

COMPARATIVE EVALUATION AND COMBINATION OF AUDIO TEMPO ESTIMATION APPROACHES

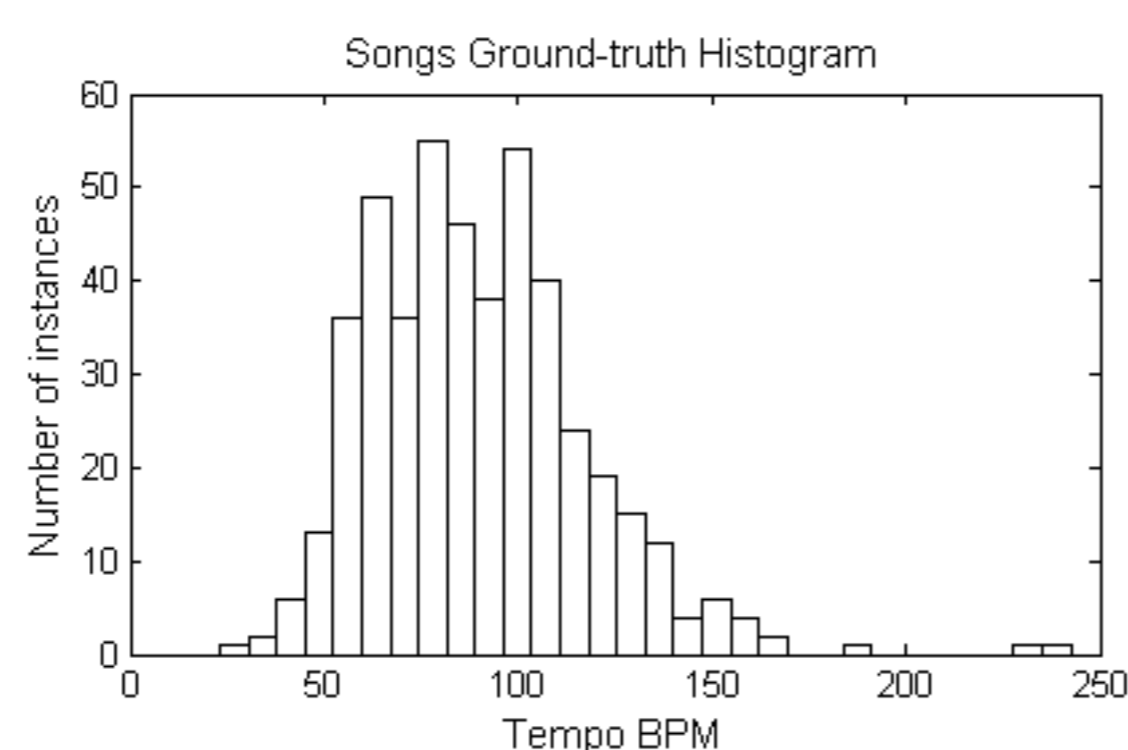
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APPROACHES

Algorithm	Aubio	BeatIt	Beatroot	Ellis	Essentia	Fixedtempo	IBT	jAudio	MIRTempo	Mpeg7-xm	Qmtempo	Tzanetakis
Author	Brossier Sonic annotator	Bonada and Gouyon Windows Binary	Dixon	Ellis	Aylon and Wack	Cannan	Oliveira et al.	McEnnis and McKay	Lartillot	Rohden	Davies and Plumbley	Tzanetakis
Infrastructure			Java	Matlab	Essentia	Sonic annotator	Marsyas	Java	Matlab	Matlab	Sonic annotator	Marsyas
Output	Beats Positions in Time	One Bpm	Beats Positions in Time	slower and faster tempo, relation between them	One Bpm	One Bpm	One Bpm	One Bpm	One Bpm	BPM value each time the estimated tempo changes	BPM value each time the estimated tempo changes	slower and faster tempo, relation between them
Feature list	Onset detection (Complex spectral difference)	Energy envelope differences for 8 bands	Onset detection (Spectral Flux)	Onsets obtained from the Mel spectrogram	Energy enveloped differences (Novelty curves) for 5 bands	Overall energy rise function.	Onset detection (Spectral Flux)	Energy envelope (256 window)	Onset curve, (10 channel gammatone filterbank and a low pass filter)	Energy envelopes for 6 bands	Onset detection (Spectral Flux)	Onset strength signal (based on Spectral Flux)
Pulse induction	ACF	ACF	IOI Clustering	ACF	ACF	ACF	ACF	ACF	ACF	ACF	ACF	ACF
Algorithm	AlonsoACF	AlonsoSP	DixonACF	DixonI	DixonT	Klapuri	Scheirer	Tzan_H	Tzan_MM	Tzan_MS	Uhle	Combination
Author	Alonso	Alonso	Dixon	Dixon	Dixon	Klapuri	Scheirer	Tzanetakis	Tzanetakis	Tzanetakis	Uhle	
Infrastructure	Matlab	Matlab	Matlab	Java	Java	Linux binary	Linux binary	Linux binary	Linux binary	Linux binary	Windows Binary	
Output	One Bpm	One Bpm	One Bpm	One Bpm	One Bpm	Beats in Time	Beats in Time	One Bpm	One Bpm	One Bpm	One Bpm	Heuristic combination of the results of Klapuri, BeatIt, Qmtempo, Ellis, IBT(c),Essentia and Mpeg7-xm
Feature list	Onsets of Notes	Onsets of Notes	Downsampled and smoothed Energy of 8 frequency bands	Energy based Onset Detector	Energy based Onset Detector	The differentials of the loudness in 36 frequency subbands	Energy envelope differences for 6 bands computed by a filterbank	Energy Envelopes of 5 octave – space frequency bands obtained by Wavelets	Energy Envelopes of 5 octave – space frequency bands obtained by Wavelets	Energy Envelopes of 5 octave – space frequency bands obtained by Wavelets	Differential smoothed Energy envelopes for logarithmical ly space frequency bands	
Pulse induction	ACF	Spectral Product	ACF	IOI Clustering	IOI Clustering	Bankcomb Filter	Bankcomb Filter	ACF	ACF	ACF	ACF	

MUSIC COLLECTION

Genre	# Songs
Rock	68
Classical	70
Electronic	59
Latin	44
Samba	42
Jazz	12
Afrobeat	3
Flamenco	13
Balkan and Greek	144
Fado	10



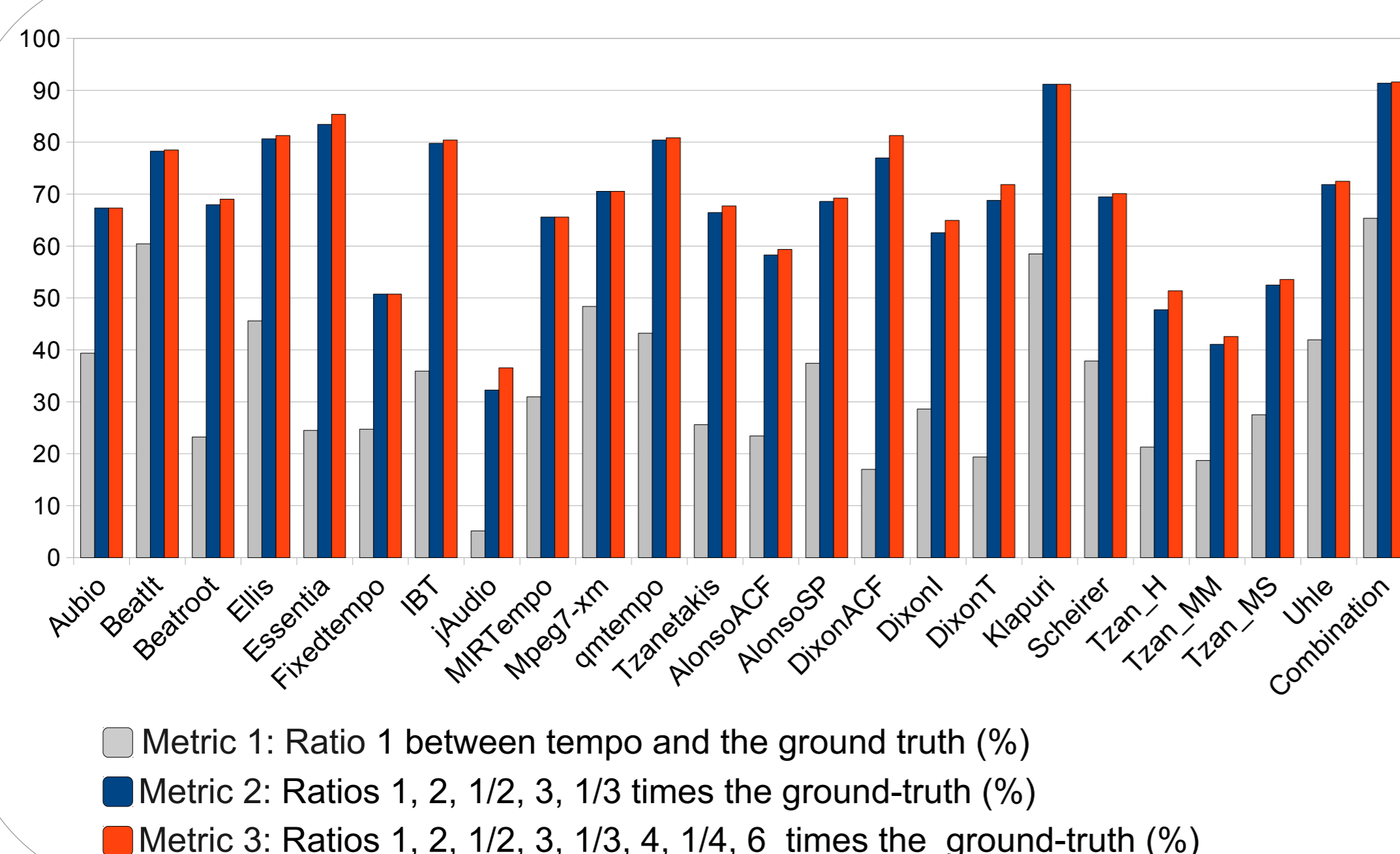
COMBINATION METHOD

```

Main() {
for n = 1:number_of_songs
Result(n) = klapuri(n);
if (Result(n) == BeatIt(n) & Result(n) == Essentia(n))
[p, Out] = Relation(n);
if (p == 1)
Result(n) = Out;
end
% To avoid the double error
if (Result(n) == BeatIt(n)*2 & BeatIt(n) == Ellis(n))
Result(n) = BeatIt(n);
end
}

function [p, Out] = Relation(n)
{ array = [BeatIt(n), Ellis(n), Qmtempo(n),
IBT(n), Essentia(n), Mpeg7-xm(n)];
if three or more values from array are equal and
if exist the double of these values in the array
p = 1;
Out = the repeat value in the array;
else
p = 0; Out = 0;
end
}
    
```

RESULTS



Algorithm	Pulse Induction	Combing Bands
Klapuri	B	After
BeatIt	A	After
Ellis	A	
qmtempo	A	
IBT	A	
Essentia	A	After
Mpeg7-xm	A	After
Uhle	A	After
Scheirer	B	After
AlonsoSP	SP	
DixonACF	A	Before
Aubio	A	
MIRTempo	A	Before
Beatroot	IOI	
DixonT	IOI	
Tzanetakis	A	
DixonI	IOI	
AlonsoACF	A	
Tzan_MS	A	Before
Fixedtempo	A	
Tzan_H	A	Before
Tzan_MM	A	After
jAudio	A	

General Ratio	%
1/4	0,03%
1/3	0,09%
1/2	3,62%
2/3	1,89%
3/4	0,62%
Good	32,12%
5/4	1,02%
4/3	5,36%
3/2	1,88%
2	29,00%
9/4	1,65%
5/2	1,14%
8/3	1,37%
3	1,80%
4	1,25%
16/3	0,07%
6	0,15%
Other	16,94%

CONCLUSIONS

- The best estimation had an overall accuracy of 91% but can be improved with a heuristic combination of the best methods
- The best performing methods (Klapuri and BeatIt) are based on the following steps: differential energy values, frequency decomposition, periodicity detection prior to the multi-band integration, and multi-tempo level detection in the pulse-induction block.
- The tested algorithms involving band decomposition and periodicity detection before the multi-band integration achieved better results.

REFERENCES

- F. Gouyon, A. Klapuri, S. Dixon, M. Alonso, G. Tzanetakis, C. Uhle, and P. Cano, "An experimental comparison of audio tempo induction methods," *Audio, Speech, and Language Processing, IEEE Transactions on*, vol. 14, 2006, p. 1832–1844.
- F. Gouyon and S. Dixon, "A Review of Automatic Rhythm Description Systems," *Computer Music Journal*, vol. 29, 2005, pp. 34-54.
- <https://sites.google.com/site/tempoandbeattracking/>