

Recursive versions of the Levenberg-Marquardt reassigned spectrogram and of the synchrosqueezed STFT

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Outline

- 1 Introduction
- 2 Filter-based reassigned and synchrosqueezed STFT
- 3 Recursive implementation
- 4 Numerical results
- 5 Conclusion and Future works

STFT implemented as a recursive filtering

Goals

Compute time-frequency representations

- candidate for real-time implementation
- allow adjustments from the user (Levenberg-Marquardt approach)
- allow modes separation and signal reconstruction (synchrosqueezing)
- filter bank approach

Principle

- special case of the STFT using a causal, infinite length window function that can be rewritten as a causal IIR recursive filtering
- the algorithmic complexity depends on the filter order and on the analyzed frequency bandwidth

State-of-the-art

Computing the STFT in terms of recursions

- Recursive STFT [Chen et al.1993], [Amin and Feng, 1995], [S. Tomazic and S.Znidar,1996, 1997]
- STFT and recursive reassignment (implemented using approximative finite differences) [Richard and Lengellé, 1997]
- **STFT and recursive reassignment using a causal infinite window** [Nilsen, 2009]

Reassignment

- [Kodera, 1976], [Auger, Flandrin, 1995]
- Levenberg-Marquardt reassigned spectrogram [Auger, Chassande-Mottin and Flandrin2012]

Synchrosqueezing

- synchrosqueezed CWT [Daubechies and Maes, 1996]
- synchrosqueezed STFT [Thakur and Wu, 2011], [Auger et al., 2013]

The STFT as a convolution product

The STFT of a signal x using a window h , denoted $F_x^h(t, \omega) = M_x^h(t, \omega) e^{j\Phi_x^h(t, \omega)}$ can be related to the linear convolution product between the analyzed signal x and the complex valued impulse response of a bandpass filter centered on ω , $g(t, \omega) = h(t) e^{j\omega t}$, $h(t)$ being a real-valued analysis window:

$$y_x^g(t, \omega) = \int_{-\infty}^{+\infty} g(\tau, \omega) x(t - \tau) d\tau = |y_x^g(t, \omega)| e^{j\Psi_x^g(t, \omega)} \quad (1)$$

$$= F_x^h(t, \omega) e^{j\omega t} = M_x^h(t, \omega) e^{j(\Phi_x^h(t, \omega) + \omega t)} \quad (2)$$

The STFT is related to the convolution product as

$$M_x^h(t, \omega) = |y_x^g(t, \omega)| \quad (3)$$

$$\Phi_x^h(t, \omega) = \Psi_x^g(t, \omega) - \omega t \quad (4)$$

Rewording the reassignment operators of the spectrogram

According to their definition [Kodera, 1976], [Auger, Flandrin, 1995], the spectrogram reassignment operators can be related to the phase of the STFT and can be reformulated using the phase of $y_x^g(t, \omega)$, denoted

$$\Psi_x^g(t, \omega) = \Phi_x^h(t, \omega) + \omega t$$

$$\hat{t}(t, \omega) = -\frac{\partial \Phi_x^h}{\partial \omega}(t, \omega) = t - \frac{\partial \Psi_x^g}{\partial \omega}(t, \omega), \quad (5)$$

$$\hat{\omega}(t, \omega) = \omega + \frac{\partial \Phi_x^h}{\partial t}(t, \omega) = \frac{\partial \Psi_x^g}{\partial t}(t, \omega). \quad (6)$$

Reassigned spectrogram

$$\text{RSP}(t, \omega) = \iint_{\mathbb{R}^2} |y_x^g(t', \omega')|^2 \delta(t - \hat{t}(t', \omega')) \delta(\omega - \hat{\omega}(t', \omega')) dt' d\omega' \quad (7)$$

where $\delta(t)$ denotes the Dirac distribution.

Rewording the Levenberg-Marquardt reassignment

[Auger, Chassande-Mottin and Flandrin, 2012]

Levenberg-Marquardt reassignment operators

$$\begin{pmatrix} \tilde{t}(t, \omega) \\ \tilde{\omega}(t, \omega) \end{pmatrix} = \begin{pmatrix} t \\ \omega \end{pmatrix} - \left(\nabla^t R_x^h(t, \omega) + \mu I_2 \right)^{-1} R_x^h(t, \omega) \quad (8)$$

with $R_x^h(t, \omega) = \begin{pmatrix} t - \hat{t}(t, \omega) \\ \omega - \hat{\omega}(t, \omega) \end{pmatrix} = \begin{pmatrix} \frac{\partial \Psi_x^g}{\partial \omega}(t, \omega) \\ \omega - \frac{\partial \Psi_x^g}{\partial t}(t, \omega) \end{pmatrix}$

$$\nabla^t R_x^h(t, \omega) = \begin{pmatrix} \frac{\partial R_x^h}{\partial t}(t, \omega) & \frac{\partial R_x^h}{\partial \omega}(t, \omega) \end{pmatrix} = \begin{pmatrix} \frac{\partial^2 \Psi_x^g}{\partial t \partial \omega}(t, \omega) & \frac{\partial^2 \Psi_x^g}{\partial \omega^2}(t, \omega) \\ -\frac{\partial^2 \Psi_x^g}{\partial t^2}(t, \omega) & 1 - \frac{\partial^2 \Psi_x^g}{\partial t \partial \omega}(t, \omega) \end{pmatrix}$$

where I_2 is the 2×2 identity matrix.

Levenberg-Marquardt reassigned spectrogram

$$\text{LMRSP}(t, \omega) = \iint_{\mathbb{R}^2} |y_x^g(t', \omega')|^2 \delta(t - \tilde{t}(t', \omega')) \delta(\omega - \tilde{\omega}(t', \omega')) dt' d\omega' \quad (9)$$

Rewording the partial derivatives of the phase

$$\frac{\partial \Psi_x^g}{\partial t}(t, \omega) = \text{Im} \left(\frac{y_x^{\mathcal{D}g}(t, \omega)}{y_x^g(t, \omega)} \right) \quad (10)$$

$$\frac{\partial \Psi_x^g}{\partial \omega}(t, \omega) = \text{Re} \left(\frac{y_x^{\mathcal{T}g}(t, \omega)}{y_x^g(t, \omega)} \right) \quad (11)$$

$$\frac{\partial^2 \Psi_x^g}{\partial t \partial \omega}(t, \omega) = \text{Re} \left(\frac{y_x^{\mathcal{D}\mathcal{T}g}(t, \omega)}{y_x^g(t, \omega)} - \frac{y_x^{\mathcal{D}g}(t, \omega) y_x^{\mathcal{T}g}(t, \omega)}{y_x^g(t, \omega)^2} \right) \quad (12)$$

$$\frac{\partial^2 \Psi_x^g}{\partial t^2}(t, \omega) = \text{Im} \left(\frac{y_x^{\mathcal{D}^2g}(t, \omega)}{y_x^g(t, \omega)} - \left(\frac{y_x^{\mathcal{D}g}(t, \omega)}{y_x^g(t, \omega)} \right)^2 \right) \quad (13)$$

$$\frac{\partial^2 \Psi_x^g}{\partial \omega^2}(t, \omega) = -\text{Im} \left(\frac{y_x^{\mathcal{T}^2g}(t, \omega)}{y_x^g(t, \omega)} - \left(\frac{y_x^{\mathcal{T}g}(t, \omega)}{y_x^g(t, \omega)} \right)^2 \right) \quad (14)$$

where y_x^g , $y_x^{\mathcal{T}g}$, $y_x^{\mathcal{D}g}$, $y_x^{\mathcal{D}\mathcal{T}g}$, $y_x^{\mathcal{T}^2g}$ and $y_x^{\mathcal{D}^2g}$ are the outputs of the filters using respectively the impulse responses $g(t, \omega)$, $\mathcal{T}g = t g(t, \omega)$, $\mathcal{D}g(t, \omega) = \frac{\partial g}{\partial t}(t, \omega)$, $\mathcal{D}\mathcal{T}g(t, \omega) = \frac{\partial}{\partial t}(t g(t, \omega))$, $\mathcal{T}^2g(t, \omega) = t^2 g(t, \omega)$ and $\mathcal{D}^2g(t, \omega) = \frac{\partial^2 g}{\partial t^2}(t, \omega)$.

Rewording the synchrosqueezed STFT

As previously defined, y_x^g admits the following signal reconstruction formula

$$x(t - t_0) = \frac{1}{h(t_0)} \int_{-\infty}^{+\infty} y_x^g(t, \omega) e^{-j\omega t_0} \frac{d\omega}{2\pi}, \quad \text{when } h(t_0) \neq 0. \quad (15)$$

Synchrosqueezed STFT [Thakur and Wu, 2011], [Auger et al., 2013]

$$Sy_x^g(t, \omega) = \int_{\mathbb{R}} y_x^g(t, \omega') e^{-j\omega' t_0} \delta(\omega - \hat{\omega}(t, \omega')) d\omega' \quad (16)$$

LMSy_x^g(t, ω) is obtained by replacing $\hat{\omega}$ by $\tilde{\omega}$.

Time-frequency representation

A sharpen time-frequency representation is provided by $|Sy_x^g(t, \omega)|^2$

Signal reconstruction from $Sy_x^g(t, \omega)$

$$\hat{x}(t - t_0) = \frac{1}{h(t_0)} \int_{\mathbb{R}} Sy_x^g(t, \omega) \frac{d\omega}{2\pi} \quad (17)$$

The STFT special case [Nilsen, 2009]

y_x^g can be recursively implemented using

$$h_k(t) = \frac{t^{k-1}}{T^k(k-1)!} e^{-t/T} U(t), \quad (18)$$

$$g_k(t, \omega) = h_k(t) e^{j\omega t} = \frac{t^{k-1}}{T^k(k-1)!} e^{pt} U(t) \quad (19)$$

with $p = -\frac{1}{T} + j\omega$, $k \geq 1$ being the filter order, T the time spread of the window and $U(t)$ the Heaviside step function.

Discretization using the impulse invariance method [Jackson,2000]

$$G_k(z, \omega) = T_s \mathcal{Z} \{g_k(t, \omega)\} = \frac{\sum_{i=0}^{k-1} b_i z^{-i}}{1 + \sum_{i=1}^k a_i z^{-i}} \quad (20)$$

with $b_i = \frac{1}{L^{k(k-1)!}} B_{k-1, k-i-1} \alpha^i$, $\alpha = e^{pT_s}$, $L = T/T_s$, $a_i = A_{k,i} (-\alpha)^i$, T_s being the sampling period. $B_{k,i} = \sum_{j=0}^i (-1)^j A_{k+1,j} (i+1-j)^k$ denotes the Eulerian numbers and $A_{k,i} = \binom{k}{i} = \frac{k!}{i!(k-i)!}$ the binomial coefficients.

Hence, using $y_k[n, m] \approx y_x^{gk}(nT_s, \frac{2\pi m}{MT_s})$ with $n \in \mathbb{Z}$ and $m = 0, 1, \dots, M-1$, we obtain

$$y_k[n, m] = \sum_{i=0}^{k-1} b_i x[n-i] - \sum_{i=1}^k a_i y_k[n-i, m] \quad (21)$$

other specific impulse responses

Recursive relations between filters

$$T_s \mathcal{Z} \{ \mathcal{T} g_k(t, \omega) \} = k T G_{k+1}(z, \omega) \quad (22)$$

$$T_s \mathcal{Z} \{ \mathcal{D} g_k(t, \omega) \} = \frac{1}{T} G_{k-1}(z, \omega) + p G_k(z, \omega) \quad (23)$$

$$T_s \mathcal{Z} \{ \mathcal{D} \mathcal{T} g_k(t, \omega) \} = k (G_k(z, \omega) + p T G_{k+1}(z, \omega)) \quad (24)$$

$$T_s \mathcal{Z} \{ \mathcal{T}^2 g_k(t, \omega) \} = k(k+1) T^2 G_{k+2}(z, \omega) \quad (25)$$

$$T_s \mathcal{Z} \{ \mathcal{D}^2 g_k(t, \omega) \} = \frac{1}{T^2} G_{k-2}(z, \omega) + \frac{2p}{T} G_{k-1}(z, \omega) + p^2 G_k(z, \omega) \quad (26)$$

for any $k \geq 1$, provided that $G_0(z, \omega) = G_{-1}(z, \omega) = 0$.

Discrete-time recursive classical reassigned spectrogram

Reassignment operators

$$\hat{n}[n, m] = n - \text{Round} \left(\text{Re} \left(\frac{T_s^{-1} y_x^{Tg}[n, m]}{y_x^g[n, m]} \right) \right) \quad (27)$$

$$\hat{m}[n, m] = \text{Round} \left(\frac{M}{2\pi} \text{Im} \left(\frac{T_s y_x^{Dg}[n, m]}{y_x^g[n, m]} \right) \right) \quad (28)$$

Reassigned spectrogram

$$\text{RSP}_k[n, m] = \sum_{n' \in \mathbb{Z}} \sum_{m'=0}^{M-1} |y_k[n', m']|^2 \delta [n - \hat{n}[n', m']] \delta [m - \hat{m}[n', m']] \quad (29)$$

where $\delta[n]$ is the Kronecker delta

Discrete-time recursive Levenberg-Marquardt reassigned spectrogram

Levenberg-Marquardt reassignment operators

$$\begin{aligned} \tilde{n}[n, m] = n - \text{Round} & \left(\frac{1}{\Lambda} \left(T_s^{-1} \frac{\partial \Psi_x^g}{\partial \omega} \right) \left(1 + \mu - \frac{\partial^2 \Psi_x^g}{\partial t \partial \omega} \right) \right. \\ & \left. - \frac{1}{\Lambda} \left(T_s^{-2} \frac{\partial^2 \Psi_x^g}{\partial \omega^2} \right) \left(\frac{2\pi m}{M} - T_s \frac{\partial \Psi_x^g}{\partial t} \right) \right) \end{aligned} \quad (30)$$

$$\begin{aligned} \tilde{m}[n, m] = m - \text{Round} & \left(\frac{M}{2\pi\Lambda} \left(T_s^{-1} \frac{\partial \Psi_x^g}{\partial \omega} \right) \left(T_s^2 \frac{\partial^2 \Psi_x^g}{\partial t^2} \right) \right. \\ & \left. + \frac{1}{\Lambda} \left(\mu + \frac{\partial^2 \Psi_x^g}{\partial t \partial \omega} \right) \left(m - \frac{MT_s}{2\pi} \frac{\partial \Psi_x^g}{\partial t} \right) \right) \end{aligned} \quad (31)$$

with $\Lambda = \left(\mu + \frac{\partial^2 \Psi_x^g}{\partial t \partial \omega} \right) \left(\mu + 1 - \frac{\partial^2 \Psi_x^g}{\partial t \partial \omega} \right) + \left(T_s^2 \frac{\partial^2 \Psi_x^g}{\partial t^2} \right) \left(T_s^{-2} \frac{\partial^2 \Psi_x^g}{\partial \omega^2} \right)$

Hence, $\text{LMRSP}_k[n, m]$ is obtained by replacing (\hat{n}, \hat{m}) by (\tilde{n}, \tilde{m}) .

Discrete-time recursive synchrosqueezed STFT

Recursive synchrosqueezed STFT

$$S y_k[n, m] = \sum_{m'=0}^{M-1} y_k[n, m'] e^{-\frac{2j\pi m' n_0}{M}} \delta [m - \hat{m}[n, m']] \quad (32)$$

where $n_0 = t_0/T_s$ can be chosen as the time instant $t_0 > 0$ when the maximum of h_k is reached (*i.e.* $n_0 = (k - 1)L$). $LMSy_k[n, m]$ is obtained by replacing (\hat{n}, \hat{m}) by (\tilde{n}, \tilde{m}) .

Signal reconstruction from $Sy_k[n, m]$

$$\hat{x}[n - n_0] = \frac{1}{MT_s h_k(n_0 T_s)} \sum_{m=0}^{M-1} S y_k[n, m]. \quad (33)$$

For modes extraction, for example [Meignen et al., 2012] can be used.

Implementation issue

Algorithm for recursive TFR computation

At time index n

- 1 Compute the required $y_k^g[n, m]$ using $x[n - i]$ and $y_k[n - j, m]$ with $i \in [0, k - 1]$, $j \in [1, k]$
- 2 Compute the other required specific filtered signals (i.e. y_k^{Tg} , y_k^{Dg} , y_k^{DTg} , $y_k^{T^2g}$ or $y_k^{D^2g}$) using y_k^g with different filter orders
- 3 Compute \hat{n}, \hat{m} (resp. \tilde{n}, \tilde{m}) provided by the reassignment operators
- 4 If $\hat{n} \leq n$ (resp. $\tilde{n} \leq n$) then update TFR $[\hat{n}, m]$ otherwise store the triplet $(y_k^g[n, m], \hat{n}, m)$ into a list
- 5 Update TFR $[n, m]$ using all previously stored triplets verifying $\hat{n} \leq n$ and remove them from the list

Time-frequency representations

Signal reconstruction quality

Reconstruction Quality Factor

$$\text{RQF} = 10 \log_{10} \left(\frac{\sum_n |x[n]|^2}{\sum_n |x[n] - \hat{x}[n]|^2} \right) \quad (34)$$

(a)	n_0	8	18	26	28	30
	RQF (dB)	9.79	24.17	26.77	26.82	26.73
(b)	M	100	200	600	1000	2400
	RQF (dB)	20.56	24.90	29.48	30.50	30.87
(c)	μ	0.30	0.80	1.30	1.80	2.30
	RQF (dB)	20.83	27.28	29.68	30.35	30.90

Signal reconstruction quality factor of the recursive synchrosqueezed STFT computed for $k = 5$, $L = 7$ at $\text{SNR} = 45$ dB. Line (a), computed for $M = 300$, Line (b) and Line (c), computed for $n_0=28$ and $M=300$. Reference recursive STFT reconstruction RQF = **35.56** dB.

Results

- The first recursive implementation of the synchrosqueezing transform
- A new generalized STFT synthesis formula allowing reconstruction using a causal analysis window
- Adjustable methods allowing to choose a trade-off between the energy concentration and the signal reconstruction quality

Future works

- New practical applications (e.g. audio speech enhancement, Music Information Retrieval, Electrical loads recognition, etc.)
- Recursive second-order synchrosqueezing [Oberlin et al., 2015]
- Usage of a different analysis window