

swaras played and hence the raga, the onset time of each note and the scale in which it is played.

The system architecture is then based on 4 major parts:

- Feature extraction and Swara Segmentation
- Swara Identification module
- Raga Identification
- Database of Ragas

A. Feature Extraction and Swara Segmentation

Any music recognition algorithm requires the segmentation of the audio signal into separate events to detect different musical notes. For this, initially the audio signal was sampled at a rate of 16 KHz and then divided into different frames [4].

1) Design challenges

The harmonium is different from other keyboard instruments in the sense that, the note or swara is played continuously. Also the time duration for which a swara is played varies according to the mood of the artist and the requirements of the performance. Hence proper onset detection and end-point detection is a very critical aspect for this system.

2) Onset detection

In this paper, we tested two different approaches for onset detection viz. using spectral flux and using fundamental frequency estimation [5]-[7].

a) Spectral flux method

In this method the change in spectral energy from each frame to the next is calculated using spectral flux. Spectral flux (SF) is defined as the squared difference between the normalized magnitudes of successive spectral distribution. It is given by equation (1).

$$SF = \sum_{k=0}^{K-1} (|X_n(k)| - |X_{n-1}(k)|)^2 \quad (1)$$

where $n-1$ and n are the frame indices; k is FFT bin number. $X_n(k)$ is the FFT of n^{th} frame.

The spectral energy peaks obtained after processing the signal using spectral flux method denote the onset of new note. But this method can lead to false detection of the swara and is hence, less accurate. To achieve greater accuracy, it is necessary to go for a different approach.

b) Fundamental frequency estimation

This approach relies on joint time-frequency domain analysis of each frame [4]. The underlying assumption in most music processing schemes is that the properties of the music signal change relatively slowly with time. This assumption leads to a variety of short-time processing methods in which short segments of the music signal are isolated and processed as if they were short segments from a sustained sound with fixed properties. This is repeated usually periodically as often as desired. Often these short segments, which are sometimes called analysis frames, overlap one another. In present investigation we have taken analysis frame of 30 ms with 20 ms overlap giving a frame rate of 100 frames/second since the harmonium fundamental frequency range is from 100 Hz - 1000 Hz. The criterion for selection of 30 ms frame size is to have at

least 2-3 cycles of fundamental frequency of swara in order to have accurate estimation. The fundamental frequency or pitch frequency for each frame is obtained by applying autocorrelation function.

Autocorrelation function provides a measure of self-similarity between different alignments of the sequence. Autocorrelation of signal is expressed as in equation (2).

$$r_{xx}(\tau) = \sum_{m=0}^{M-1} x(m)x(m+\tau) \quad (2)$$

where,

$r_{xx}(\tau)$ = Autocorrelation of signal $x(m)$.

$x(m)$ = m^{th} speech sample, and τ = sample lag.

Autocorrelation based fundamental frequency estimation algorithm is reliable and the pitch frequencies achieved are quite accurate. Table I shows the fundamental frequencies of 36 swaras in an ascending order. The 36 swaras came from 3 octaves namely lower, middle and upper. The table also list the popular name of swara and symbol, which is used for string detection algorithm. Each swara is represented by two characters, the first one indicating actual position in the octave and the second letter (L, M, U) representing the lower, middle or upper octave respectively. Position of 12 swaras in an octave in an ascending Order: S, r, R, g, G, M, m, P, d, D, n, N; where capital letters S, R, G, M, P, D, N indicate normal swaras, lower case letters 'r, g, d, n' indicate komal swaras and 'm' indicates teevra swara.

TABLE I: THE FUNDAMENTAL FREQUENCIES OF 36 SWARAS IN AN ASCENDING ORDER

Sr. no.	Octave	Swara Name	Symbol	Frequency (Hz)
1	Lower	Shadja	SL	130
2	Lower	Komal Rishab	rL	141
3	Lower	Shuddha Rishab	RL	148
4	Lower	Komal Gandhar	gL	158
5	Lower	Shuddha Gandhar	GL	167
6	Lower	Shuddha Madhyam	ML	176
7	Lower	Tivra Madhyam	mL	186
8	Lower	Pancham	PL	196
9	Lower	Komal Dhaivat	dL	209
10	Lower	Shuddha Dhaivat	DL	222
11	Lower	Komal Nishad	nL	235
12	Lower	Shuddha Nishad	NL	251
13	Middle	Shadja	SM	260
14	Middle	Komal Rishab	rM	280
15	Middle	Shuddha Rishab	RM	296
16	Middle	Komal Gandhar	gM	314
17	Middle	Shuddha Gandhar	GM	332
18	Middle	Shuddha Madhyam	MM	351
19	Middle	Tivra Madhyam	mM	372
20	Middle	Pancham	PM	394
21	Middle	Komal Dhaivat	dM	419
22	Middle	Shuddha Dhaivat	DM	443
23	Middle	Komal Nishad	nM	470
24	Middle	Shuddha Nishad	NM	500
25	Upper	Shadja	SU	520
26	Upper	Komal Rishab	rU	558
27	Upper	Shuddha Rishab	RU	590
28	Upper	Komal Gandhar	gU	626
29	Upper	Shuddha Gandhar	GU	664
30	Upper	Shuddha Madhyam	MU	702
31	Upper	Tivra Madhyam	mU	744
32	Upper	Pancham	PU	787
33	Upper	Komal Dhaivat	dU	837
34	Upper	Shuddha Dhaivat	DU	884
35	Upper	Komal Nishad	nU	941
36	Upper	Shuddha Nishad	NU	995

Fig. 3(a) shows a typical signal $x(m)$, consisting of 12 *swaras* of middle octave in ascending order. The signal is then divided into frames and the corresponding pitch frequency of each frame is as shown in Fig. 3(b). The range of pitch frequencies for middle octave is from 260 Hz to 500 Hz.

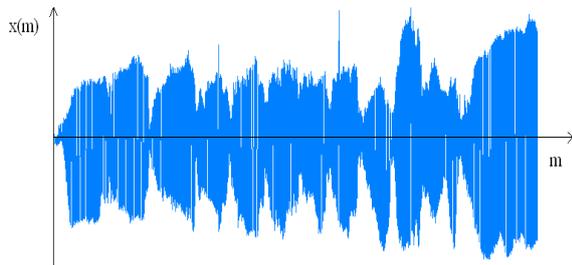


Fig. 3(a). Middle octave 12 swaras in ascending order.

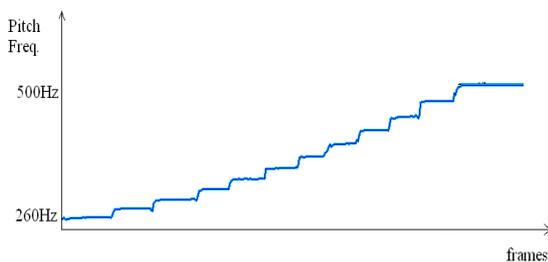


Fig. 3(b). Corresponding pitch frequencies of frames (Range of 260-500Hz).

The gradient function applied to the pitch frequency gives the proper onset detection peaks as shown in Fig.4.

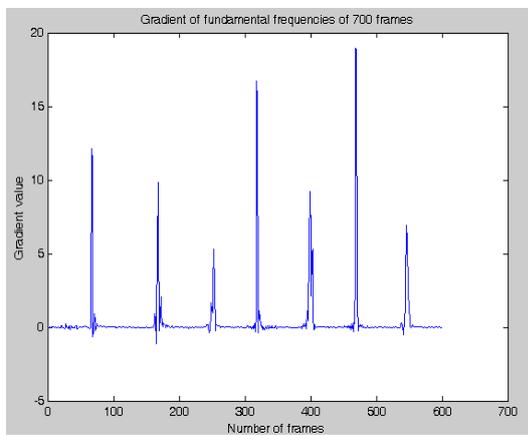


Fig. 4. Gradient of fundamental/pitch frequencies showing onset detection of 7 swaras.

Frame prior to the onset of next *swara* is considered as a last frame (end-point) of previous *swara*. Hence no separate end-point detection is necessary. The pitch frequencies thus obtained are comparable to those obtained by using Praat software [8]. As can be seen, sharp peaks are obtained which correspond to the onset of a new *swara*. This approach has very good accuracy and there is hardly any false detection. Hence, we decided to continue with this approach.

B. Swara Identification

As said earlier, in the feature extraction algorithm, pitch frequency of *swaras* is extracted. These frequencies are further used to identify each incoming musical note as a

particular *swara*. For this, initially, a database is created which contains all the possible *swaras* with their corresponding values of pitch frequencies. The different pitch frequencies present in the audio file are mapped to the *swaras* according to prepared database [1]. The tolerance permitted for fundamental frequency of each *swara* is ± 2 Hz from the standard frequency given in table I. This tolerance range has given almost 100% accurate identification accuracy which is an essential part for highly accurate *raga* identification [3], [9].

C. Raga Identification

Every *raga* has two distinct patterns viz. the first pattern which is played one way going up (from *swaras* of lower frequencies to those of higher frequencies) called '*arohi*' and the other which is played one way going down called '*avarohi*'.

In live concerts, the entire *aarohi* or *avrohi* is very rarely presented; instead different rhythmic sequential combinations of the *swaras* from the *aarohi* or *avrohi* of that *raga* are played on the harmonium. These melodic combinations though apparently random, are always within the grammar restrictions of that particular *raga* with the result that some combinations are unique to that *raga* [2]. This concept is used for making the structure of *raga* which has as many numbers of templates as allowed by the grammar of that *raga*. Identification of *raga* problem is thus brought down to a simple template matching algorithm. In this algorithm dynamic programming method is used to compute the minimum edit distance between the input *swara* string and all the templates present in the *ragas* database structure as given in II (D).

D. Database of Ragas

A database is specially structured for eight different popular *ragas* viz. '*Bheempalas*', '*Bageshri*', '*Yaman*', '*Bhoop*', '*Durga*', '*Des*', '*Kafee*' and '*Khamaj*' along with their identifying melodic combinations. A template is grammatically valid *swara* combination. A *raga* is a series of such templates. Popularly used *swara* combinations by the artists for the four *ragas* under investigation are listed below. Each template is separated by curly brackets, which represents a silence portion in the input audio. A *raga* structure used in *raga* identification algorithm is as shown below.

Raga Bheempalas : {{'nL'}, {'SM'}}, {{'gM'}, {'MM'}, {'PM'}, {'gM'}, {'MM'}}, {{'gM'}}, {{'RM'}, {'SM'}}, {{'gM'}, {'RM'}, {'SM'}}, {{'MM'}, {'PM'}} , {{'gM'}, {'MM'}, {'PM'}} , {{'MM'}}, {{'RM'}}, {{'SM'}}, {{'nL'}, {'SM'}, {'gM'}}, {{'PM'}, {'nM'}}, {{'DM'}, {'PM'}}, {{'nM'}, {'DM'}, {'PM'}}, {{'gM'}, {'MM'}}, {{'nL'}, {'SM'}, {'gM'}, {'MM'}}, {{'PM'}, {'nM'}, {'SU'}}, {{'nM'}, {'SU'}, {'gU'}, {'RU'}, {'SU'}}

Raga Bageshri : {{'nL'}, {'DM'}, {'nL'}, {'SM'}}, {{'AB'}, {'nL'}, {'DL'}, {'nL'}, {'RM'}, {'nM'}, {'SM'}}, {{'DM'}, {'nL'}, {'SM'}, {'gM'}, {'RM'}, {'SM'}}, {{'SM'}}, {{'SM'}, {'gM'}, {'MM'}}, {{'MM'}, {'gM'}, {'RM'}, {'SM'}, {'nM'}, {'SM'}}, {{'SM'}, {'gM'}, {'MM'}, {'DM'}} , {{'GM'}, {'MM'}, {'DM'}} , {{'MM'}, {'DM'}, {'MM'}, {'DM'}} , {{'gM'}, {'DM'}, {'gM'}, {'MM'}}, {{'SM'}, {'gM'}, {'MM'}, {'gM'}, {'RM'}, {'SM'}}, {{'gM'}, {'MM'}, {'DM'}, {'nM'}, {'nM'}, {'DM'}}, {{'MM'}, {'nM'}, {'DM'}}, {{'gM'}, {'DM'}, {'MM'}}, {{'SM'}, {'MM'}, {'gM'}, {'RM'}, {'SM'}}, {{'gM'}, {'MM'}, {'DM'}, {'nM'}, {'SU'}}, {{'MM'}, {'DM'}, {'nM'}, {'SU'}}, {{'SU'}, {'RU'}, {'SU'}, {'SU'}, {'SU'}, {'DM'}, {'MM'}}, {{'CD'}}, {{'gM'}, {'MM'}}, {{'gM'}, {'RM'}, {'SM'}}

Raga Yaman : {{'SM'}, {'NL'}, {'DL'}, {'NL'}, {'SM'}}, {{'NL'}, {'RM'}, {'SM'}}, {{'NL'}, {'RM'}, {'GM'}}, {{'RM'}, {'GM'}} ,

harmonium performance that is around 10-12 swaras/second.

The current algorithm for raga identification does not assign any priorities to the templates in each raga in the database. A method to give preference or weighted probability to the most popularly used combinations in a raga can be developed in further research. As the computational time requirement for template matching program is proportional to number of templates in database a synchronous search algorithm will be developed which will process all the templates simultaneously which will make it independent of number of templates in the database and hence enabling a real-time applications.

In future, this system may also be used for more generic purposes by using a filtering mechanism initially which will filter out only the audio signals pertaining to the harmonium from a medley of signals normally present in an Indian Classical orchestra.

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robotics etc.

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