NEW APPROACH FOR INITIAL DETECTION OF POWER QUALITY DISTURBANCES BASED ON CONTINUOUS S-TRANSFORM(CST)

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ABSTRACT

This paper produces a new approach for power quality analysis using a windowing technique based on Continuous S-transform (CST). This half-cycle window technique approach can detect almost correctly for initial detection of disturbances i.e. voltage sags and swells. S-transform is a time-frequency representation whose analyzing function is the product of a fixed Fourier sinusoid with a scalable, translatable window. S-transform has better time frequency and localization property than traditional and also has ability to detect the disturbance under noisy condition correctly. The excellent time-frequency resolution characteristic of the S-transform makes it the most an attractive candidate for analysis of power system disturbances signals.

Keywords-; power quality disturbances; initial detection; windowing; S-transform

1. INTRODUCTION

Power Quality Disturbances (PQD) issue has become an increased concern for electric utilities and their customers in last decades. By increasing use of solid state switching devices, non linear and power electronically switched loads, unbalanced power systems, lighting controls, computer and data processing equipment, as well as industrial plant rectifiers and inverters is resulting to poor power quality. Disturbances in quality of electric power supply is normally caused by power line disturbances such as voltage sags/swells with or without harmonics, momentary interruption, harmonic distortion, flicker, notch, spike and transients. All this disturbances causing the problems such as malfunctions, short lifetime, instabilities, failure of electrical equipments and so on.

The important issues in power quality analysis are to detect correctly and classify disturbance signals automatically in a efficient manner. Using signal processing technique, various types of PQ disturbances can be detected among others, in time, frequency and time-frequency domains. Several different techniques have been used so far in the literatures to detect power quality disturbance events. The most common technique used for detecting purpose is the calculation of the root mean square (RMS) value of the voltage supply. The main advantage of this technique in terms of calculation, it is simple, fast and much sensitive in sags and swells but not able to detect during transients [8-9]. But, the drawbacks of this technique it is dependence on the size of the sample window. A small window makes the RMS parameter less relevant, as it follows the tendency of the temporal signal, and loses the meaning of mean value of power [9].

Common frequency domain tools that are widely used are the fast Fourier transform (FFT) and the short time Fourier transform (STFT) [1]. This consists of the decomposition of the signal into a sum of sinusoid signals of different frequencies. This analysis can be viewed as a mathematical transformation from the time domain to the frequency domain. FFT is very useful in the analysis of harmonics and is an essential tool for filter design. However, there are some disadvantages such as losses of temporal information, so that it can only be used in the

steady state, and cannot show the moment when the event is produced [12]. STFT has been used in power quality analysis due to its applicability to non-stationary signals. The most advantage of this technique is its ability to give the harmonic content of the signal at every time-period specified by a defined window. But, STFT also has the limitation of fixed window width chosen apriority and this causes limitations for low-frequency and high-frequency non-stationary signals analysis at the same time, the location on the time series may be lost or incorrect [10].

However, using wavelets transform (WT), both time and frequency information of the disturbance can be obtained [13]. The WT on the other hand uses a basis function which dilates and contracts with frequency. It uses short windows at high frequency and long windows at low frequency. Although WT has the capability to extract feature from the signal in both time and frequency domain simultaneously and has been applied in the detection and classification of power quality, it exhibits some disadvantages like excessive computation, sensitivity to noise level and the dependency of its accuracy on the chosen basis wavelet [2].

S-transform (ST) was also introduced recently in [4-6] as an effective technique for PQ disturbances signal processing. It is method for the feature extraction and also detection of PQ disturbances. ST is a extension to the ideas of continuous wavelet transform and is based on a moving and scalable localizing window and has characteristics superior to other transforms. This transform has the ability to detect the disturbance correctly in the presence of noise [2]. The other advantage of S-transform is that it avoids the requirement of testing various families of wavelets so as identify the best one for accurate classification [6]. Further, the decomposition of the disturbance signals at different resolution levels is not required in the S-transform, thereby reducing the memory size and computational over head [3].

This paper proposed a technique based on S-transform for efficient detection of power quality disturbance especially voltage sags and swells. This technique called half-cycle window that applied to the power quality disturbances signal based on continuous S-transform (CST). Each samples of half-cycle window for entire disturbance signal are analyzed based on ST-contour matrices. All the sample windows obtained from the half-cycle technique are

analyzed continuously based on Continuous S-transform. The Continuous S-transform is used to extract the features that can characterize the voltage sags and swells into s-matrices form. So, the significant features from the disturbance signals are continuous extracted by half-cycle windows technique based on continuous S-transform. The most significance contribution of this paper is the new approach technique that applied to detect the voltage sags and swells by an initial detection properly. This approach applied based on Continuous S-transform to get an automatic detection of power quality disturbances type.

2. CONTINUOUS S-TRANSFORM

2.1 S-transform

The S-transform [7] of a time series h(t) is defined as

$$S(\tau, f) = \int_{-\infty}^{\infty} h(t) g(\tau - t, f) e^{-i2\pi ft} dt \qquad (1)$$

Where f is the frequency, τ and t are both time.

The Gaussian modulation function g(τ,f) is given by

$$g(\tau, f) = \frac{|f|}{\sqrt{2\pi}} e^{-(t^2/2\sigma^2)}. \qquad (2)$$

And $\sigma = \frac{1}{|f|}$.

The final expression as follow

$$S(\tau, f) = \int_{-\infty}^{\infty} h(t) \left\{ \frac{|f|}{\sqrt{2\pi}} e^{-((\tau-t)^2 f^2/2)} e^{-i2\pi ft} \right\} dt \qquad (3)$$

The CWT W(τ,d) of a h(t) function is defined as

$$W(\tau, d) = \int_{-\infty}^{\infty} h(t) w(t - \tau, d) dt \qquad (4)$$

The S-transform is obtained by multiplying the CWT with a phase factor as

$$S(\tau, f) = e^{i2\pi f\tau} W(\tau, d) \qquad (5)$$

The final form of the continuous S-transform is obtained as

$$S(\tau, f) = \int_{-\infty}^{\infty} h(t) \left\{ \frac{1}{\sigma(f)\sqrt{2\pi}} e^{-((\tau-t)^2 f^2 / 2\sigma(f)^2)} e^{-i2\pi ft} \right\} dt \quad (6)$$

The width of the Gaussian window is

$$\sigma(f) = T = \frac{1}{|f|} \quad .$$

The inverse S-transform is like

$$h(t) = \int_{-\infty}^{\infty} [\int_{-\infty}^{\infty} S(\tau, f) d\tau] e^{i2\pi ft} df \quad . \quad (7)$$

If additive noise is added to the signal h(t), the operation of the S-transform as

$$S\{h_{noisy}(t)\} = S\{h(t)\} + S\{\eta(t)\} \quad (8)$$

The discrete Fourier transform of the time series h(t) is obtained as

$$H[\frac{n}{NT}] = \frac{1}{N} \sum_{k=0}^{N-1} h(kt) e^{-i2\pi nk} \quad (9)$$

Where n=0,1,…,N-1, (N≥1)

The generalized S-transform of a discrete time series h(t) is derived by letting $\tau \rightarrow jT$ and $f \rightarrow n/NT$ is like

$$S(jT, \frac{n}{NT}) = \sum_{m=0}^{N-1} H[\frac{m+n}{NT}] G(m,n) e^{i2\pi mj/N} \quad . \quad (10)$$

Where $G(m,n) = e^{-2\pi^2 m^2 \alpha^2 / n^2}$, j , m and

$n = 0,1,...,N-1$.

Based on the DFT, the discrete inverse of the S-transform is obtained as

$$h[kt] = \frac{1}{N} \sum_{n=0}^{N-1} \left\{ \sum_{j=0}^{N-1} S\left[\frac{n}{NT}, jT \right] \right\} e^{i2\pi nk} \quad . \quad (11)$$

Where j and $n = 0,1,...,N-1$.

I. DETECTION CAPABILITY OF THE WINDOWING TECHNIQUE BASED CONTINUOUS s-TRANSFORM

The S-transform is having edge over the wavelet transform in detecting a disturbance under a noisy condition. It has the ability to detect the occurrence of disturbance correctly in the presence of noise. The S-transform performs multiresolution analysis on a time varying signal as its window width varies inversely with frequency. This gives high time resolution at high

frequency and high frequency resolution at low frequency [11]. Since power quality disturbances make the power signal a non-stationary one, the S-transform can be applied effectively.

In this paper, the disturbance signals are generated from IEEE13 bus using PSCAD simulation. Two types of power quality disturbances, i.e. voltage sags and swells are generated and the features of all types of disturbances are extracted from the S-contour matrix by using MATLAB programming. From the S-contour matrix, important information in terms of magnitude, frequency, standard deviation and phase can be extracted. To demonstrate the detection capability of this technique based on half-cycle window based on continuous S-transform, two types of disturbances, i.e. voltage sag and voltage swell along with some of the important features are presented in Figs. 2 -11. For simplicity, only two disturbances (sags and swells) are shown here. In Fig.2, represents the original voltage sags signal generated from PSCAD simulation. Fig.3 represents the frequency contour of the S-matrix for the voltage sags signal. Also, in Fig.8, represents the output of new approach in this paper; the initial detection of power quality disturbances for voltage sags. This technique presents the ability of initial detection for the voltage sags correctly. Similarly, Fig. 9-11 represents the above described characteristics for a voltage swells disturbance.
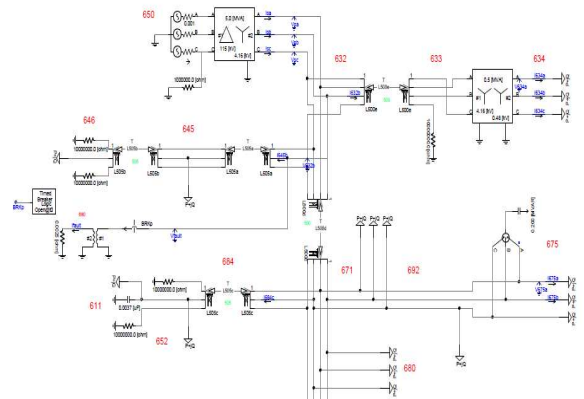

Figure 1.   IEEE 13 bus PSCAD simulation.
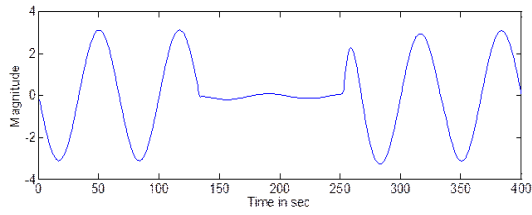
Figure 2.   Voltage sags signal
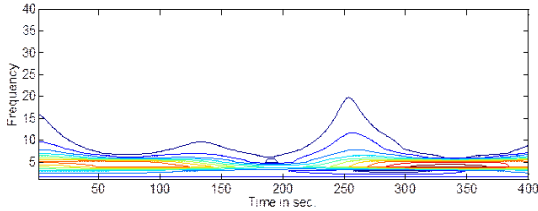

Figure 3.   S-matrix contour of voltage sags.
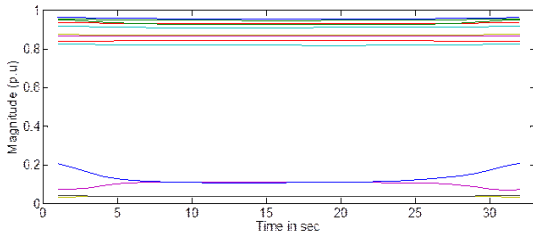

Figure 4.   Standard deviation of real value of samples voltage sags.
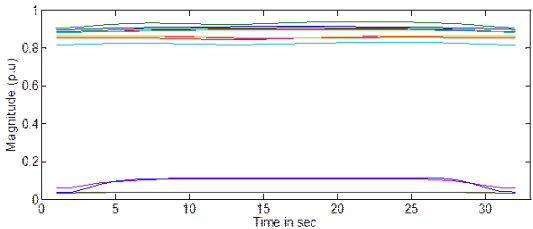

Figure 5.   Standard deviation of absolute value of samples voltage sags.
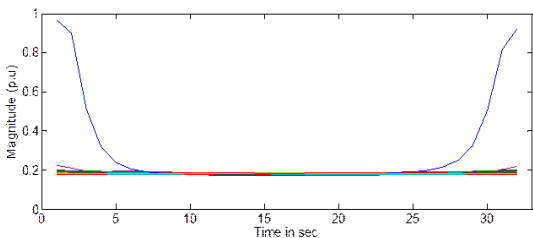

Figure 6.   Absolute value of feature extraction for voltage sags.
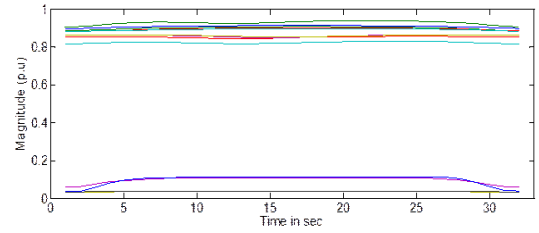

Figure 7.   Feature extraction of voltage sags.
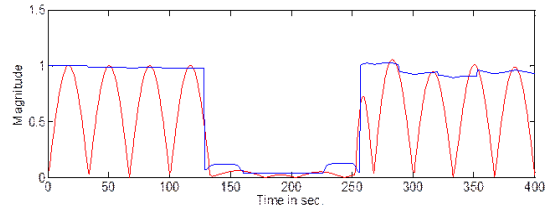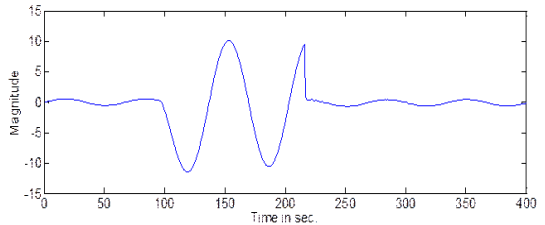

Figure 8.   Initial detection of voltage sags.

Figure 9.   Voltage swells signal.
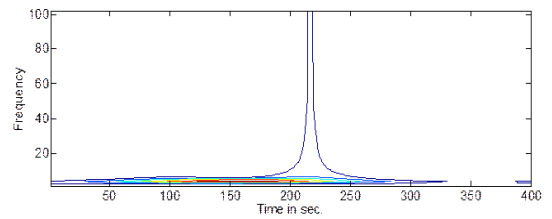

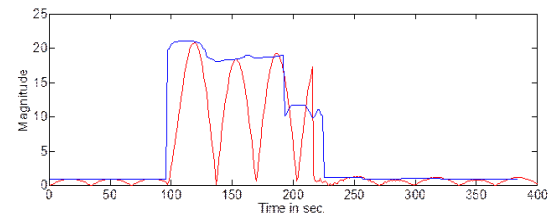Figure 10.   S-matrix contour of swells signal.

Figure 11.   Initial detection of voltage swells.

3. CONCLUSION

In this paper, an attempt has been made to extract efficient feature and detect the PQ disturbances (sags and swells) using half-cycle windowing technique based continuous S-

transform. This paper proposed an improvement technique for initial detection of voltage sags and swells. It is observed that by half-cycle window technique can obtained a correctly detection for PQ disturbances. And the technique could be proved as feasible and effective by more simulation results for another types PQ disturbances. Therefore, the proposed technique can be used as PQ event detection.

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