

# Melodic Pattern Extraction in Large Collections of Music Recordings Using Time Series Mining Techniques

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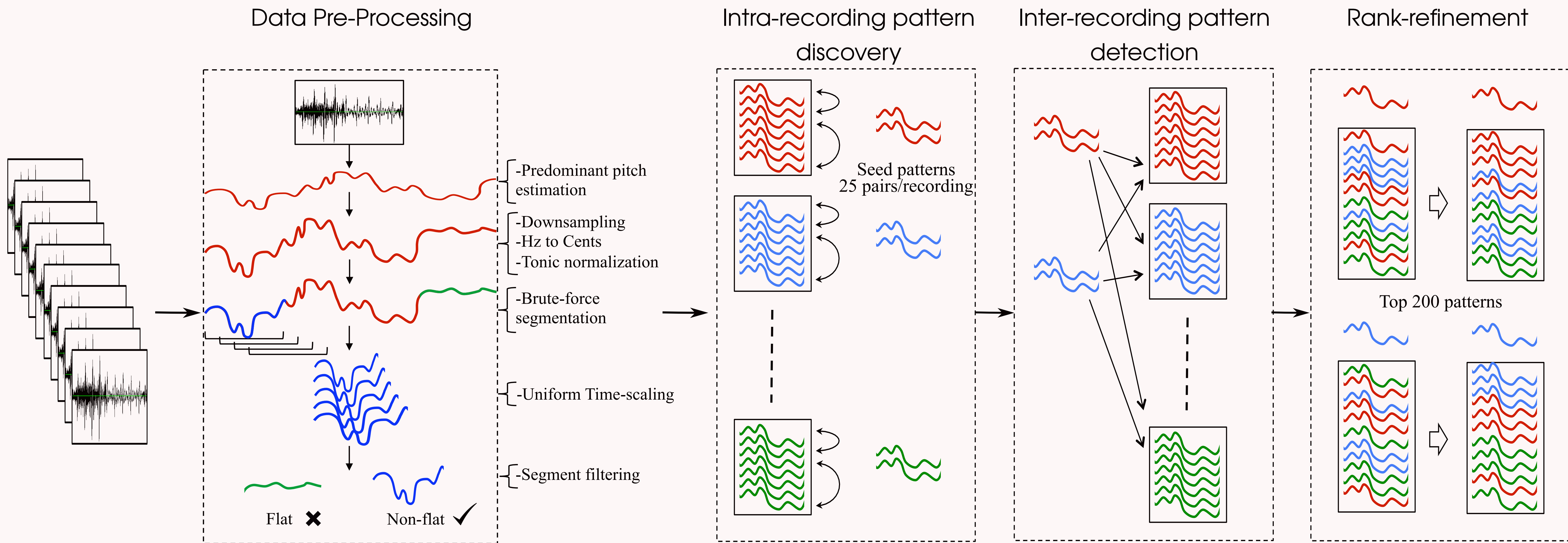
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## Goals

- Discover short-time melodic patterns in audio collections of Indian art music (IAM)
- Assess scalability of melodic similarity measures based on DTW
- Evaluate four variants of DTW cost function for rank refinement
- Assess the tightness of lower bounding techniques for DTW-based distance measure for the given data

## Methodology

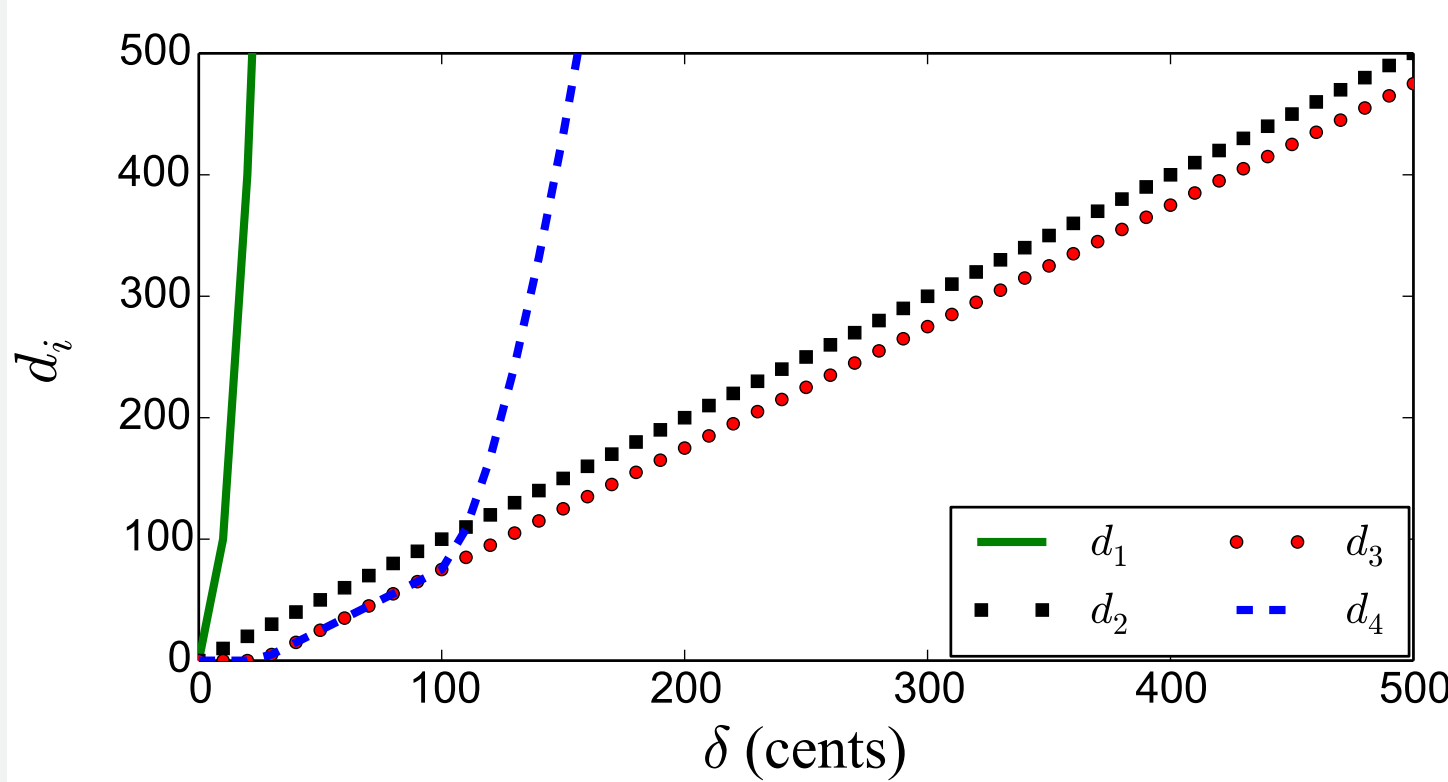


## Melodic similarity

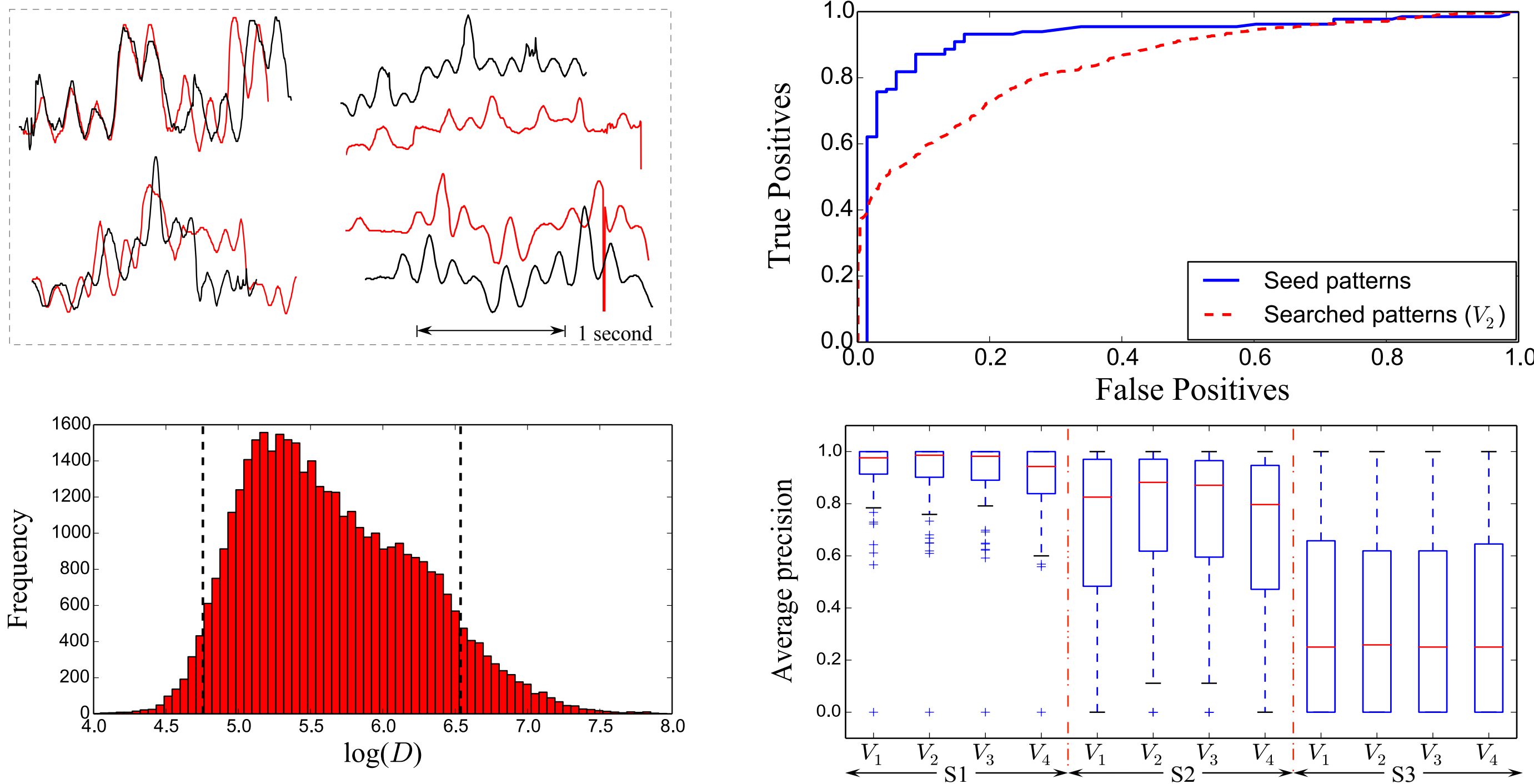
- Whole sequence matching DTW-based distance measure
- Square Euclidean distance as DTW cost function
- No local constraint and transition cost
- Step condition:  $\{(1,0), (1,1), (0,1)\}$
- Sakoe-Chiba global band constraint (10% pattern length)
- Use first-last (FL) lower bound
- Use LB\_Keogh lower bound

### Rank-refinement

- Local constraint
- Step condition:  $\{(2,1), (1,1), (1,2)\}$
- 4 different DTW cost functions
- No lower bounding used



## Results and Discussion



Intra-recording discovery:  $1.4 \cdot 10^{12}$  DTW computations (75% avoided)

Inter-recording search:  $12.4 \cdot 10^{12}$  DTW computations (99% avoided)

## Evaluation

### Dataset:

- CompMusic Carnatic music collection
- 1764 polyphonic audio recordings
- 365 hours of music material covering different forms, rāgas and artists
- Over 300 million pattern candidates

### Evaluation:

- 79,000 seed patterns & 15 million search patterns
- Randomly sample 200 seed patterns and top 10 search patterns
- Total of 8000 patterns pairs evaluated by a professional musician
- Mean average precision (MAP) to quantify musician's assessment
- ROC curves for the analysis of distance distribution

## Browse melodic patterns

The screenshot shows a table of search results for a chosen seed from Nadamadi. The table includes columns for Similarity, Seed\_id, Start(t), End(t), Pair\_id, Start(t), End(t), Musicianship ID (searched file), and Distance. The results are sorted by similarity, showing various patterns and their corresponding distances.



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