

# The synchrosqueezing generalized S-transform algorithm: Application to the Sichuan Basin Ying Hu<sup>1,2</sup>, Hui Chen<sup>1</sup>, and Kurt. J. Marfurt<sup>2</sup>, <sup>1</sup>Chengdu University of Technology; <sup>2</sup>University of Oklahoma

### Summary

A new time-frequency analysis method—Synchrosqueezing **GST** Generalized S-transform (SSGST)—is proposed to meet the needs of high-resolution seismic signal processing and interpretation. The basic wavelet of the generalized Stransform (GST) is a modulated harmonic wave with four undetermined parameters that can be constructed by adjusting the four parameters to make the GST more suitable for seismic signals processing. The SSGST method squeezes and reconstructs the complex coefficient spectra of GST results along the frequency direction so that the energy distributions on the time-frequency spectra are concentrated around the real instantaneous frequency of the signal; thus, the time-frequency resolution can be improved. Based on mathematical theory, we strictly deduced the mathematical expressions of the positive transformation and lossless inverse transformation of SSGST. The experimental results of numerical signals illustrate that the proposed method can correctly decompose signals with different spectral characteristics into a high time-frequency resolution spectrum and can recovery the original signal from the time-frequency spectrum with satisfying reconstructing accuracy. Application on field seismic data shows the superiority of the new method in seismic time-frequency analysis for hydrocarbon detection.

# Inroduction

Time-frequency analysis, a powerful tool for seismic data analysis, plays a significant role in oil and gas exploration and development. It maps a 1D signal in the time domain into a 2D time-frequency spectrum, which can effectively reveal important details of seismic data and provide valuable information for reservoir characterization.

Synchrosqueezing transform (SST) is a relatively new Equation (4) can be derived as follows: technique based on the combination of time-frequency methods followed by a reassignment step (Auger, 2013; Tary,  $\Box GS$ 2017) method. SST can improve the energy concentration of time-frequency representation by applying a post-processing reallocation to the original representation. At present, many efforts have been made to use this method for seismic data processing and interpretation.

Herrera et al. (2014) identified channels from seismic data using SST successfully. In 2015, they also used SST to separate the P and S waves of microseismic data. Mousavi et conjugate of Fourier Transform result of  $\psi(t)$ .  $\hat{x}(\omega)$  denotes al. (2016) used it for microseismic detection. Mousavi and the Fourier Transform result of the signal x(t). Langston (2017) showed that SST can be used to improve the denoising of seismic data. Tary et al. (2017) used it for attenuation estimation.

The generalized S-transform (GST) introduced by Gao et SS al. (2003) overcomes the dilemma of the fixed wavelet in ST by introducing four undetermined parameters (amplitude, energy decay rate, energy delay time, and video rate) to construct the basic wavelet adaptively to the non-stationary  $L_f$  is the half length of frequency range  $[f_l - L_f, f_l + L_f]$  centered signal characteristics in practical application. Due to having on the frequency point  $f_l$ .  $f_k$  is the discrete frequency points in no restriction of the time window length, GST can obtain real frequency ranges of the GST, and  $\Delta f_k = f_k - f_{k-1}$ . time-frequency spectra with excellent time-frequency resolution, which provides more possibility and higher accuracy for the detailed information extraction of complicated non-stationary signals.

Inspired by the theory of SST and the advantages of the GST, we propose a novel time-frequency analysis method that we havenamed synchrosqueezing generalized Stransform (SSGST).

# **Principles**

Gao et al. (2003) use a modulated harmonic wave with four undetermined parameters to replace the basic wavelet in ST to overcome the disadvantage caused by the fixed basic wavelet function. The modulated harmonic wave is written as,

$$w_f(t) = A \left| f \right| \exp \left[ -\alpha \left( ft - \beta \right)^2 - i2\pi f_0 ft \right]$$
(1)

So the generalized S-transform (GST) is:

$$ST_{x}(f,b) = A \left| f \right| \int_{-\infty}^{\infty} x(t) \exp\left\{-\alpha \left[ f(t-b) - \beta \right]^{2} \right\} \exp\left(-i2\pi f_{0}ft\right) dt$$
(2)

Where, x(t) represents a signal,  $GST_x$  denotes the generalized S-Transform of x(t). f is frequency, t is time and b is the time shift. A is amplitude of the basic wavelet,  $\alpha$  is energy attenuation ratio ( $\alpha > 0$ ),  $\beta$  is energy delay time and  $f_0$  is video frequency of the basic wavelet.

### SSGST

First, the Equation (2) can be reformulated as,

$$GST_{x}(f,b) = A |f| \exp(-i2\pi f_{0}fb)$$

$$\times \int_{-\infty}^{\infty} x(t) \exp\{-\alpha [f(t-b) - \beta]^{2}\} \exp[-i2\pi f_{0}f(t-b)]dt$$

$$(3)$$

t: 
$$\psi(t) = A \exp\left[-\alpha(t-\beta)\right] \exp(i2\pi f_0 t)$$

then Equation (3) is expressed as,

$$FST_{x}(f,b) = \left| f \right| \exp\left(-i2\pi f_{0}fb\right) \int_{-\infty}^{\infty} x(t) \overline{\psi\left[f(t-b)\right]} dt \qquad (4)$$

According to the Parseval theorem and transformation properties of scale and translation in Fourier Transform, the

$$ST_{x}(f,b) = \frac{1}{2\pi} \exp\left(-i2\pi f_{0}fb\right) \int_{-\infty}^{\infty} \hat{x}(\omega) \overline{\hat{\psi}(f^{-1}\omega)} e^{ib\omega} d\omega \quad (5)$$

Then, we can calculate the instantaneous frequency of the signal x(t) by using Equation (6):

$$f_{x}(f,b) = f_{0}f + [i2\pi GST_{x}(f,b)]^{-1} \frac{\partial GST_{x}(f,b)}{\partial b}$$
(6)

 $\overline{\psi(t)}$  is the complex conjugate of  $\psi(t)$ .  $\widehat{\psi}(\omega)$  is the complex

SSGST is defined as the Equation (7) according to the theories of synchrosqueezing.

$$GST_{x}(f_{l},b) = L_{f}^{-1}\sum_{f_{k}:|f_{x}(f_{k},b)-f_{l}| \leq \Delta f/2} GST_{x}(f_{k},b) \exp(i2\pi f_{0}f_{k}b)f_{k}^{-1}\Delta f_{k}$$

Where,  $f_1$  is the frequency of the SSGST result, b is the time.

The equation represents that the time-frequency spectra of other time frequency transforms as references, values among the frequency range  $[f_l - L_f, f_l + L_f]$  are superimposed on the frequency point  $f_{l}$ , so that the SSGST has higher accuracy of time-frequency decomposition ability.

The inverse transform of SSGST is given as,

$$\kappa(b) = \operatorname{Re}\left[C_{\psi}^{-1}\sum_{l} SSGST_{x}(f_{l},b)L_{f}\right]$$
(8)

MSE Signal Signal<sub>2</sub> Signal₃

## Synthetic Examples



#### ouble Hyperbolic Chirped Signal



thetic Seismic Signal



be seen, although all energy centers ne true instantaneous frequencies of the ne energy of the CWT and GST results eavily. In contrast, due to the effect of the ng," SSCWT and SSGST squeeze all uency coefficients into the time-frequency which makes the spectra more energyated. In other words, SSCWT and SSGST her time-frequency resolution than CW7 The arrows show that there is no distinct

the SSGST result compared with the result of SSCWT.

We adopted the mean square error (MSE) to validate the reconstruction ability of the SSGST method. As shown in Table, taking the MSE values the SSGST method can reconstruct the original signal well with a lower reconstruction error.

CWT	GST	SSCWT	SSGST
0.0121	3.6979 × 10 <sup>-27</sup>	2.3060 × 10-5	1.8626 × 10-5
0.0181	3.2963 × 10 <sup>-27</sup>	0.0214	0.0083
0.0221	5.0622 × 10 <sup>-26</sup>	7.5347 × 10 <sup>-8</sup>	$1.0925 \times 10^{-4}$

# Field Data Examples

When seismic waves propagate through hydrocarbon reservoirs, waves induced by fluid flow can lead to abnormal attenuation of energy and frequency. The phenomenon of abnormal attenuation mainly shows the loss of high-frequency energy and the conservation of strong low-frequency energy. Spectral decomposition technology can be utilized to identify hydrocarbon reservoirs by analyzing the different frequency response characteristics among different scale geological bodies. Here, we apply the SSGST method to seismic field data from the ZhongJiang Gas Field located in the western Sichuan Basin, China, in order to identify hydrocarbon reservoirs by analyzing the abnormal instantaneous frequency and energy.

In Fig4, the green curve in the horizontal direction at around 1.35ms is a seismic horizon and the green vertical line represents a gas well. The area within the blue ellipse is a gas-bearing reservoir, which is our study area. Fig5 is a histogram based on comprehensive analysis of log data. There are two sets of hydrocarbon reservoirs in the study area: the first set of reservoirs (JS33-1) is located in near a depth of 2560 m and thickness of about 22 m, and the second set of reservoirs (JS33-2) is located in near a depth of 2600 m and thickness of about 18 m.







igure 9. Constant frequency sections based on SSGST (a) at 40 Hz and (b) at 50 Hz. Figure 10. The details of Figure 9a near hydrocarbon reservoir. Although the analysis results of the CWT and GST can all observe the abnormal attenuation in the hydrocarbon reservoir, they offer rough reservoir information due to the poor time-frequency resolution of CWT and GST. Compared with the CWT and GST methods, the SSCWT method can improve the time-frequency resolution and obtain the accurate reservoir information but cannot identify the boundaries of the two sets of hydrocarbon reservoirs, as shown in Figure 8.

we have locally enlarged the 40 Hz constant frequency section in Fig9 and combined it with the red acoustic velocity logging curve to analysis. The strong energy group in the two black-dashed, rectangular areas in Fig10 represent the two hydrocarbon reservoirs, namely JS33-1 and JS33-2, respectively. According to the comparison with CWT, GST, and SSCWT, we can find that SSGST can provide higher time-frequency resolution and can more accurately extract the abnormal response characteristics of seismic signals. Therefore, the SSGST method can locate hydrocarbon reservoirs effectively and depicts the reservoir boundary accurately with higher precision.

Acknowledgment





Thanks to all sponsors of AASPI consortium group for their generous sponsorship.