbor

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# Performance on a large Database

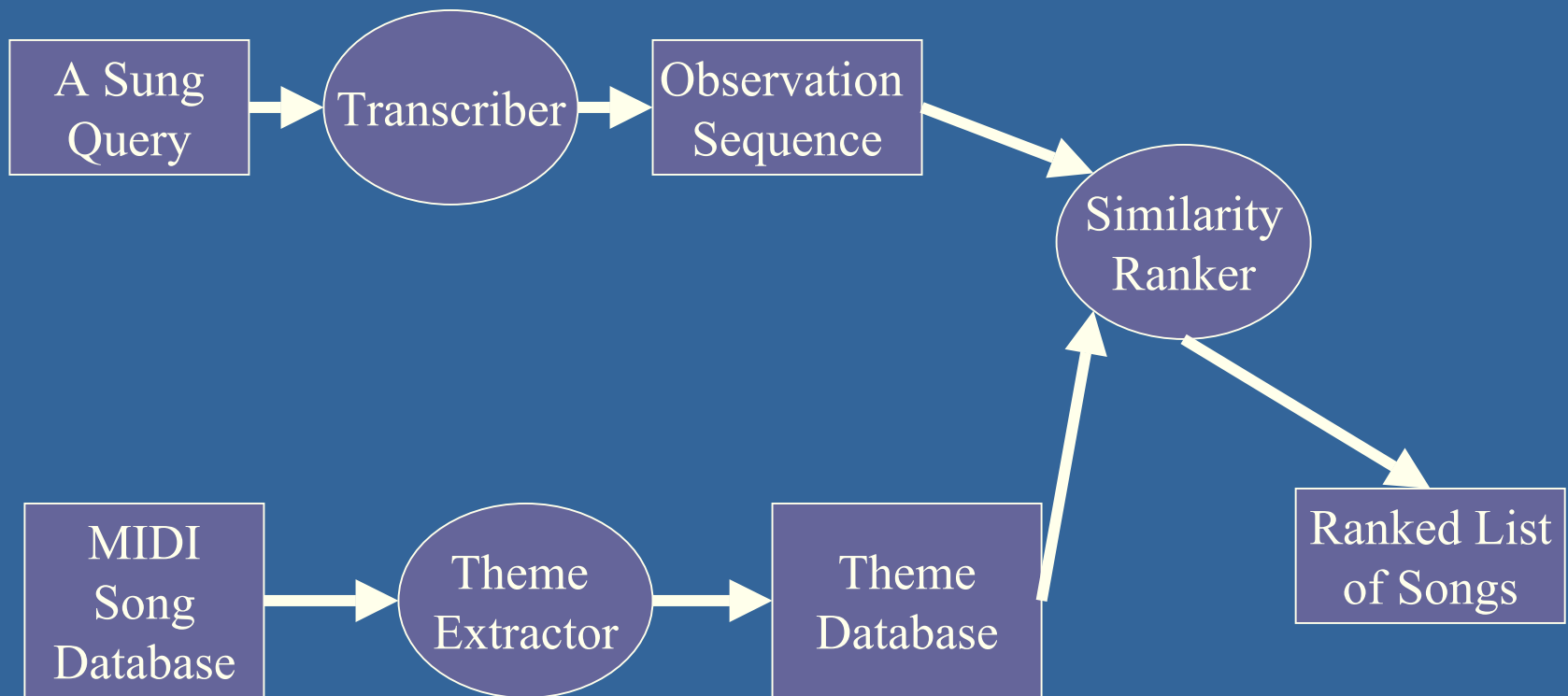
- “real-world” system contain a large database
  - 100-gigabyte hard drive can hold 22,000 mp3s
- Are individual themes distinguishable at all?
- Can queries modeled for our system be retrieved?
- Will “real-world” modeled queries be identified by our system?

# Presentation Outline

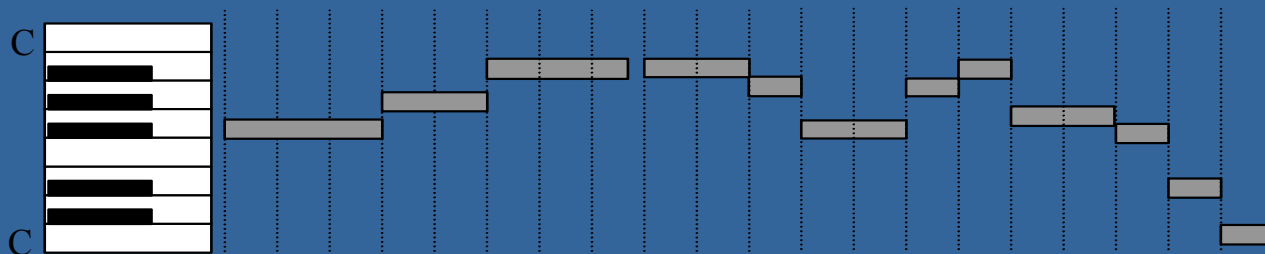
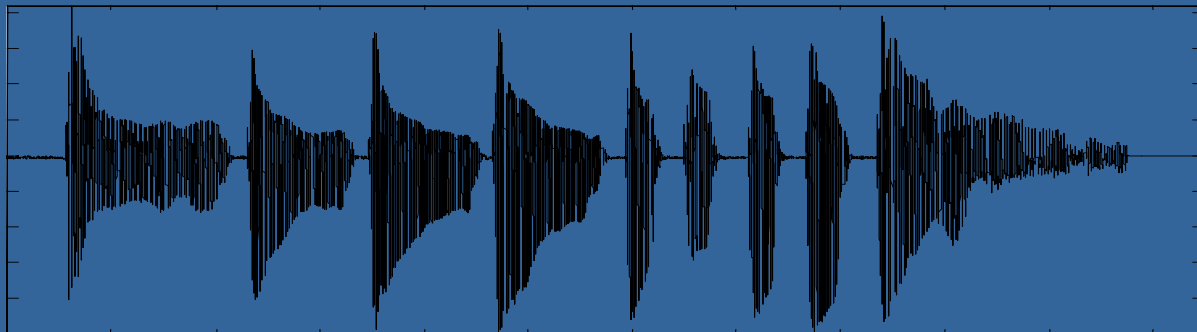
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- Overview of System
- Large Database Creation
- Query Creation
- Experimental Results
- Conclusions

# Introduction to Existing System



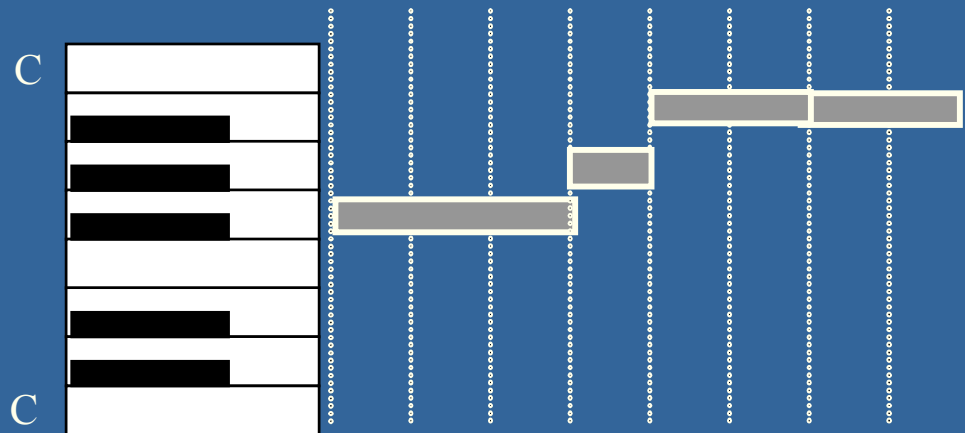
# Sung Queries



delta pitch	2	2	0	-2	-2	2	2	-4	-1	-3	-3
IOI (100 ms units)	3	2	3	2	1	2	1	1	2	1	1
IOI ratio	1.5	.66	1.5	2	.5	2	1	1	2	1	1

# Query Representation

- Sequence of duples
  - Change in Pitch (Delta Pitch)
  - Rhythmic Ratio (Inter Onset Interval Ratio)



Delta Pitch	2	2	0	
IOI	3	1	2	2
IOI Ratio	3	0.5	1	

# Targets – Themes

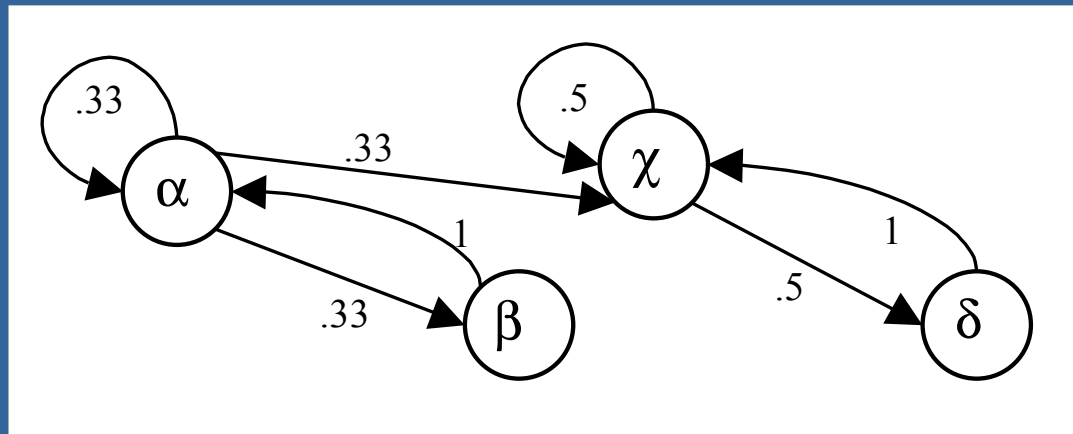
Theme



States

Delta pitch	2	2	1	2	-2	-1	-2	-2
IOI	1	1	1	1	1	1	1	1
IOI ratio	1	1	1	1	1	1	1	1
State	$\alpha$	$\alpha$	$\beta$	$\alpha$	$\chi$	$\delta$	$\chi$	$\chi$

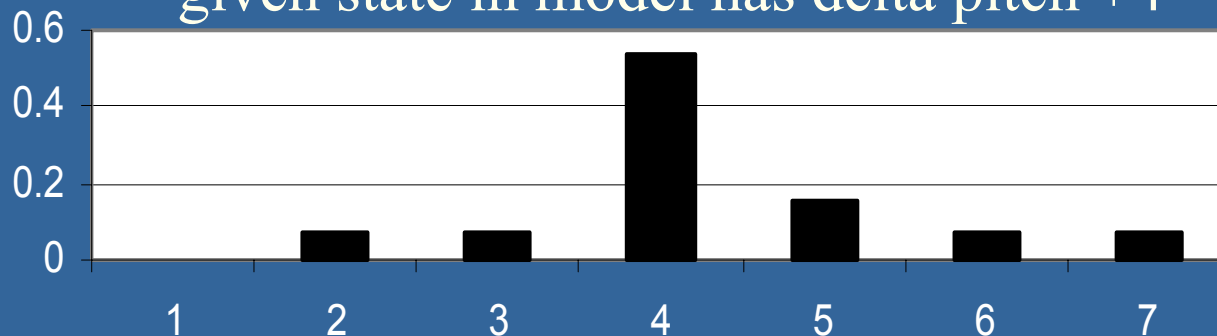
Markov Model



# Matcher

- Forward algorithm
  - determines probability target generated observation
- Pitch and duration assumed conditionally independent in observation tables

Probability of observing Delta Pitch  $i$ , given state in model has delta pitch +4



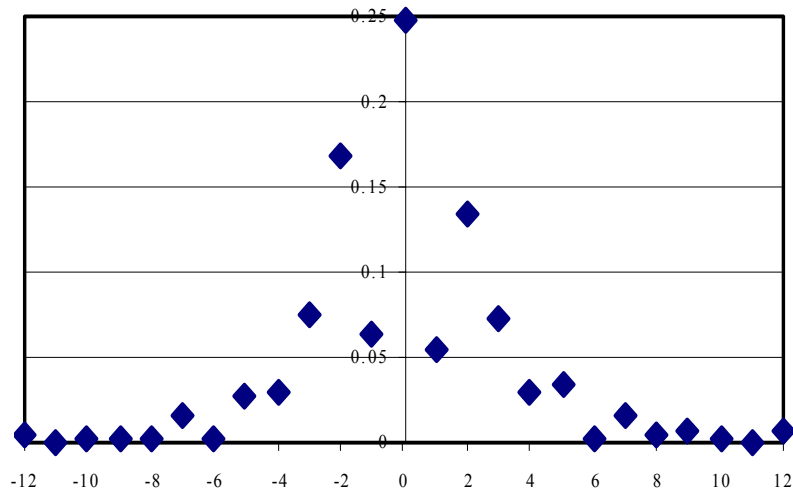


# Synthetic Target Database

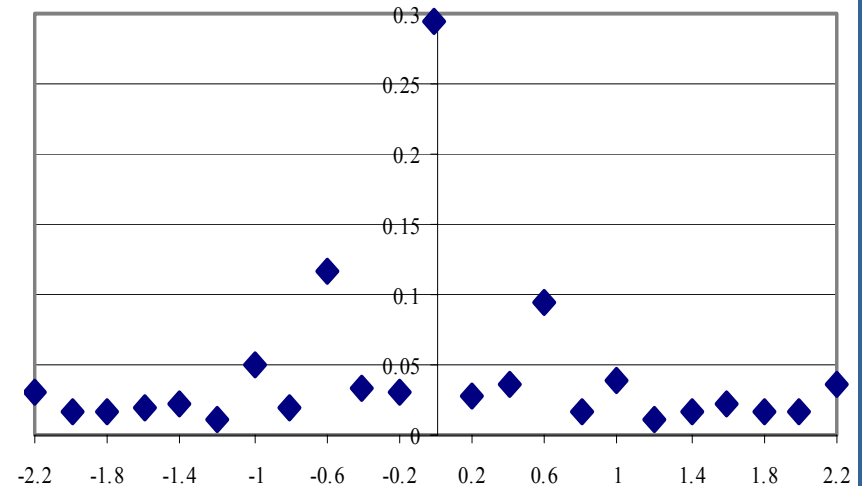
- Created synthetic database from Beatles database
  - Beatles Database has 284 themes from 260 songs
- Examined *length*, *deltaPitch* and *IOratio*
- Median theme length is 40 notes
  - standard deviation of 20

# Database Generation

Fraction of Database by deltaPitch



Fraction of Database by ln(IOIratio)



- Created 50,000 themes

# Synthetic Query Generation

- Three query sets containing 5100 queries
  - Query length ranges from 5 to 55 notes
  - 100 queries of each length
  - How many notes necessary to distinguish target?
- Queries are notes and durations
  - Equivalent to segmented, pitch-tracked sung queries

# Query Set 1: Perfect Queries

- Are individual themes distinguishable at all?
- Perfect queries created by exacting excerpt of length  $n$  from target in database
- Provide baseline

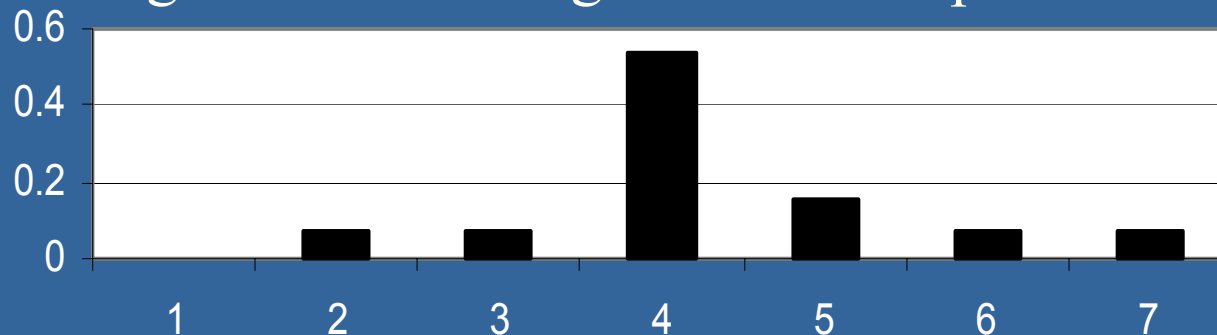
## Query Set 2: Imperfect Queries

- Can queries modeled for our system be retrieved?
- Assumption: Perfect model of singer and transcription error
- Generate queries based on singer-error statistics
- Represent best-case real world scenario

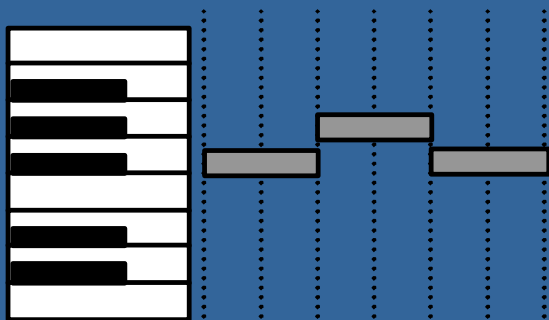
# Imperfect Query Generation

- Select a length- $n$  subsequence in target
- Transform selected portion into duples of
  - $\langle \text{deltaPitch}, \text{IOIratio} \rangle$
- Generate query from observation tables

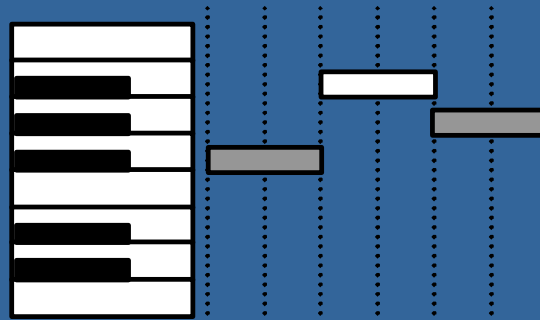
Probability query has a of Delta Pitch  $i$ , given state in target has a delta pitch +4



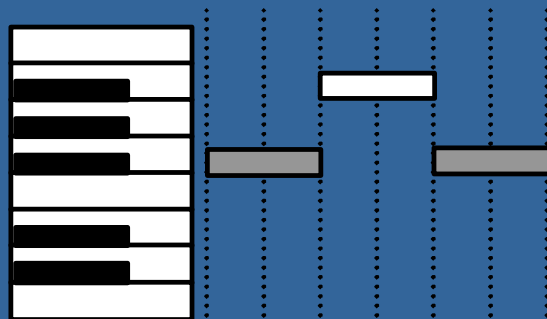
# Imperfect Queries: Assume Cumulative Error only



*deltaPitch:* +2 -2  
Target on Piano Roll



*deltaPitch:* +4 -2  
Query with Cumulative Error



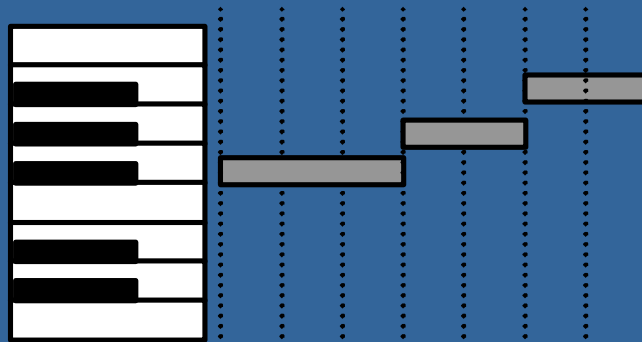
*deltaPitch:* +4 -4  
Query with Local Error

# Query Set 3: Imperfect Queries with Insertions and Deletions

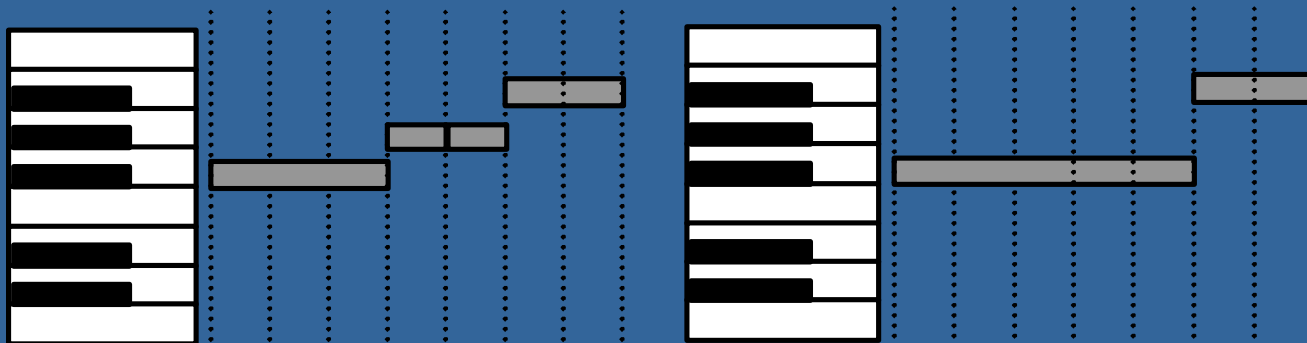
- Will “real-world” modeled queries be identified by our system?
- Insertions and Deletions present in real-world queries
- Sources:
  - People
  - pitch-trackers/segmenters



# Insertions and Deletions



Query on Piano Roll



Insertion on Second Note

Deletion on Second Note

# Insertion and Deletion Generation

- Used same method as Imperfect Queries
  - Insert or delete notes based on probabilities

$$P(\text{“no edit”}) = 0.81$$

$$P(\text{“insertion”}) = 0.06$$

$$P(\text{“deletion”}) = 0.13$$

# Matcher Modifications

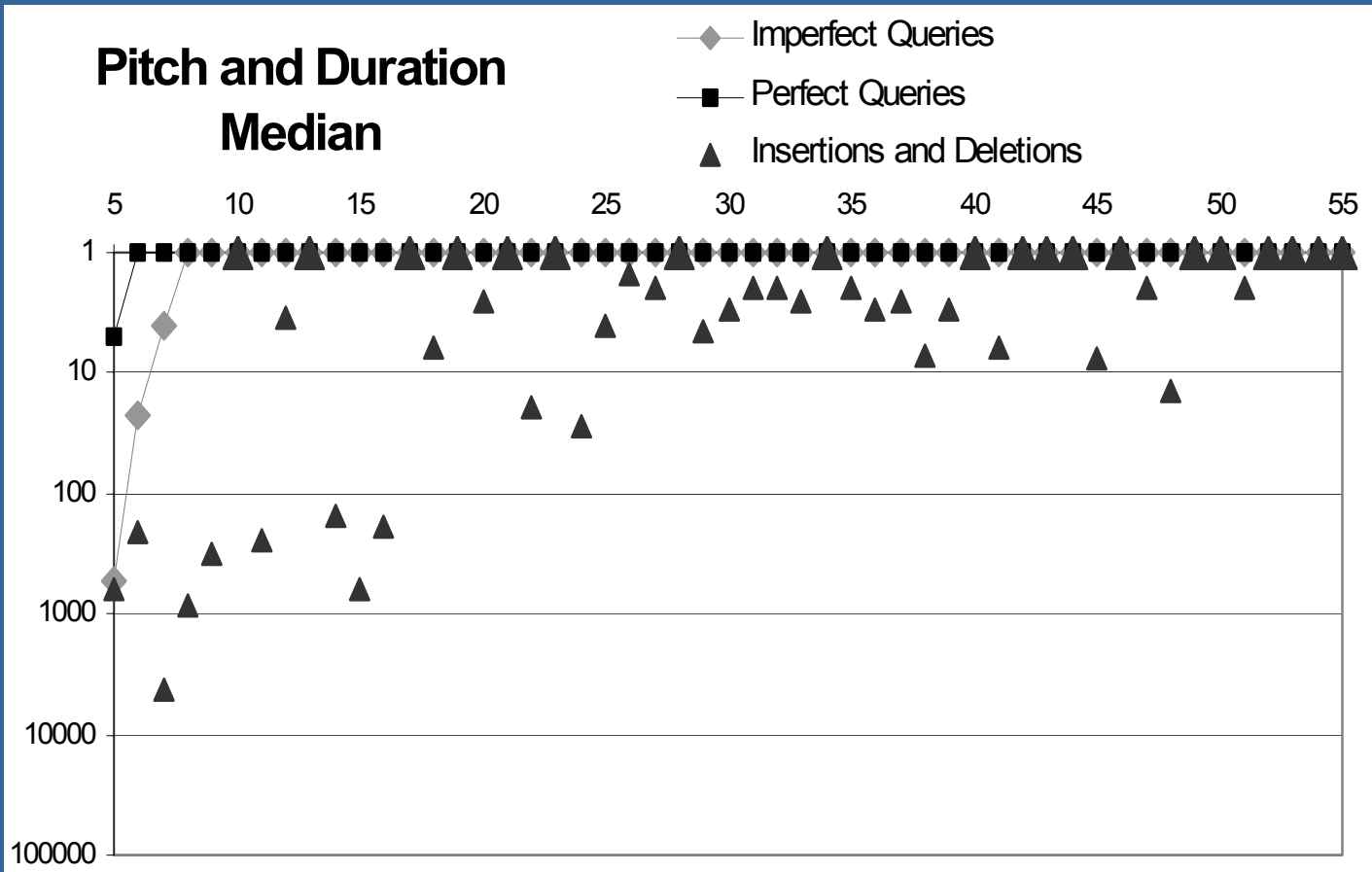
- Pitch and Duration Matcher
  - No modifications
- Pitch Only Matcher
  - Only consider deltaPitch in observation table
- Duration Only Matcher
  - Only consider IOIratio in observation table
- Show effectiveness of various representations

# Results:

## Pitch and Duration Matcher

Significant Benefit from longer queries

Huge drop-off with insertions and deletions

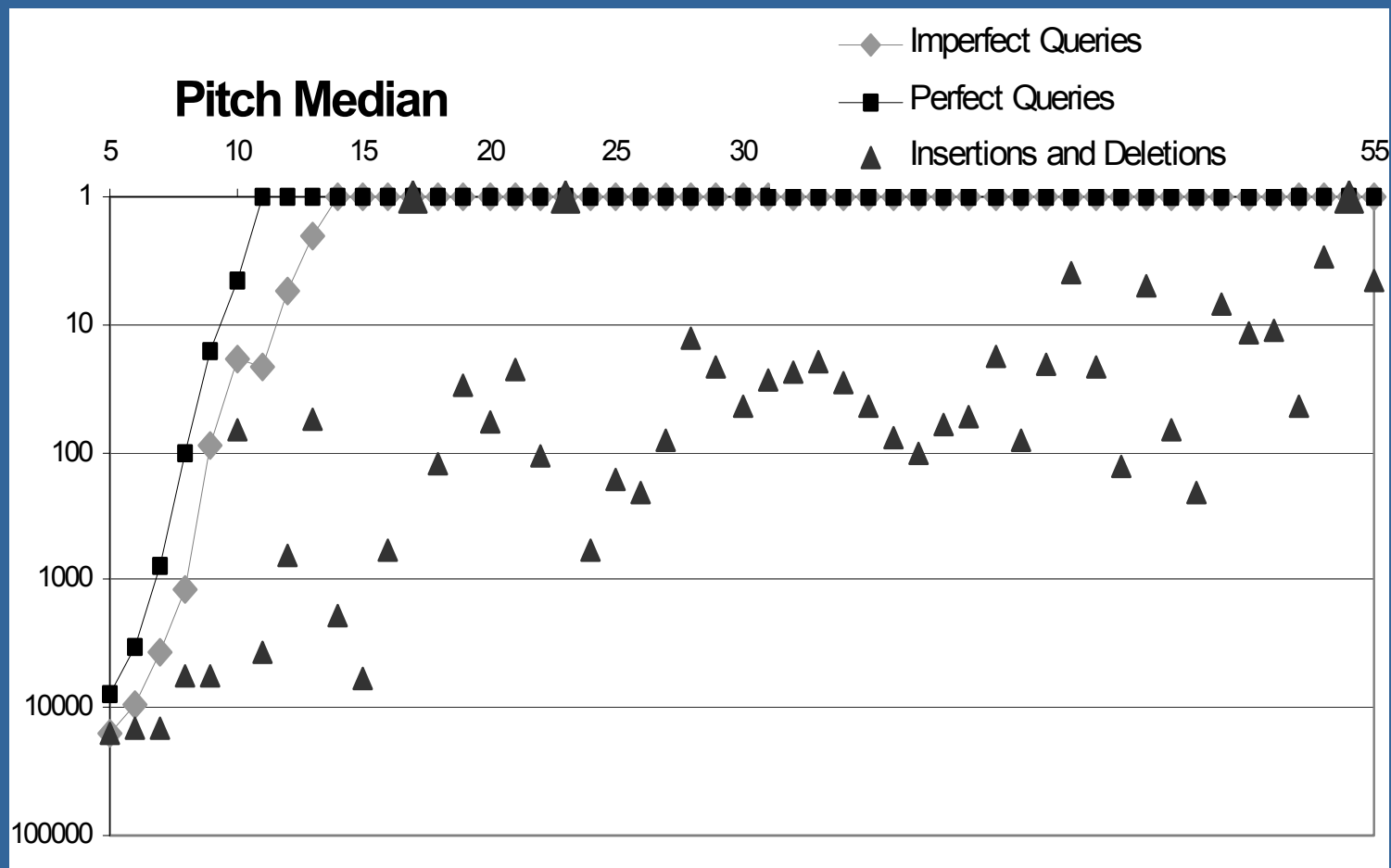


# Results:

## Pitch Only Matcher

Similar results to Pitch and Duration

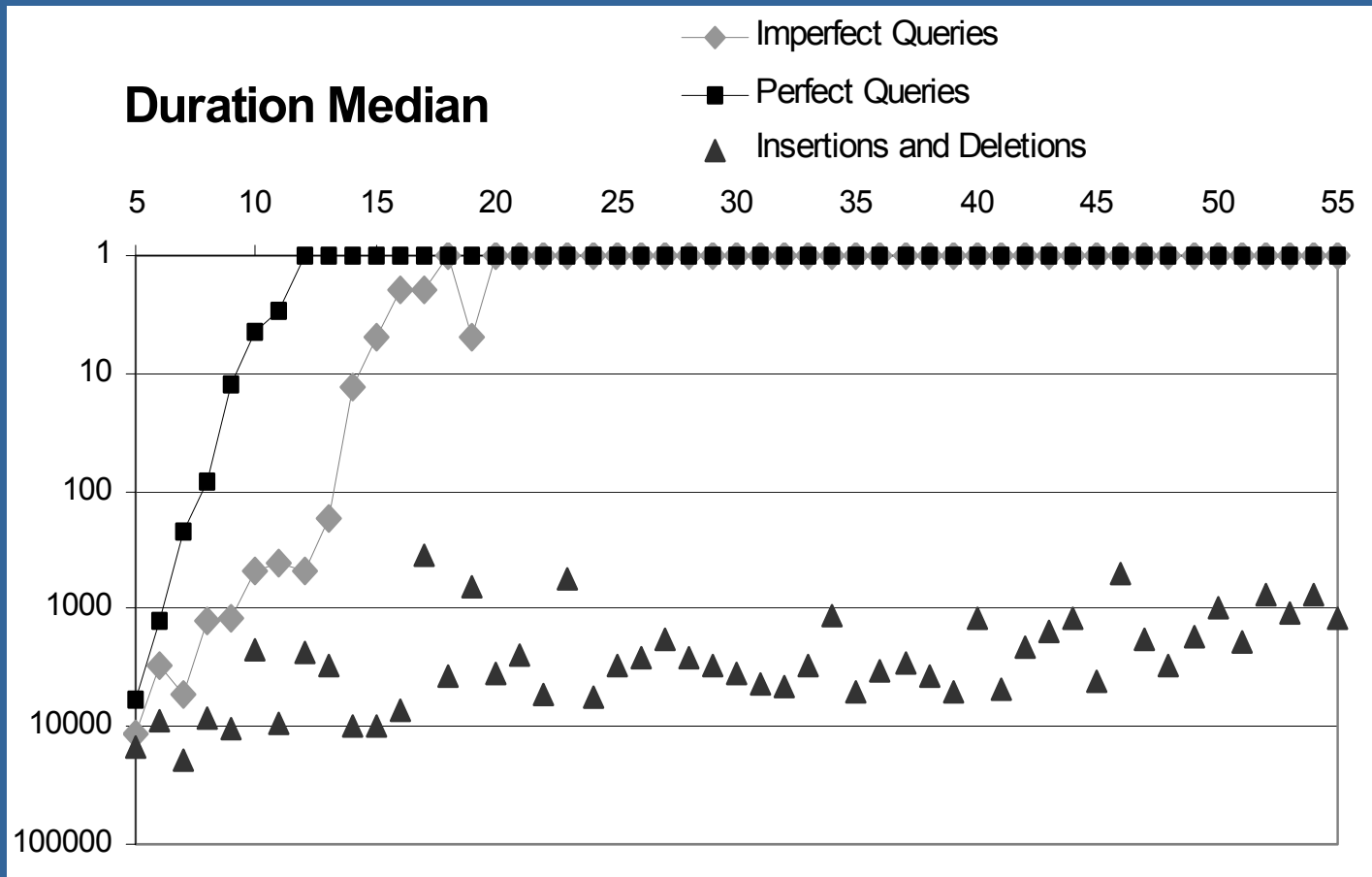
Requires more notes to distinguish queries



# Results: Duration Matcher

Significant Drop-off

Terrible performance by Insertions and Deletions



# Conclusions and Future Work

- Query length is significant
  - Saturation point
  - Query length used as confidence factor
- Topological Improvements
  - Insertion and Deletion states
- Suggests a change to the duration model

# Conclusions

- Are individual themes distinguishable at all?
  - Yes. Perfect Queries performed quite well
- Can queries modeled for our system be retrieved?
  - Yes. Results were very encouraging.
- Will “real-world” modeled queries be identified by our system?
  - Not reliably. Needs topological modifications.



# Questions

