



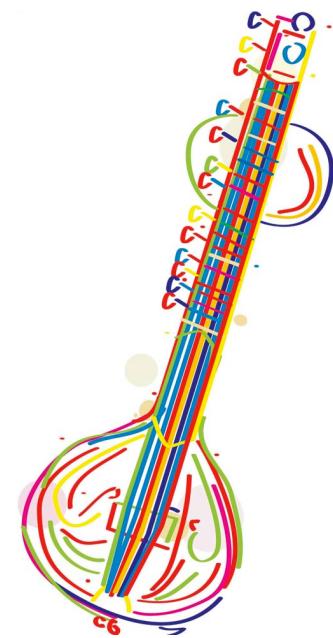
TOPICS IN COMPUTER MUSIC

Automatic Raga Recognition in Hindustani Classical Music

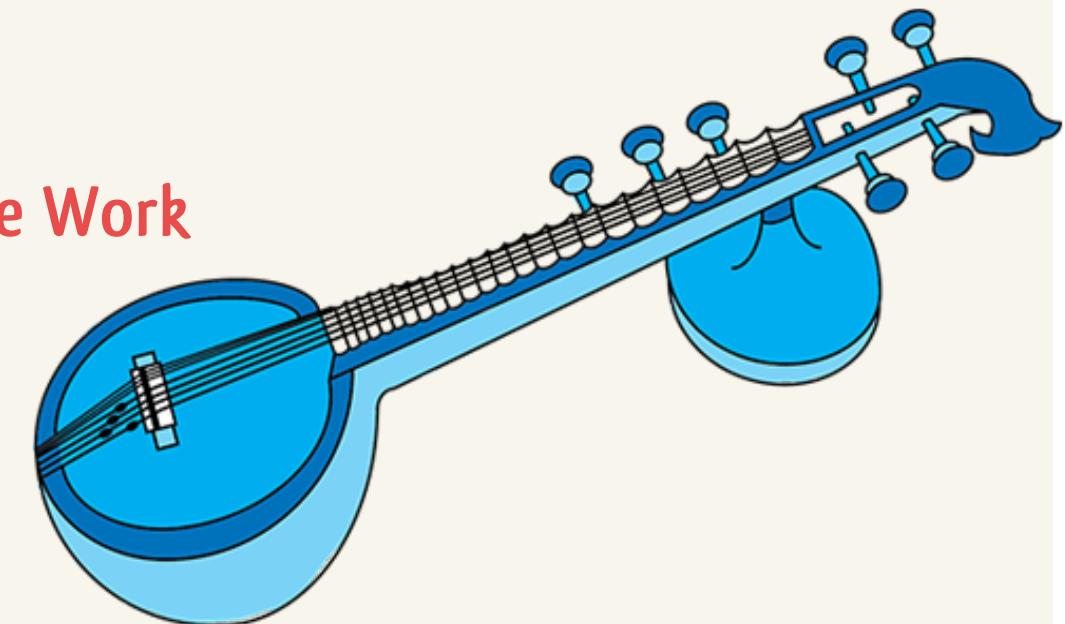
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1. Introduction
2. Approaches for Raga Recognition
3. Experiments & Results
4. Conclusion and Scope for Future Work
5. References



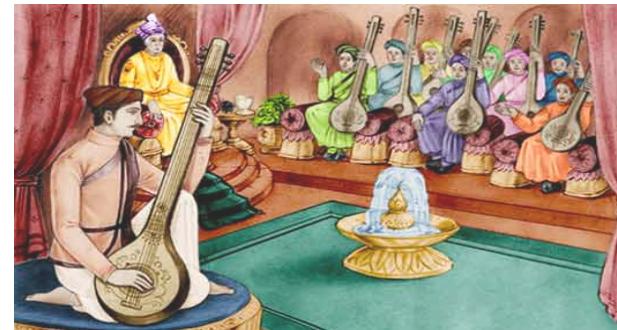


INTRODUCTION

1. What is Hindustani Music?
2. Common Instruments
3. What is a Raga?
4. Why is Raga Recognition challenging?

WHAT IS HINDUSTANI MUSIC?

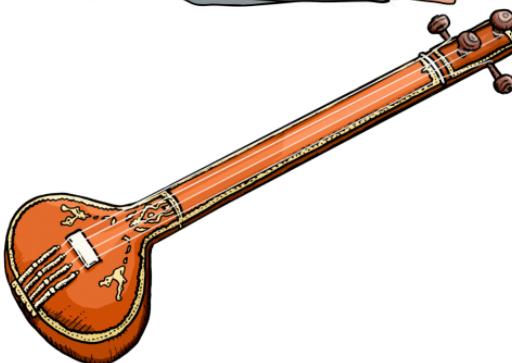
- Ancient form of music originating in Northern India
- Existence has been known and documented since 1500 B.C.
- Broadly classified into Vocal and Instrumental styles
- Traditionally passed down the generations by a 'guru'
- Characterized by imagination and improvisation



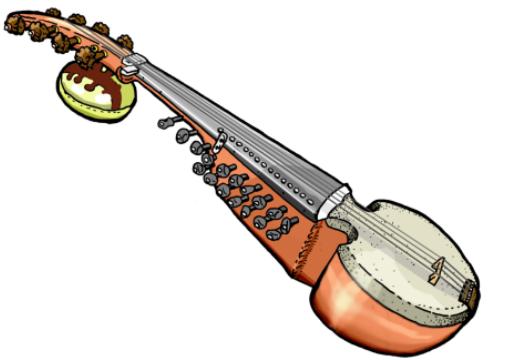
COMMON INSTRUMENTS



Sitar (सितार)
Stringed Instrument
used mostly in solo
performances



Tanpura (तानपुरा)
Stringed Instrument
used as a drone /
accompaniment



Sarod (सरोद)
Stringed Instrument
(fretless) used in solo
performances

Swarmandal (स्वर्मंडल)
Stringed Instrument
used as a drone /
accompaniment

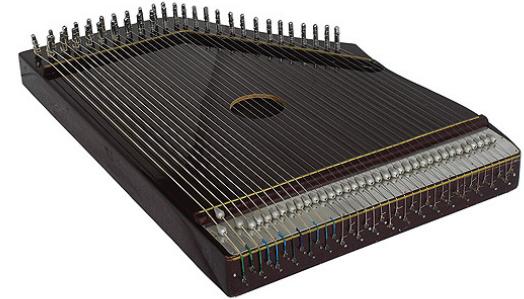
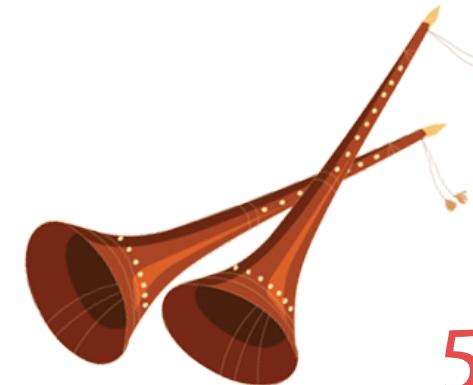


Tabla (तबला)
Percussion instrument
Keeps rhythm and
cyclic time signatures



Shehnai (शहनाई)
Wind instrument
similar to an oboe for
solo renditions



WHAT IS A RAGA?

- A collection of melodic gestures, along with techniques for developing them^[2]
- Compositions are bound to a rhythmic cycle*
- Longer phrases can be built by joining these melodic atoms together^[2]
- Ragas are always associated with a time of the day to be sung in (sometimes the season) based on the mood they invoke

(७६)

श्यामकल्याण—त्रिताल (द्रुत).
स्थायी.

गम पध	मंप गम	गम म	रे सा	रे - सा सा	रे - नि सा
श्यां SS	SS SS	SS S	म क	ल्या ं S श	गा ं व त
.	.	3	x	2	
मं मं प प	मं प ध प	मं प ध प	ध - - म	प - म ग	
नि स दि न	च तु र गु	नी तु र गु	नी ं S S S	S S S S I	
.	3	x	2		

अन्तरा.

प - सां सां	सां - सां -	प नि सां रे	सां नि ध प
मे ं ल क	ल्या ं शी ं	अ' ं श ख	र ज क र
.	3	x	2
म - प प	मं प ध प	ध - - म	प - म ग
मं ध य म	जु गु ल गु	नी ं S S S	S S S S I
.	3	x	2

श्यामकल्याण—भपताल (मध्यलय)
स्थायी.

ग	-	मग	म रे	सा	रे	रे नि सा सा
प		नों	०	हे	०	श्या ं म
स	S	०	अ	०	०	३
x		२				
प						
म	प	म	प - ध	प	ध	मं प म ग
इ	म	म	०	०	०	
x	त	ती	नी ं S मो	री	वि	नं ं S ती॒०
	२					

* Alap is a notable exception

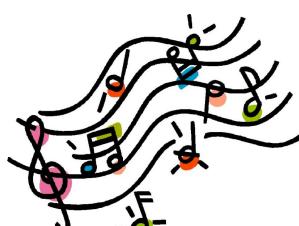
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SOME RAGAS - #1



RAGA CHARUKESHI

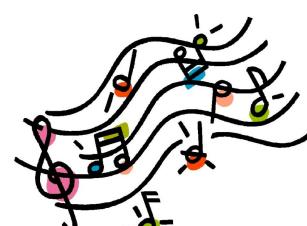
- Time : Late Morning – Noon (9:00 – 12:00)
- Ascent : सा रे ग म प ध नि सां
In C : (C D E F G G[#] A[#] C)
- Descent : सां नि ध प म ग रे सा, ध नि सा
In C : (C A[#] G[#] G F D[#] D C, G[#] A[#] C)
- Stressed Notes : सा (C) and म (F)
- Emotions : Pathos , Devotion , Beauty



Performed on: Sitar
Performed by: Ustad Shahid Pervez

RAGA BHIMPALASI

- Time : Afternoon
- Ascent : नि सा गु म प नि सां
In C : (A[#] C D[#] F G A[#] C)
- Descent : सां नि ध प म गु रे सा
In C : (C A[#] A G F D[#] D C)
- Stressed Notes : म (F) and सा (C)
- Emotions : Happiness, Pleasure



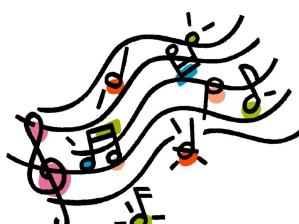
Performed on: Vocals
Performed by: Kaushiki Chakrabarty

SOME RAGAS - #2



RAGA PILU

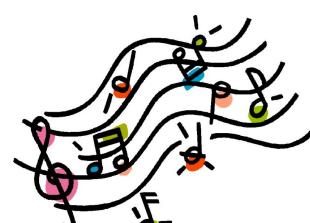
- Time : Night or Late Afternoon
- Ascent : नि सा गु म प नि सां
In C : (B C D[#] F G B C)
- Descent : सां नि ध प, ध म गु, रे सा नि गु सा
In C : (C A[#] A G, G[#] F D[#], D C B D[#] C)
- Stressed Notes: गु (D[#]), नि (B)
- Emotions : Love, Devotion, Pathos



Performed on: Sitar
Performed by: Sahana Banerjee

RAGA JOG

- Time : Late Night
- Ascent : सा ग म प नि सां
In C : (C E F G A[#] C)
- Descent : सां नि प म ग म, सा गु सा
In C : (C A[#] G F E F, C D[#] C)
- Stressed Notes : म (F)
- Emotions : Calmness, Love



Performed on: Flute
Performed by: Pandit Ronu Majumdar

- 
1. Richness, Diversity and Complexity of melodic types, making feature extraction difficult [2]
 2. Prevalence of Continuous Pitch Motions and ornamentations such as vibrato (अंदोलन) and portamento (मीड)
 3. Arbitrarily tuned tonic pitch, depending on the instrument and/or the voice of the vocalist
 4. Subtle and intricate differences between a few ragas introduces ambiguity in recognition



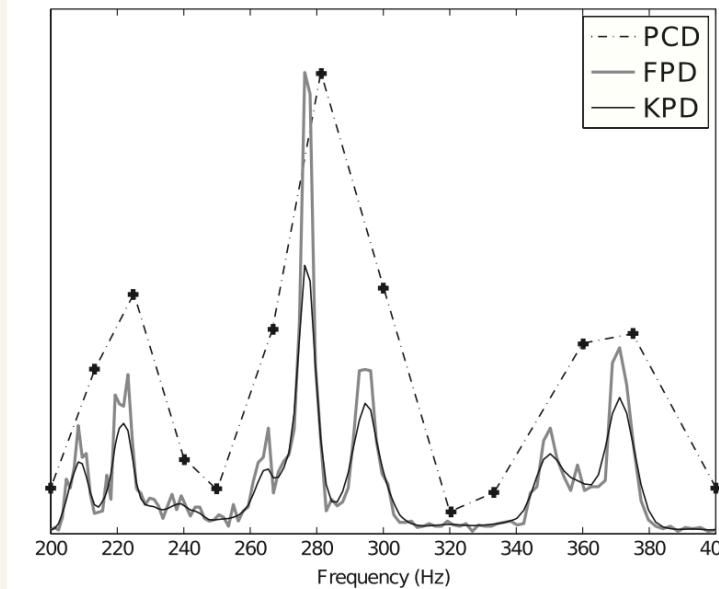
APPROACHES FOR RAGA RECOGNITION

1. Based on Pitch Distributions (Chordia et al.)
2. Using Vector-Space Models (Gulati et al.)

APPROACH 1 (Chordia et al.) AT A GLANCE^[2]

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- Based on Pitch Distributions (PD)
- Performs simultaneous recognition of raga and tonic
- Uses a Kernel-density pitch distribution apart from the standard 12-dimensional PCD
- Uses a Nearest-Neighbour classifier with Bhattacharya distance, attaining a 4.2% tonic rate and a 10.3% raga error rate



FLOWCHART

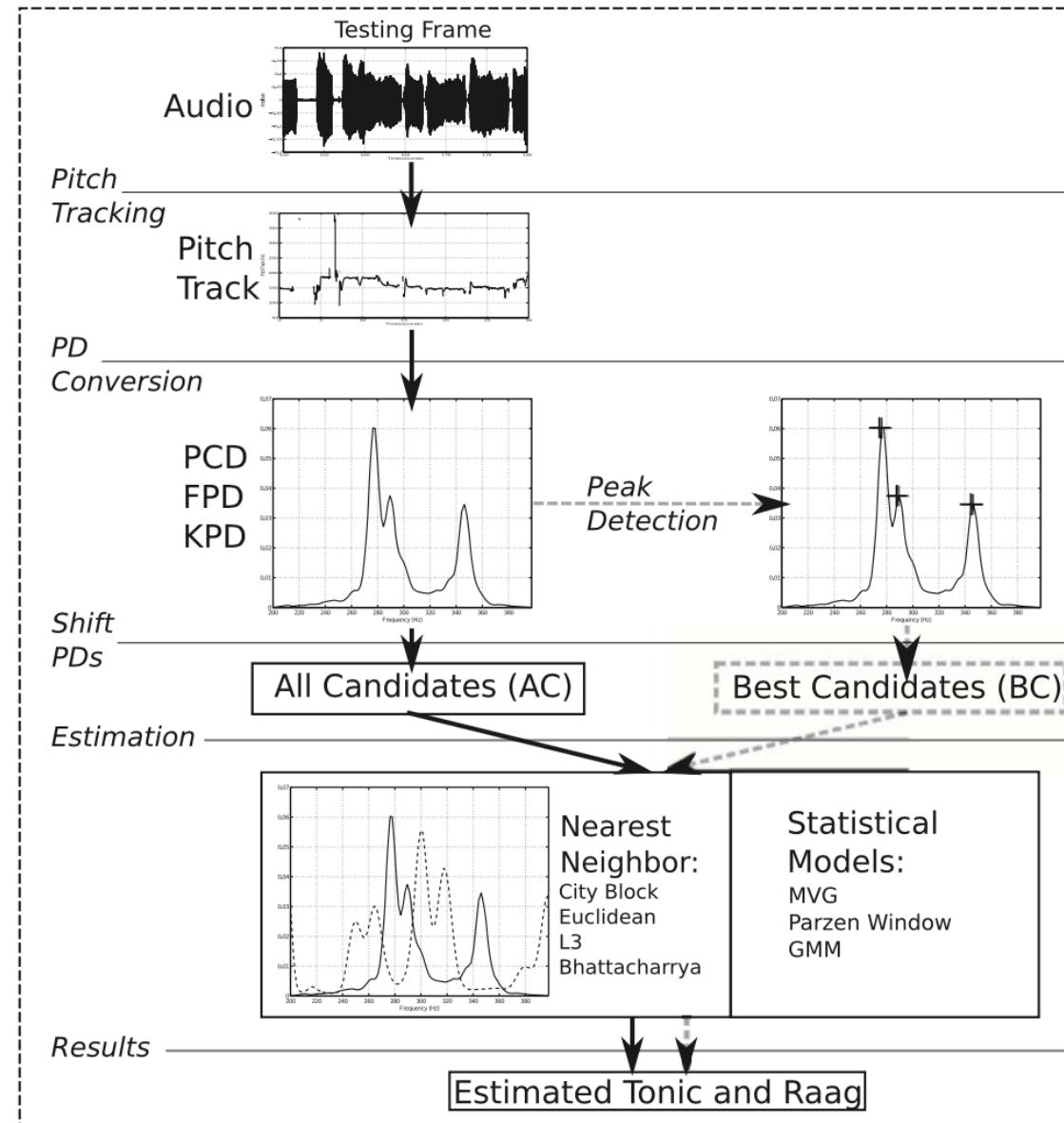


STEP - 1

STEP - 2

STEP - 3

STEP - 4



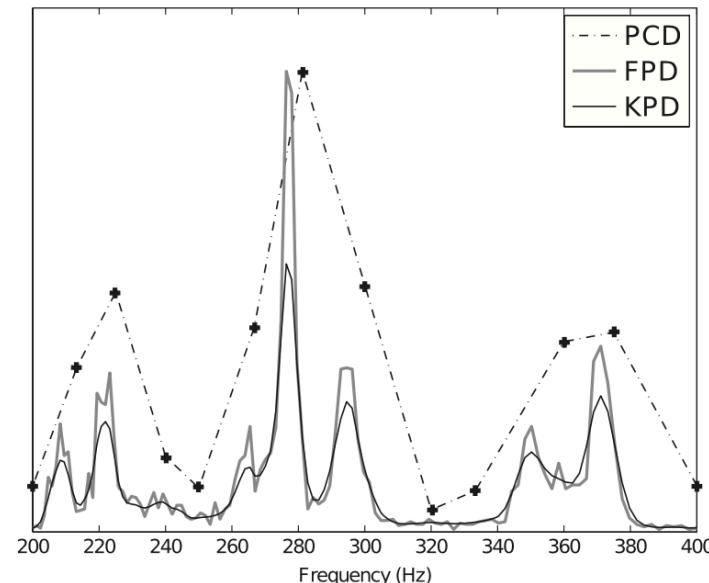
STEP 1: Pitch Tracking

- 
1. Sawtooth-waveform-inspired pitch estimator (SWIPE')^[5] is used.
Pitch is estimated as the fundamental frequency of sawtooth that best matches input signal
 2. Sound clip divided into 30 second chunks, after converting it to mono (to remove accompanying instruments)
 3. Pitch is estimated every 10 ms, and kept within range 73.4 - 587.2 Hz using a resolution of 48 steps per octave
 4. SWIPE' also returns an estimate of the pitch strength, which is a number between 0 and 1.

STEP 2: Extracting Tonal Features



- Three types of pitch distributions (PD) are used: a 12-D Pitch Class Distribution (PCD), Fine-Grained PD (FPD) and Kernel-Density PD (KPD)
- Each pitch estimate from the previous step is assigned in the following ways:
 - PCD assigns them to 12 chromatic pitch classes
 - FPD uses 120 or 240 bins (width of 10 or 5 cents)
 - KPD approximates a continuous PD function
- KPD centers a Gaussian window on the pitch value, and sum of all curves gives overall density



$$\widehat{f}_h(x) = \frac{1}{nh} \sum_{i=1}^n \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-x_i)^2}{2h^2}}$$

where

K is the kernel with a kernel width of h ,
 x_i is the value of the i th pitch value, and
 n is the total number of pitch values.

STEP 3: Tonic Estimation

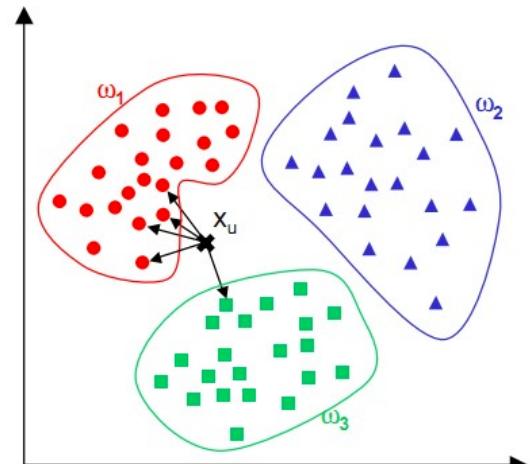
1. Based on calculating PD for different tonic pitches and finding one that gives the best match with the database
2. In All-Candidates approach, Brute Force method is used. PD is calculated for all 120/240 candidates in a circular fashion
3. The Best Candidates approach uses the fact that stable notes have peaks in HPDs. Only the 7 highest peaks are considered
4. PD is compared with all samples in the training database, and the nearest neighbour is found. The one with least overall distance is taken as the tonic



STEP 4: Raga Recognition



- Raga Recognition takes two alternative approaches: Nearest Neighbour Classification (k -NN) and Statistical Classifiers
- k -NN uses several alternative distance measures : City-Block, Euclidean, L3 norm and Bhattacharya distance
- Bhattacharya Distance is very popular for comparing probability densities, and is given by the formula
- For the statistical approach, Bayes' Rule^[6] is used
- For estimating the prior probabilities, multivariate Gaussian models are used with maximum-likelihood approach



$$D_B(p, q) = -\ln \left(\sum_{i=1}^n \sqrt{p_i q_i} \right)$$

where $p = (p_1, p_2, \dots, p_n)$ and $q = (q_1, q_2, \dots, q_n)$.

$$P(raag_i|x) = \frac{P(x|raag_i)P(raag_i)}{\sum_j P(x|raag_j)P(raag_j)}$$

EXPERIMENTS AND RESULTS



- For our experiments, our input are audio files of classical pieces, ranging from 3-60 minutes. The database used, is called the GTraagDB*
- The expected output from the algorithm is the correct tonic and the raga
- For both tonic and raga recognition, a 10-fold cross-validation is used
- The configurable parameters for the experiments were the precision (for tonic estimation), granularity, distance algorithm, PD used, all/best candidates
- Minimum error rate of 8.5% was attained using KPD with 5-cent granularity, 15% precision Nearest-neighbour with Bhattacharya Distance & all-candidates

EXPERIMENTS AND RESULTS

	PCD			FPD			KPD		
	Ground	All	Best	Ground	All	Best	Ground	All	Best
City Block	29.17	46.00	39.17	21.00	26.50	44.33	22.67	25.67	36.00
Euclidean	29.33	49.17	41.67	30.00	36.50	62.33	32.33	37.00	50.00
L3	32.50	48.83	43.33	39.50	45.67	71.67	40.17	46.33	58.83
Bhattacharyya	12.50	30.00	21.83	8.67	14.50	33.50	8.50	12.50	19.17
MVG	35.17	57.67	50.83	26.00	37.00	51.83	35.17	44.67	44.17
Parzen	33.50	53.17	46.67	27.50	34.17	53.00	33.67	42.33	43.00
GMM	31.67	43.83	39.50						

"Ground" is the error rate when the tonic is known in advance; "All" and "Best" are the all-candidates and best-candidate methods, respectively.

	Ground Truth		All Candidates		Best Candidates	
	10 Cents	5 Cents	10 Cents	5 Cents	10 Cents	5 Cents
30-sec frames	19.09	18.13	26.45	26.26	37.62	35.70
60-sec frames	11.10	12.33	15.85	15.95	27.80	22.93
120-sec frames	8.50	10.85	12.50	11.86	19.17	18.64

Error rates calculated using KPD with NNB and 15-cent precision.

Results for Raga Recognition for different configurable parameters

CONCLUSION AND SCOPE FOR FUTURE WORK

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- Evidence that melodic estimation is possible with good accuracy in a complex musical genre with continuous pitch movements and diversity of scale types
- Rich, fine-grained pitch distributions performed significantly better than PCD. Out of all methods, k-NN classification with Bhattacharya distance easily outperformed the others
- Some ragas which were misclassified, e.g. Desh and Khamaj, Asavari and Darbari etc. are sometimes difficult even for seasoned listeners
- Modeling sequential information using HMMs can be an improvement for this approach. N-gram modeling is also a promising option

REFERENCES

- [1] S. Chakraborty, G. Mazzola, S. Tewari, and M. Patra, 'Computational Musicology in Hindustani Music'.
- [2] P. Chordia and S. Sentürk, 'Joint Recognition of Raag and Tonic in North Indian Music'
- [3] S. Gulati, J. Serra, V. Ishwar, S. Sentürk, and X. Serra, 'Phrase-based raga recognition using vector space modeling'
- [4] Blondel, Vincent D., Jean-Loup Guillaume, Renaud Lambiotte, and Etienne Lefebvre. 'Fast unfolding of communities in large networks'
- [5] Camacho Lozano, Arturo. 'SWIPE: A sawtooth waveform inspired pitch estimator for speech and music.'
- [6] Duda, Richard O., Peter E. Hart, and David G. Stork. 'Pattern classification'. John Wiley & Sons, 2012.



THANK YOU FOR YOUR ATTENTION

QUESTIONS ?

WHAT IS A RAGA?

Video Link for PDF Version: <https://goo.gl/photos/jWR17cxJnUDpdSd77>

Western Classical Music

1. Concerts are never performed as an extempore
2. Percussion is not very important
3. The tonic never changes. E.g. a composition in C minor will always be performed in C minor
4. Major-Minor tonal systems, harmonies and counterpoints are important

Hindustani Music

1. The composition is always improvised upon. There is no script
2. Percussion is extremely important
3. The pitch of the tonic changes according to the instrument/voice but relative distances are constant
4. Harmony is not emphasized, there are no counterpoints, generally a single-melody instrument/voice



APPROACH 2 in BRIEF

- Uses the vector-space modeling analogy from Text Information Retrieval
- Motivation behind the approach is the way seasoned listeners identify ragas
- Melodic patterns from a collection of audio recordings in an unsupervised way
- Similar patterns are grouped, a directed graph is created based on similarity thresholds and clustered using a Community Detection Method^[4]
- Each recording is represented as a Vector, frequency and inverse frequencies are extracted and fed to a classifier such as SVM/Logistic Regression

