

# Rhythm Classification Using Spectral Rhythm Patterns



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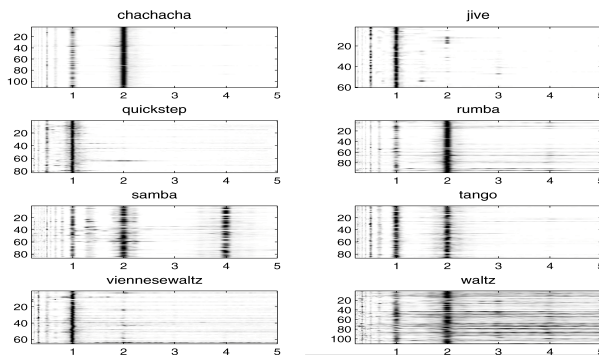
IRCAM - Sound Analysis/Synthesis Team - Semantic HiFi

## Objectives

- Study the use of spectral patterns to represent the characteristics of the rhythm
- Three spectral patterns derived from the onset function
  - Discrete Fourier Transform
  - Auto-Correlation Function
  - Product of DFT and Frequency-Mapped ACF
- Evaluation for the task of rhythm classification

## State of the art

- **Rhythm representations from audio signal**
  - type of information being represented,
  - how they are represented
- **Footo 2001** : beat spectrum
- **Tzanetakis 2002** : beat histogram
- **Paulus 2002** : sequence of audio features, Dynamic Time Warping
- **Gouyon 2004** : 73 features from periodicity histogram, Inter-Onset-Interval Histogram, 8 music genres of ballroom dancer database: 90,1% (ground-truth tempo), 78,9% (estimated tempo)
- **Gouyon 2004**: tempo estimation errors: 67,8 %
- **Dixon 2004**: Gouyon + temporal rhythmic patterns (energy evolution inside a bar): 96% (pattern+all features), 50% (only pattern)



Spectral rhythm patterns

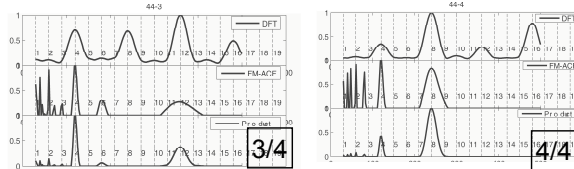
## Proposed method

### 1. Tempo estimation

- **Onset energy function**: reassigned spectral energy flux (see ICMC 2005)
- **Periodicities estimation**:
  - DFT,
  - ACF,
  - product DFT and Frequency-Mapped ACF
- **Product DFT - FM-ACF ?**

Two measures of periodicities: DFT  $F(w_k, t_i)$ , ACF  $A(l, t_i)$  with inverse octave uncertainties -> combined both

  - 1)  $l$  is mapped to the frequency domain:  $sr/w_k$ . In order to obtain the same frequencies as for the DFT: interpolation of ACF at  $l=sr/w_k$
  - 2) Compute the product of the DFT and ACF at each frequency  $w_k$



- **Tempo estimation**: usage of the ground-truth tempo

### 2. Spectral Rhythm Patterns

- **Rhythm**: position, duration, acoustical properties
  - **Here**: representation of sequence of duration
  - **Sensitiveness to the sequence of duration**
    - obtained through complex DFT phase relationships
  - **Independence of the tempo**
    - $Y(w_k, t_i)$  = either the DFT, ACF or DFT/FM-ACF
    - $w_{bpm}(t_i)$  = the current tempo
    - Normalized frequencies  $w_k' = w_k / w_{bpm}(t_i)$  -> resampling
    - Mean of  $Y(w_k', t_i)$  over time
    - Normalization to unit sum
  - **Compactness**
    - keep only musically meaningful frequencies: 1/4, 1/3, 1/2, 2/3, 3/4, 1, 1.25, 1.5, 1.75, 2, 2.25, 2.75, 3, 3.25, 3.5, 3.75, 4
    - lower components = measure subdivision
    - upper components = beat subdivision
- = **Spectral Rhythm Patterns**

## Evaluation: Music Genre Classification

- **Data**: Ballroom dancer database, 698 tracks, 30 s., 8 music genres (ChaChaCha, Jive, Quickstep, Rumba, Samba, Tango, VienneseWaltz, Waltz)
- **Features**:
  - DFT (18 dim.) / + tempo (19 dim.)
  - ACF (18 dim.) / + tempo (19 dim.)
  - product DFT/FM-ACF (18 dim.) / + tempo (19 dim.)
- **Classification algorithm**:
  - C4.5 decision tree algorithm,
  - Partial Decision Tree algorithm,
  - Classification via Regression algorithm
- **Results**:
  - Best classifiers: Classification via Regression
  - Best feature set: DFT
- **Comparison to the state of the art**:
  - Here: without tempo **81%**, with tempo **90,4%**
  - Gouyon 79,6%, 90,1%
  - Dixon: 50% (only pattern), 96%

	J48	PART	ClassViaReg
DFT	75,64	73,78	<b>80,8</b>
ACF	69,34	70,34	76,64
DFT/FM-ACF	65,9	65,32	75,5
DFT + tempo	90,4	88,96	<b>90,4</b>
ACF + tempo	86,67	86,67	90,25
DFT/FM-ACF + tempo	86,38	86,24	90,25
tempo	77,79	77,36	77,93

### Confusion matrix

classified as -->	C	J	Q	R	S	T	VW	W
ChaChaCha	<b>87,4%</b>			4,5%	0,9%	7,2%		
Jive		<b>86,7%</b>	1,7%			6,7%	5,0%	
Quickstep		1,2%	<b>97,6%</b>	1,2%				
Rumba	2,0%			<b>79,6%</b>	2,0%	3,1%		13,3%
Samba	1,2%			7,0%	<b>89,5%</b>	1,2%		1,2%
Tango	3,5%			1,2%	1,2%	<b>94,2%</b>		
Viennese Waltz		3,1%					<b>95,4%</b>	
Waltz		0,9%		4,5%				<b>94,5%</b>

- **Best features** (CFS algorithm): 1/3, 2/3, 1, 2, 3, 3.75, 4. Recognition rate: 75.5%, 89,54%

## Conclusion

- The use of simple spectral patterns allows to achieve a high recognition rate (close to the results obtained with more complex methods proposed so far)
- Future works: use estimated tempo, Evaluation on a larger set of music genres