# High-Order Adaptive Synchrosqueezing Transform 

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# High-Order Adaptive Synchrosqueezing Transform 

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A dissertation submitted to the Graduate School of the University of Missouri-St. Louis
in partial fulfillment of the requirements for the degree Doctor of Philosophy in Mathematical and Computational Sciences with an emphasis in Mathematics.

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#### Abstract

The prevalence of the separation of multicomponent non-stationary signals across many elds of research makes this concept an important subject of study. The synchrosqueezing transform (SST) is a particular type of reassignment method. It aims to separate and recover the components of a multicomponent non-stationary signal. The short time Fourier transform (STFT)-based SST (FSST) and the continuous wavelet transform (CWT)based SST (WSST) have been used in engineering and medical data analysis applications. The current study introduces the dierent versions of FSST and WSST to estimate instantaneous frequency (IF) and to recover components. It has a good concentration and reconstruction for a wide variety of amplitude and frequency modulated multicomponent signals. Earlier studies have improved existing FSSTs by computing more accurate estimates of the IFs of the modes making up the signal. The higher order approximations for both the amplitude and phase were used. Therefore, there is a better concentration and reconstruction for a wider variety of AM-FM modes than what was possible with current synchrosqueezing techniques. In this study, we propose to improve the adaptive FSST, the adaptive WSST, and to introduce a new type of 2nd-order FSST with a new phase transformation. We use higher order approximations for both the amplitude and phase function. We study the higher order adaptive FSST and adaptive WSST. The result shows an even better concentration and reconstruction for a wider variety of AM-FM modes with the higher order adaptive SSTs. We also study the theoretical analysis of the 2nd-order FSST with a new phase transformation. The new phase transformation introduced by us is much simpler than the convectional one, while the performance in IF estimation and component recovery of the new 2nd-order FSST is comparable with, and even better in some cases than, that of the conventional 2nd-order FSST.


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## Chapter 1

## Introduction

Non-stationary signals can be modeled as superpositions of band limited, amplitude and frequency modulated (AM-FM) sub-signals. Non-stationary signals can be shown as,

$$
\begin{equation*}
x(t)=A_{0}(t)+\sum_{k=1}^{K} A_{k}(t) \cos \left(2 \pi \phi_{k}(t)\right) \tag{1.0.1}
\end{equation*}
$$

with $A_{k}(t)$ and $\phi_{k}^{\prime}(t)>0$, where $A_{k}(t)$ is the instantaneous ampllitude (IA), and $\phi_{k}^{\prime}(t)$ is the instantaneous frequenc (IF) of $x_{k}(t)$. The study of separating multicomponent non-stationary signals is a significant research topic in many different fields such as engineering and medical data analysis applications.

In non-stationary signal analysis [1, 2, 3], one of the most important tools is the time-frequency analysis (TFA). The continuous wavelet transform (CWT) and the short time Fourier transform (STFT) are the most common methods in TFA. CWT and STFT are linear time-frequency representations and applicable to component reconstruction. CWT and STFT have suffered because of the uncertainly principle that imposes an unavoidable tradeoff between time and frequency resolutions (see e.g. [3, 4, 5] ).

The synchrosqueezing transform (SST) is a particular type of reassignment method on the CWT and STFT used to sharpen the time-frequency representation of signals and recover the components of a multicomponent. In previous research, the SST has introduced and further developed (see $[6,7])$. Using the CWT-based SST(WSST) [7, 8], Thakur and Wu pro-
posed an extention of the SST given by the STFT-based SST (FSST) [9]. Then, researchers investigated the SST to greater lengths. The FSST and the WSST were also topics of study for Daubechies, Meignen, Wu, Iatsenko $[7,10,11,12,13]$. The 2nd-order SST is proposed in [10, 11]. Likewise, the adaptive FSST, the adaptive 2nd-order FSST with a time-varying, the adaptive WSST, and the adaptive 2nd-order WSST with time-varying have been studied $[14,15]$. Recently, Sheu, Hsu, Chou, and Wu introduced a method to select the time-varying window width for sharp SST representation by minimizing the Renyi entropy [16]. The SST is a useful tool in engineering and medical data application. This includes radar, sonar, anesthesia evaluation, and heartbeat classification. In 2017, Pham and Meignen improved existing STFT-based SSTs by computing more accurate estimates of the instantaneous frequencies (IFs) of the modes making up the signal [17]. They used higher order approximations both for the amplitude and phase. They concluded that there is a better concentration and reconstruction for a wider variety of AM-FM modes than what was possible with current synchrosqueezing techniques. In this study, we used the same technique. We propose to improve the WSST, the adaptive WSST, the adaptive FSST, and the FSST with a new phase transformation by using higher order approximations both for the amplitude and phase. The result also shows an ideal concentration and reconstruction for a wider variety of AM-FM modes.

The theorretical analysis of the 2nd-order FSST was proposed in [18]. Very recently, the theoretical analysis of the FSST obtains the error bounds for the instantaneous frequency (IF) estimation and component recovery with the conventional 2nd-order FSST as was introduced in [16, 19]. We study the theoretical analysis of the 2nd-order FSST with a new phase transformation.

The organization of the remainder of this dissertation is as follows. First, we start with an overview of CWT-based SST and the adaptive CWT-based SST with a time-varying parameter as shown in Chapter 2. The STFT-based SST and the adaptive STFT-based SST with a time-varying parameter are also derived in Chapter 2. In Chapter 3, we propose the higher-order WSS and FSST and the new higher order FSST with the new phase transformation. We address the numerical simulation in Chapter 4. The theoretical analysis of the new formulation of the FSST is described in Chapter 5. Finally, we present the conclusion and future work in Chapter 6.

## Chapter 2

## Preliminaries

In this chapter, some of the basic concepts are presented. The continuous wavelet transform-based synchrosqueezed transform (WSST) are discussed in Section 2.1. WSST with a time varying parameter is studied in Section 2.2. In addition, Section 2.3 is devoted to the short-time Fourier transformbased synchrosqueezed transform (FSST). FSST with a time varying parameter are discussed in Section 2.4. The analysis of FSST and adaptive FSST are provided in Section 2.5 and 2.6.

### 2.1 Continuous wavelet transform-based synchrosqueezed transform (WSST)

A funcation $\psi(t) \in L_{2}(\mathbb{R})$ is called a continuous wavelet if it satisfies (see e.g. $[20,21])$ the following condition

$$
C_{\psi}:=\int_{-\infty}^{\infty}|\hat{\psi}(\xi)|^{2} \frac{d \xi}{|\xi|}<\infty \text { and } \psi_{a, b}(t)=\frac{1}{a} \psi\left(\frac{t-b}{a}\right)
$$

where $\hat{\psi}$ is the Fourier transform of a signal $\psi(t)$ is defined by

$$
\hat{\psi}(\xi)=\int_{-\infty}^{\infty} \psi(t) e^{-i 2 \pi \xi t} d t
$$

Definition 2.1.1. We can define the continuous wavelet transform (CWT) of a signal $x(t) \in L_{2}(\mathbb{R})$ with a continuous wavelet $\psi(t)$ as

$$
\begin{equation*}
W_{x}(a, b)=\int_{-\infty}^{\infty} x(t) \frac{1}{a} \psi \overline{\left(\frac{t-b}{a}\right)} d t=\int_{-\infty}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{2 i \pi b \xi} d \xi \tag{2.1.1}
\end{equation*}
$$

In this instance, $a$ is the scale variable, and $b$ is the time variable.
Note that Fourier transform $\hat{\psi}_{a, b}(\xi)$ of $\psi_{a, b}(t)$ is $\hat{\psi}(a \xi) e^{-i 2 \pi b \xi}$. For more details see [7].
proposition: A function $x(t)$ is an analytic signal if it satisfies $\hat{x}(\xi)=0$ for $\xi<0$. This is defined by

$$
\begin{equation*}
W_{x}(a, b)=\int_{0}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{2 i \pi b \xi} d \xi \tag{2.1.2}
\end{equation*}
$$

Proof. If $a$ and $b$ are two real numbers, then the Fourier transform is

$$
\hat{\psi}_{a, b}(\xi)=\int_{-\infty}^{\infty} \psi_{a, b}(t) e^{-2 i \pi b \xi t} d t=\int_{-\infty}^{\infty} \frac{1}{a} \psi\left(\frac{t-b}{a}\right) e^{-2 i \pi b \xi t} d t
$$

A change of variable $y=\frac{t-b}{a}$, which implies

$$
\begin{aligned}
\hat{\psi}_{a, b}(\xi) & =\int_{-\infty}^{\infty} \psi(y) e^{-2 i \pi(a y+b) \xi} d y \\
& =e^{-2 i \pi b \xi} \int_{-\infty}^{\infty} \psi(y) e^{-2 i \pi a y \xi} d y=e^{-2 i \pi b \xi} \hat{\psi}(a \xi)
\end{aligned}
$$

Then we have

$$
\begin{aligned}
W_{x}(a, b) & =<x, \psi_{a, b}>=<\hat{x}, \hat{\psi}_{a, b}> \\
& =\int_{-\infty}^{\infty} \hat{x}(\xi) \bar{\psi}(a \xi) e^{-2 i \pi b \xi} d \xi \\
& =\int_{-\infty}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{2 i \pi b \xi} d \xi
\end{aligned}
$$

If $x(t)$ is analytic or $\psi(t)$ is analytic, then for $a>0$ and $\xi<0, \hat{\psi}(a \xi)=0$.

$$
\begin{aligned}
W_{x}(a, b) & =\int_{-\infty}^{0} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{2 i \pi b \xi} d \xi+\int_{0}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{2 i \pi b \xi} d \xi \\
& =\int_{0}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{2 i \pi b \xi} d \xi
\end{aligned}
$$

properties(Fourier Transform). Let $x, y \in L_{2}(\mathbb{R})$ are two signals, then we have

$$
\begin{aligned}
\int_{-\infty}^{\infty} x(\xi) \overline{y(\xi)} d \xi & =\int_{-\infty}^{\infty} \hat{x}(\xi) \overline{\hat{y}(\xi)} d \xi \\
\int_{-\infty}^{\infty} \hat{x}(\xi) y(\xi) d \xi & =\int_{-\infty}^{\infty} x(\xi) \hat{y}(\xi) d \xi
\end{aligned}
$$

Example: Let $a$ be a positive real. We define the function $f_{a}(x)=e^{-a x^{2}}$, then we have

$$
\hat{f}_{a}(\xi)=\int_{-\infty}^{\infty} f(t) e^{-2 i \pi \xi t} d t=\int_{-\infty}^{\infty} e^{-2 i \pi \xi t} e^{-a t^{2}} d t
$$

Consider the function

$$
g(y)=\int_{-\infty}^{\infty} e^{-a t^{2}+y t} d t, \text { for } y \in \mathbb{R}
$$

Completing squares, we have

$$
\begin{aligned}
g(y) & =\int_{-\infty}^{\infty} e^{-a\left(t-\frac{y}{2 a}\right)^{2}+\frac{y^{2}}{4 a}} d t=\frac{1}{\sqrt{a}} e^{\frac{y^{2}}{4 a}} \int_{-\infty}^{\infty} e^{-x^{2}} d x \\
& =\frac{1}{\sqrt{a}} e^{\frac{y^{2}}{4 a}} \sqrt{\pi}=\sqrt{\frac{\pi}{a}} e^{\frac{y^{2}}{4 a}}
\end{aligned}
$$

Let $h(y)=\sqrt{\frac{\pi}{a}} e^{\frac{y^{2}}{4 a}}$ be extended to be entire analytic functions, and since they agree on $\mathbb{R}$ as shown, they must agree on the complex plane $\mathbb{C}$. In particular, by setting $y=-2 i \pi \xi$, we have

$$
\begin{align*}
\hat{f}_{a}(\xi) & =\int_{-\infty}^{\infty} e^{-2 i \pi \xi t} e^{-a t^{2}} d t=\sqrt{\frac{\pi}{a}} e^{\frac{(-2 i \pi \xi)^{2}}{4 a}} \\
& =\sqrt{\frac{\pi}{a}} e^{\frac{-4 \pi^{2} \xi^{2}}{4 a}} \tag{2.1.3}
\end{align*}
$$

## Examples of Continuous Wavelet Transform

Bump Wavelet: The Bump wavelet is defined by

$$
\begin{equation*}
\hat{\psi}(\xi)=e^{1-\frac{1}{1-\sigma^{2}(\xi-\mu)^{2}}} \chi_{\left(\mu-\frac{1}{\sigma}, \mu+\frac{1}{\sigma}\right)} \tag{2.1.4}
\end{equation*}
$$

where $\sigma, \mu>0$, such that $\sigma \mu>1$.
Morlet Wavelet: Morlet wavelet $\psi_{\sigma}$ is defined by the function

$$
\begin{equation*}
\psi_{\sigma}(t)=\frac{1}{\sigma \sqrt{2 \pi}} e^{-\frac{t^{2}}{2 \sigma^{2}}}\left(e^{i 2 \pi \mu t}-e^{-2 \pi^{2} \mu^{2} \sigma^{2}}\right) \tag{2.1.5}
\end{equation*}
$$

where $\sigma>0, \mu$ is a constant.
Lemma 2.1.2. Let $\sigma>0$ and $\mu$ two constant, then the Fourier transform of Mortlet wavelet given by

$$
\hat{\psi}_{\sigma}(\xi)=e^{-2 \pi^{2} \sigma^{2}(\xi-\mu)^{2}}-e^{-2 \pi^{2} \sigma^{2}\left(\xi^{2}+\mu^{2}\right)}
$$

Proof. Let $\psi_{\sigma}$ Morlet's complex wavelet, the Fourier transform of $\hat{\psi_{\sigma}}$ is defined by

$$
\begin{aligned}
\hat{\psi}_{\sigma}(\xi) & =\frac{1}{\sigma \sqrt{2 \pi}}\left(\int_{-\infty}^{\infty} e^{-2 i \pi \xi t} e^{-\frac{t^{2}}{2 \sigma^{2}}} e^{2 i \pi \mu t} d t-e^{-2 \pi^{2} \mu^{2} \sigma^{2}} \int_{-\infty}^{\infty} e^{-2 i \pi \xi t} e^{-\frac{t^{2}}{2 \sigma^{2}}} d t\right) \\
& =\frac{1}{\sigma \sqrt{2 \pi}}\left(\int_{-\infty}^{\infty} e^{-2 i \pi(\xi-\mu)} e^{-\frac{t^{2}}{2 \sigma^{2}}} d t-e^{-2 \pi^{2} \mu^{2} \sigma^{2}} \int_{-\infty}^{\infty} e^{-2 i \pi \xi t} e^{-\frac{t^{2}}{2 \sigma^{2}}} d t\right)
\end{aligned}
$$

According to the example of Fourier transform (2.1.3)

$$
\begin{aligned}
\hat{\psi}_{\sigma}(\xi) & =\frac{1}{\sigma \sqrt{2 \pi}}\left(\hat{f}_{\frac{1}{2 \sigma^{2}}}(\xi-\mu)-e^{-2 \pi^{2} \mu^{2} \sigma^{2}} \hat{f}_{\frac{1}{2} \sigma^{2}}(\xi)\right) \\
& =\frac{1}{\sigma \sqrt{2 \pi}}\left(\sqrt{\pi 2 \sigma^{2}} e^{-2 \pi^{2} \sigma^{2}(\xi-\mu)}-e^{-2 \pi^{2} \mu^{2} \sigma^{2}} \sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \xi^{2}}\right) \\
& =\left(e^{-2 \pi^{2} \sigma^{2}(\xi-\mu)}-e^{-2 \pi^{2} \sigma^{2}\left(\xi^{2}+\mu^{2}\right)}\right)
\end{aligned}
$$

Maxican hat Wavelet:
Let $\sigma>0$, then the Maxican hat wavelet is defined by

$$
\begin{equation*}
\psi_{\sigma}(t)=\left(1-\frac{t^{2}}{\sigma^{2}}\right) e^{-\frac{t^{2}}{2 \sigma^{2}}} \tag{2.1.6}
\end{equation*}
$$

Lemma 2.1.3. The Fourier Transform of Maxican hat wavelet is given by

$$
\hat{\psi}_{\sigma}(\xi)= \begin{cases}(2 \pi \sigma \xi)^{2} \sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}} & \text { for } \xi>0  \tag{2.1.7}\\ 0 & \text { for } \xi \leq 0\end{cases}
$$

Proof. The Fourier Transform of Maxican hat wavelet function $\psi_{\sigma}$

$$
\begin{aligned}
\hat{\psi}_{\sigma}(\xi) & =\int_{-\infty}^{\infty}\left(1-\frac{t^{2}}{\sigma^{2}}\right) e^{-\frac{t^{2}}{2 \sigma^{2}}} e^{-2 i \pi t \xi} d t \\
& =\underbrace{\int_{-\infty}^{\infty} e^{-\frac{t^{2}}{2 \sigma^{2}}} e^{-2 i \pi t \xi} d t}_{I_{1}}+\underbrace{\int_{-\infty}^{\infty} \frac{t^{2}}{\sigma^{2}} e^{-\frac{t^{2}}{2 \sigma^{2}}} e^{-2 i \pi t \xi} d t}_{I_{2}}
\end{aligned}
$$

Calculation of the integral $I_{1}=\mathcal{F}\left(f_{\frac{1}{2 \sigma^{2}}}\right)(\xi)=\sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}$.
Calculation of the integral $I_{2}=\mathcal{F}\left(t^{2} e^{-\frac{t^{2}}{2 \sigma^{2}}}\right)(\xi)$. Using the Fourier transform derivation formula $\mathcal{F}\left((-2 i \pi t)^{m} f(t)\right)(\xi)=(\mathcal{F}(f))^{(m)}(\xi)$

$$
\begin{aligned}
\mathcal{F}\left((-2 i \pi t)^{2} f_{\frac{1}{2 \sigma^{2}}}(t)\right)(\xi) & =\left(\mathcal{F}\left(f_{\frac{1}{2 \sigma}}\right)(\xi)\right)^{(2)}=\frac{d^{2}}{d^{2} \xi}\left(\sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}\right) \\
& =\frac{d}{d \xi}\left(\sqrt{2 \pi \sigma^{2}}\left(-4 \pi^{2} \sigma^{2} \xi\right) e^{-2 \pi^{2} \sigma^{2} \xi^{2}}\right) \\
& =-4 \pi^{-2} \sigma^{2} \sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi}+4^{2} \pi^{4} \sigma^{4} \xi^{2} \sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}
\end{aligned}
$$

This result in

$$
\begin{aligned}
\mathcal{F}\left(\frac{t^{2}}{\sigma^{2}} f_{\frac{1}{2 \sigma}}(t)\right)(\xi) & =\frac{1}{-4 \pi^{2} \sigma^{2}}\left(\sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}\right)\left(-4 \pi^{2} \sigma^{2}+4^{2} \pi^{4} \sigma^{4} \xi^{2}\right) \\
& =\sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}\left(1-\pi^{2} \sigma^{2} \xi^{2}\right)
\end{aligned}
$$

We group the integrals calculation above

$$
\begin{aligned}
\hat{\psi}(\xi) & =\sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}-\sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}\left(1-4 \pi^{2} \sigma^{2} \xi^{2}\right) \\
& =\sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}\left(1-1^{2}+4 \pi^{2} \sigma^{2} \xi^{2}\right) \\
& =(2 \pi \sigma \xi)^{2} \sqrt{2 \pi \sigma^{2}} e^{-2 \pi^{2} \sigma^{2} \xi^{2}}
\end{aligned}
$$

## Inverse Continuous Wavelet Transform

Theorem 2.1.4. (Inverse CWT)
Let $x(t) \in L_{2}(\mathbb{R})$ and $\psi(t)$ is continuous wavelet, the inverse wavelet transform can be recover the signal $x(t)$ (see e.g. [20, 22, 23, 24])

$$
\begin{equation*}
x(t)=\frac{1}{C_{\psi}} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} W_{x}(a, b) \psi_{a, b}(t) d b \frac{d a}{|a|} \tag{2.1.8}
\end{equation*}
$$

where $C_{\psi}=\int_{-\infty}^{\infty}|\hat{\psi}(\xi)|^{2} \frac{d \xi}{|\xi|}<\infty$
If $x(t)$ is analytic or $\psi(t)$ is real, then:

$$
\begin{equation*}
x(t)=\frac{1}{\tilde{C}_{\psi}} \int_{0}^{\infty} \int_{-\infty}^{\infty} W_{x}(a, b) \psi_{a, b}(t) d b \frac{d a}{a} \tag{2.1.9}
\end{equation*}
$$

where $\tilde{C}_{\psi}=\int_{0}^{\infty}|\hat{\psi}(\xi)|^{2} \frac{d \xi}{|\xi|}<\infty$.

Proof. If $x(t) \in L_{2}(\mathbb{R})$ and $\psi(t)$ a continuous wavelet, then we have

$$
\int_{0}^{\infty} \int_{-\infty}^{\infty} W_{x}(a, b) \psi_{a, b}(t) d b \frac{d a}{a}=\int_{0}^{\infty} \int_{-\infty}^{\infty} \int_{0}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{i 2 \pi b \xi} d \xi \psi_{a, b}(t) d b \frac{d a}{a}
$$

$$
\begin{align*}
& =\int_{0}^{\infty} \int_{0}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} \int_{-\infty}^{\infty} \psi_{a, b} e^{i 2 \pi b \xi}(t) d b \frac{d a}{a} d \xi \\
& =\int_{0}^{\infty} \int_{0}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} \hat{\psi}(a \xi) e^{i 2 \pi t \xi} d b \frac{d a}{a} d \xi  \tag{2.1.10}\\
& =\int_{0}^{\infty} \hat{x}(\xi) e^{i 2 \pi t \xi} \int_{0}^{\infty}|\hat{\psi}(a \xi)|^{2} d b \frac{d a}{a} d \xi \\
& =\tilde{C}_{\psi} \int_{0}^{\infty} \hat{x}(\xi) e^{i 2 \pi t \xi} d \xi=\tilde{C}_{\psi} x(t)
\end{align*}
$$

note we assume that a continuous wavelet $\psi(t)$ also satisfies

$$
\begin{equation*}
0 \neq c_{\psi}=\int_{0}^{\infty} \overline{\hat{\psi}(\xi)} \frac{d \xi}{\xi}<\infty \tag{2.1.11}
\end{equation*}
$$

Theorem 2.1.5. (Inverse CWT involving a for analytic signal ) Suppose $x(t) \in L_{2}(\mathbb{R})$ and $\psi(t)$ is a continuous wavelet. If $x(t)$ is analytic, then

$$
\begin{equation*}
x(b)=\frac{1}{c_{\psi}} \int_{0}^{\infty} W_{x}(a, b) \frac{d a}{a} \tag{2.1.12}
\end{equation*}
$$

where $C_{\psi}$ is defined $0 \neq C_{\psi}=\int_{0}^{\infty} \overline{\hat{\psi}(\xi)} \frac{d \xi}{\xi}<\infty$.
Proof. From (2.1.2), we have

$$
\begin{aligned}
\int_{0}^{\infty} W_{x}(a, b) \frac{d a}{a} & =\int_{0}^{\infty} \int_{0}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{2 i \pi b \xi} d \xi \frac{d a}{a} \\
& =\int_{0}^{\infty} \hat{x}(\xi) e^{2 i \pi b \xi} \int_{0}^{\infty} \overline{\hat{\psi}(a \xi)} \frac{d a}{a} d \xi \\
& =\int_{0}^{\infty} \hat{x}(\xi) e^{2 i \pi b \xi} d \xi \int_{0}^{\infty} \overline{\hat{\psi}(a)} \frac{d a}{a} \\
& =c_{\psi} \int_{0}^{\infty} \hat{x}(\xi) e^{2 i \pi b \xi} d \xi=c_{\psi} x(b)
\end{aligned}
$$

Furthermore, a real signal $x(t) \in L_{2}(\mathbb{R})$ can be recovered from its CWT with an analytic continuous wavelet by the following formula which does not involve the time variable $b$ either.

Theorem 2.1.6. (Inverse CWT involving a for a real signal (refer to [7])) Suppose $x(t) \in L_{2}(\mathbb{R})$ and $\psi(t)$ is a continuous wavelet. If $x(t)$ is real and $\psi$ is analytic, then

$$
\begin{equation*}
x(b)=\operatorname{Re}\left(\frac{2}{c_{\psi}} \int_{0}^{\infty} W_{x}(a, b) \frac{d a}{a}\right) \tag{2.1.13}
\end{equation*}
$$

where $C_{\psi}$ is defined $0 \neq C_{\psi}=\int_{0}^{\infty} \overline{\hat{\psi}(\xi)} \frac{d \xi}{\xi}<\infty$.
Proof. When $x(t)$ is real, we have $\overline{\hat{x}(\xi)}=\hat{x}(-\xi)$, thus,

$$
\int_{-\infty}^{0} \hat{x}(\xi) e^{i 2 \pi b \xi} d \xi=\int_{0}^{\infty} \hat{x}(-\xi) e^{-i 2 \pi b \xi} d \xi=\overline{\int_{0}^{\infty} \hat{x}(\xi) e^{i 2 \pi b \xi}} d \xi
$$

and hence

$$
\begin{align*}
x(b) & =\int_{-\infty}^{\infty} \hat{x}(\xi) e^{i 2 \pi b \xi} d \xi=\int_{-\infty}^{0} \hat{x}(\xi) e^{i 2 \pi b \xi} d \xi+\int_{0}^{\infty} \hat{x}(\xi) e^{i 2 \pi b \xi} d \xi  \tag{2.1.14}\\
& =2 \operatorname{Re}\left(\int_{0}^{\infty} \hat{x}(\xi) e^{i 2 \pi b \xi} d \xi\right)
\end{align*}
$$

From (2.1.1) and the proof of theorem 2.1.4, we have

$$
\int_{0}^{\infty} W_{x}(a, b) \frac{d a}{a}=c_{\psi} \int_{0}^{\infty} \hat{x}(\xi) e^{i 2 \pi b \xi} d \xi
$$

Therefore,

$$
\begin{equation*}
x(b)=2 \operatorname{Re}\left(\int_{0}^{\infty} \hat{x}(\xi) e^{i 2 \pi b \xi} d \xi\right)=\operatorname{Re}\left(\frac{2}{c_{\psi}} \int_{0}^{\infty} W_{x}(a, b) \frac{d a}{a}\right) \tag{2.1.15}
\end{equation*}
$$

### 2.1.1 1st-order WSST

The 1st-order WSST was introuced in [14]. The WSST is designed to reassign the scale variable $a$ to the frequency variable. Let's start with the most classic example. The CWT of $x(t)=A \cos (2 \pi c t)$ where $c$ is a positive constant. Then the Fourier transform of $x$ is defined by

$$
\hat{x}(\xi)=A \int_{-\infty}^{\infty} \cos (2 \pi c t) e^{-2 i \pi \xi t} d t
$$

We use the fact that $\cos (2 \pi c t)=\frac{e^{2 i \pi c t}+e^{-2 i \pi c t}}{2}$

$$
\begin{align*}
\hat{x}(\xi) & =\frac{A}{2}\left(\int_{-\infty}^{\infty} e^{2 i \pi c t} e^{-2 i \pi \xi t} d t+\int_{-\infty}^{\infty} e^{-2 i \pi c t} e^{-2 i \pi \xi t} d t\right)  \tag{2.1.1}\\
& =\frac{A}{2}\left(e^{-2 i t \pi(\xi-c)}+e^{-2 i t \pi(\xi+c)}\right)=\frac{A}{2}\left(\delta_{c}(\xi)+\delta_{-c}(\xi)\right)
\end{align*}
$$

Thus for $a>0$

$$
W_{x}(a, b)=\int_{-\infty}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}(a \xi)} e^{2 i \pi b \xi} d \xi=\frac{1}{2} A \overline{\hat{\psi}(a c)} e^{2 i \pi b c}
$$

The instantaneous frequency (IF) of $x(t)$ is represented as $c$

$$
\frac{\partial}{\partial b} W_{x}(a, b)=\frac{1}{2} A \hat{\psi}(a c) e^{i 2 \pi b c}=2 i \pi c W_{x}(a, b)
$$

This implies that

$$
\frac{\frac{\partial}{\partial b} W_{x}(a, b)}{2 i \pi W_{x}(a, b)}=c
$$

For a general $x(t)$, at $(a, b)$ for which $W_{x}(a, b) \neq 0$, a good approximation for the phase transformation (also called theinstantaneous frequency (IF)) of $x$ is $\omega_{x}^{1 s t}$

$$
\omega_{x}^{1 s t}(a, b)=\frac{\frac{\partial}{\partial b} W_{x}(a, b)}{2 i \pi W_{x}(a, b)}, \quad \text { for } \quad W_{x}(a, b) \neq 0
$$

The 1st-order WSST of a signal $x(t)$ is defined by

$$
\begin{equation*}
T_{x}^{1 s t}(\xi, b)=\int_{a \in \mathbb{R}_{+}: W_{x}(a, b) \neq 0} W_{x}(a, b) \delta\left(\omega_{x}^{1 s t}(a, b)-\xi\right) \frac{d a}{a} \tag{2.1.2}
\end{equation*}
$$

where $\xi$ is the frequency variable.
The input signal $x(t)$ can be recovered from its WSST. For the analytic $x(t) \in L_{2}(\mathbb{R})$, we have

$$
x(b)=\frac{1}{c_{\psi}} \int_{0}^{\infty} T_{x}^{1 s t}(\xi, b) d \xi ;
$$

and for a real-valued $x(t) \in L_{2}(\mathbb{R})$

$$
x(b)=\operatorname{Re}\left(\frac{2}{c_{\psi}} \int_{0}^{\infty} T_{x}^{1 s t}(\xi, b) d \xi\right),
$$

where $C_{\psi}$ is defined $0 \neq C_{\psi}=\int_{0}^{\infty} \overline{\hat{\psi}(\xi)} \frac{d \xi}{\xi}<\infty$.
For a multicomponent signal $x(t)=A_{0}(t)+\sum_{k=1}^{K} x_{k}(t), x_{k}(t)=A_{k}(t) \cos \left(r \pi \phi_{k}(t)\right)$ with $A_{0}(t)=0$, each component $x_{k}(b)$ can be recovered from WSST:

$$
x(b) \approx \operatorname{Re}\left(\frac{2}{c_{\psi}} \int_{\left|\xi-\phi_{k}^{\prime}(b)\right|<\Gamma} T_{x}^{1 s t}(\xi, b) d \xi\right),
$$

for certain $\Gamma>0$.

### 2.1.2 2nd-order WSST

The 2nd-order WSST was proposed in [14]. This defines a new phase transformation more precise than that of 1st-order $\omega_{x}^{2 n d}$. This 1st-order is associated with the 2nd order partial derivatives of the CWT of $x(t)$; when $x(t)$ is a linear frequency modulation signal (linear chirp), then $\omega_{x}^{2 n d}$ is exactly the IF of $x(t)$. Therefore, $x(t)$ is a linear frequency modulation (LFM) signal if

$$
x(t)=A(t) e^{i 2 \pi \phi(t)}=A(t) e^{p t+\frac{q}{2} t^{2}} e^{i 2 \pi\left(c t+\frac{1}{2} r t^{2}\right)}
$$

with phase function $\phi(t)=c t+\frac{1}{2} r t^{2}$, IF $\phi^{\prime}(t)=c+r t$ and chirp rate $\phi^{\prime \prime}(t)=r$, IA $A(t)=A e^{p t+\frac{q}{2} t^{2}}$. In the following we derive the phase transformation $\omega_{x}^{2 n d}$. For a given wavelet $\psi, W_{x}(b, a)$ is the CWT of $x(t)$ with $\psi$ defined by (2.1.1). For $\psi_{1}(t)=t \psi(t), W_{x}^{\psi_{1}}(b, a)$ denotes the CWT of $x(t)$ with $\psi_{1}(t)$ and represents the integral on the right-hand side of (2.1.1) with $\psi(t)$ replaced by $\psi_{1}(t)$. The derivative of the signal $x$ is given by

$$
\begin{aligned}
& x^{\prime}(t)=(p+q t+i 2 \pi(c+r t)) x(t) \\
& W_{x}(b, a)=\int_{-\infty}^{\infty} x(b+a t) \overline{\psi(t)} d t .
\end{aligned}
$$

The derivative of the CWT with respect to the variable $b$ is

$$
\begin{align*}
\left.\partial_{b} W_{x}^{( } b, a\right) & =\int_{-\infty}^{\infty} x^{\prime}(b+a t) \overline{\psi(t)} d t \\
& \left.=\int_{-\infty}^{\infty}(p+q(b+a t)+i 2 \pi(c+r b+r a t)) x(b+a t)\right) \overline{\psi(t)} d t \\
& =(p+q b+i 2 \pi(c+r b)) W_{x}(b, a)+(q+i 2 \pi r) a W_{x}^{\psi_{1}}(b, a) .
\end{align*}
$$

At $(a, b)$, on which $W_{x}(a, b) \neq 0$, we then denote that

$$
\frac{\partial_{b} W_{x}(b, a)}{W_{x}(b, a)}=p+q b+i 2 \pi(c+r b)+(q+i 2 \pi r) a \frac{W^{\psi_{1}}(b, a)}{W_{x}(b, a)}
$$

$$
\frac{\partial}{\partial a}\left(\frac{\partial_{b} W_{x}(b, a)}{W_{x}(b, a)}\right)=(q+i 2 \pi r) U(a, b)
$$

where

$$
U(a, b)=\frac{\partial}{\partial a}\left(\frac{a W_{x}^{\psi_{1}}(b, a)}{W_{x}(b, a)}\right)=\frac{W_{x}^{\psi_{1}}(b, a)}{W_{x}(b, a)}+a \frac{\partial}{\partial a}\left(\frac{W_{x}^{\psi_{1}}(b, a)}{W_{x}(b, a)}\right) .
$$

If $U(a, b) \neq 0$, then

$$
q+i 2 \pi r=\frac{1}{U(a, b)} \frac{\partial}{\partial a}\left(\frac{\partial_{b} W_{x}(b, a)}{W_{x}(b, a)}\right) .
$$

Then, we have

$$
\frac{\partial_{b} W_{x}(b, a)}{W_{x}(b, a)}=p+q b+i 2 \pi(c+r b)+a \frac{W_{x}^{\psi_{1}}(b, a)}{W_{x}(b, a) U(a, b)} \frac{\partial}{\partial a}\left(\frac{\partial_{b} W_{x}(b, a)}{W_{x}(b, a)}\right)
$$

Thus,

$$
\phi^{\prime}(b)=c+r b=\operatorname{Re}\left\{\frac{\partial_{b} W_{x}(b, a)}{i 2 \pi W_{x}(b, a)}\right\}-a \operatorname{Re}\left\{\frac{W_{x}^{\psi_{1}}(b, a)}{W_{x}(b, a) U(a, b)} \frac{\partial}{\partial a}\left(\frac{\partial_{b} W_{x}(b, a)}{i 2 \pi W_{x}(b, a)}\right)\right\} .
$$

Therefore, we define the second-order phase transformation $\omega_{x}^{2 n d}$ by the formula

$$
\omega_{x}^{2 n d}(a, b)= \begin{cases}\operatorname{Re}\left\{\frac{\partial_{b} W_{x}(b, a)}{i 2 \pi W_{x}(b, a)}\right\}-a \operatorname{Re}\left\{\frac{W^{\psi_{1}}(b, a)}{W_{x}(b, a) U(a, b)} \frac{\partial}{\partial a}\left(\frac{\partial_{b} W(b, a)}{i 2 \pi W_{x}(b, a)}\right)\right\}, & \text { if } \quad U(a, b) \neq 0, W_{x}(b, a) \neq 0 \\ \operatorname{Re}\left\{\frac{\partial_{b} W_{x}(b, a)}{i 2 \pi W_{x}(b, a)}\right\}, & \text { if } \quad U(a, b)=0, W_{x}(b, a) \neq 0\end{cases}
$$

We define the 2nd-order WSST of a signal $x(t)$ as

$$
T_{x}^{2 n d}(\xi, b)=\int_{a \in \mathbb{R}_{+}: W_{x} \neq 0} W_{x}(a, b) \delta\left(\omega_{x}^{2 n d}(a, b)-\xi\right) \frac{d a}{a}
$$

where $\xi$ is the frequency variable.
To recover the input signal $x(t)$ from its WSST for analytic $x(t) \in L_{2}(\mathbb{R})$, we say that

$$
x(b)=\frac{1}{C_{\psi}(b)} \int_{0}^{\infty} T_{x}^{2 n d}(\xi, b) d \xi
$$

and for a real valued $x(t) \in L_{2}(\mathbb{R})$,

$$
x(b)=\operatorname{Re}\left(\frac{2}{C_{\psi}(b)} \int_{0}^{\infty} T_{x}^{2 n d}(\xi, b) d \xi\right)
$$

where $c_{\psi}(b)$ is defined by

$$
0 \neq C_{\psi}=\int_{0}^{\infty} \overline{\hat{\psi}(\xi)} \frac{d \xi}{\xi}<\infty
$$

For a multicomponent signal $x(t)$, we can recover each component $x_{k}(b)$ from the WSST:

$$
x_{k}(b)=\operatorname{Re}\left(\frac{2}{C_{\psi}(b)} \int_{\left|\xi-\phi^{\prime}(b)\right|<\Gamma} T_{x}^{2 n d}(\xi, b) d \xi\right)
$$

for certain $\Gamma>0$.

### 2.2 WSST with a time varying parameter

Continuous wavelets $\psi_{\sigma}$ are dependent on the function $\sigma$ defined by

$$
\psi_{\sigma}(t)=\frac{1}{\sigma} g\left(\frac{t}{\sigma}\right) e^{i 2 \pi \mu t}
$$

The CWT of a signal $x(t)$ with a time-varying parameter is defined by

$$
\widetilde{W}_{x}(a, b)=\int_{-\infty}^{\infty} x(t) \frac{1}{a} \overline{\psi_{\sigma(b)}\left(\frac{t-b}{a}\right)} d t=\int_{-\infty}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}_{\sigma(b)}(a \xi)} e^{i 2 \pi b \xi} d \xi
$$

We call $\widetilde{W}_{x}(a, b)$ the adaptive CWT of $x(t)$ with $\psi_{\sigma}$. If $\psi_{\sigma}$ is an analytic wavelet or $x(t)$ is analytic, then we have $a>0$

$$
\widetilde{W}_{x}(a, b)=\int_{0}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}_{\sigma(b)}(a \xi)} e^{i 2 \pi b \xi} d \xi
$$

An analytic signal $x(t)$ can be recovered back from its CWT

$$
x(b)=\frac{1}{c_{\psi(b)}} \int_{0}^{\infty} \widetilde{W}_{x}(a, b) \frac{d a}{a}
$$

where $C_{\psi}(b)$ is defined by

$$
C_{\psi}(b)=\int_{0}^{\infty} \overline{\hat{\psi}_{\sigma(b)}(\xi)} \frac{d \xi}{\xi}=\int_{0}^{\infty} \overline{\hat{g}(\xi-\sigma(b) \mu)} \frac{d \xi}{\xi}
$$

If $x(t)$ is a real signal, then we have

$$
x(b)=\operatorname{Re}\left(\frac{2}{C_{\psi}(b)} \int_{0}^{\infty} \widetilde{W}_{x}(a, b) \frac{d a}{a}\right) .
$$

### 2.2.1 Adaptive 1st-order WSST

The adaptive 1st-order WSST was introduced in [14]. In order to define the adaptive CWT, we must first define the $\omega_{x}^{1 s t, a d p}$. Let $\psi_{\sigma}(t)$ and $\psi_{\sigma}^{2}(t)=$ $\frac{t}{\sigma^{2}} g^{\prime}\left(\frac{t}{\sigma}\right) e^{i 2 \pi \mu t}$ be the continuous wavelet. Denote $\widetilde{W}_{x}^{\psi^{2}}(a, b)$ defined by

$$
\widetilde{W}_{x}^{\psi^{2}}(a, b)=\int_{-\infty}^{\infty} x(t) \frac{1}{a} \overline{\psi_{\sigma(b)}^{2}\left(\frac{t-b}{a}\right)} d t=\int_{-\infty}^{\infty} x(b+a t) \frac{t}{\sigma^{2}(b)} \overline{g^{\prime}\left(\frac{t}{\sigma}\right)} e^{-i 2 \pi \mu t} d t
$$

One can obtain that

$$
\hat{\psi}_{\sigma}^{2}(\xi)=-\hat{g}(\sigma(\xi-\mu))-\sigma(\xi-\mu)(\hat{g})^{\prime}(\sigma(\xi-\mu))
$$

To define the phase transformation $\omega_{x}^{a d p}(a, b)$, we consider $x(t)=A e^{i 2 \pi c t}$. From

$$
\widetilde{W}_{x}(a, b)=\int_{-\infty}^{\infty} x(b+a t) \overline{\psi_{\sigma(b)}(t)} d t=A \int_{-\infty}^{\infty} e^{i 2 \pi c(b+a t)} \frac{1}{\sigma(b)} g \overline{\frac{t}{\sigma(b)}} e^{-i 2 \pi \mu t} d t
$$

We take the partial derivative $\frac{\partial}{\partial b}$ to both sides, and we have

$$
\begin{aligned}
\partial_{b} \widetilde{W}_{x}(a, b) & \left.=A \int_{-\infty}^{\infty}(i 2 \pi c) e^{i 2 \pi c(b+a t)} \frac{1}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right.}\right) e^{-i 2 \pi \mu t} d t \\
& +A \int_{-\infty}^{\infty} e^{i 2 \pi c(b+a t)}\left(\frac{-\sigma^{\prime}(b)}{\sigma(b)^{2}}\right) g\left(\frac{t}{\sigma(b)}\right) e^{-i 2 \pi \mu t} d t \\
& +A \int_{-\infty}^{\infty} e^{i 2 \pi c(b+a t)}\left(\frac{-\sigma^{\prime}(b)}{\sigma(b)^{3}}\right) g^{\prime}\left(\frac{t}{\sigma(b)}\right) e^{-i 2 \pi \mu t} d t \\
& =i 2 \pi c \widetilde{W}_{x}(a, b)-\frac{\sigma^{\prime}(b)}{\sigma(b)} \widetilde{W}_{x}(a, b)-\frac{\sigma^{\prime}(b)}{\sigma(b)} \widetilde{W}_{x}^{\psi^{2}}(a, b) .
\end{aligned}
$$

If $\widetilde{W}_{x}(a, b) \neq 0$, then

$$
\frac{\frac{\partial}{\partial b} \widetilde{W}_{x}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}=c-\frac{\sigma^{\prime}(b)}{i 2 \pi \sigma(b)}-\frac{\sigma^{\prime}(b)}{\sigma(b)} \frac{\widetilde{W}_{x}^{\psi^{2}}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}
$$

Thus, the instantaneous frequency (IF) $c$ of $x(t)$ can be denoted by

$$
c=\operatorname{Re}\left\{\frac{\frac{\partial}{\partial b} \widetilde{W}_{x}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}\right\}+\frac{\sigma^{\prime}(b)}{\sigma(b)} \operatorname{Re}\left\{\frac{\widetilde{W}_{x}^{\psi^{2}}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}\right\}
$$

For a general signal $x(t)$, at $(a, b)$ for which $\widetilde{W}_{x}(a, b) \neq 0$, denote

$$
\widetilde{\omega}_{x}^{1 s t, a d p}(a, b)=\operatorname{Re}\left\{\frac{\frac{\partial}{\partial b} \widetilde{W}_{x}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}\right\}+\frac{\sigma^{\prime}(b)}{\sigma(b)} \operatorname{Re}\left\{\frac{\widetilde{W}_{x}^{\psi^{2}}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}\right\} .
$$

The quantity $\omega_{x}^{1 s t, a d p}(a, b)$ is called the "phase transformation." The WSST with a time-varying parameter (also called the adaptive WSST) of $x(t)$ is defined by

$$
\begin{equation*}
T_{x}^{1 s t, a d p}(\xi, b)=\int_{\left\{a \in \mathbb{R}_{+}: \widetilde{W}_{x}(a, b) \neq 0\right\}} \widetilde{W}_{x}(a, b) \delta\left(\widetilde{\omega}_{x}^{1 s t, a d p}(a, b)-\xi\right) \frac{d a}{a} \tag{2.2.1}
\end{equation*}
$$

where $\xi$ is the frequency variable.
The input signal $x(t)$ can be recovered from its adaptive WSST for the analytic $x(t) \in L_{2}(\mathbb{R})$. We have

$$
x(b)=\frac{1}{C_{\psi}(b)} \int_{0}^{\infty} T_{x}^{1 s t, a d p}(\xi, b) d \xi
$$

and for a real valued $x(t) \in L_{2}(\mathbb{R})$

$$
x(b)=\operatorname{Re}\left(\frac{2}{C_{\psi}(b)} \int_{0}^{\infty} T_{x}^{1 s t, a d p}(\xi, b) d \xi\right)
$$

where $c_{\psi}(b)$ is defined by

$$
C_{\psi}(b)=\int_{0}^{\infty} \overline{\hat{\psi}_{\sigma(b)}(\xi)} \frac{d \xi}{\xi}=\int_{0}^{\infty} \overline{\hat{g}(\xi-\sigma(b) \mu)} \frac{d \xi}{\xi} .
$$

The following formula can be used to recover the $k$-th component $x_{k}(b)$ of a multicomponent signal from the adaptive WSST

$$
x_{k}(b)=\operatorname{Re}\left(\frac{2}{C_{\psi}(b)} \int_{\left|\xi-\phi^{\prime}(b)\right|<\Gamma_{1}} T_{x}^{1 s t, a d p}(\xi, b) d \xi\right)
$$

for certain $\Gamma_{1}>0$.
Remark: Here, we may say that if $\psi_{\sigma}$ is the simplified version of Morlet's wavelet given by

$$
\hat{\psi}_{\sigma}(\xi)=e^{-2 \pi^{2} \sigma^{2}(\xi-\mu)^{2}} \text { and } \hat{\psi}_{\sigma}^{2}(\xi)=\left(4 \pi^{2} \sigma^{2}(\xi-\mu)^{2}-1\right) e^{-2 \pi^{2} \sigma^{2}(\xi-\mu)^{2}}
$$

Next, we calculate:

$$
\begin{aligned}
\widetilde{W}_{x}^{\psi^{2}}(a, b) & =\int_{-\infty}^{\infty} \hat{x}(\xi) \overline{\hat{\psi}_{\sigma(b)}^{2}(a \xi)} e^{i 2 \pi b \xi} d \xi=A \hat{\psi}_{\sigma(b)}^{2}(a c) e^{i 2 \pi b c} \\
& =A\left(4 \pi^{2} \sigma^{2}(b)(a c-\mu)^{2}-1\right) e^{i 2 \pi^{2} \sigma^{2}(b)(\xi-\mu)^{2}} \\
& =\left(4 \pi^{2} \sigma^{2}(b)(a c-\mu)^{2}-1\right) \widetilde{W}_{x}(a, b) .
\end{aligned}
$$

Thus,

$$
\frac{\widetilde{W}_{x}^{\psi^{2}}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}=\frac{1}{i 2 \pi}\left(4 \pi^{2} \sigma^{2}(b)(a c-\mu)^{2}-1\right)
$$

In this case the phase transformation $\omega_{x}$ of the Morlet's wavelet is defined by

$$
\widetilde{\omega}_{x}^{1 s t, a d p}(a, b)=\operatorname{Re}\left\{\frac{\partial_{b} \widetilde{W}_{x}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}\right\} .
$$

### 2.2.2 Adaptive 2nd-order WSST

The adaptive 2nd-order WSST is proposed in [14]. To find the adaptive 2nd-order WSST, the CWT is a time-varying parameter. Let $\psi_{\sigma}(t)$ of the continuous wavelet be defined as

$$
\psi_{\sigma}(t)=\frac{1}{\sigma} g\left(\frac{t}{\sigma}\right) e^{i 2 \pi \mu t}
$$

Denote $\psi_{\sigma}^{2}(t)=\frac{t}{\sigma^{2}} g^{\prime}\left(\frac{t}{\sigma}\right) e^{i 2 \pi \mu t}$, and let $\widetilde{W_{x}} \psi^{2}(a, b)$ denote the CWT. We define

$$
\psi_{\sigma}^{1}(t)=\frac{t}{\sigma} \psi_{\sigma}(t)=\frac{t}{\sigma^{2}} g\left(\frac{t}{\sigma}\right) e^{i 2 \pi \mu t} .
$$

Let $W_{x}^{\psi_{1}}(a, b)$ denote the CWT of $x(t)$ with $\psi_{\sigma}^{1}(t)$

$$
\begin{gather*}
\widetilde{W}_{x}^{\psi^{1}}(a, b)=\int_{-\infty}^{\infty} x(t) \frac{1}{a} \overline{\psi_{\sigma}^{1}(t)\left(\frac{t-b}{a}\right)} d t=\int_{-\infty}^{\infty} x(b+a t) \frac{t}{\sigma^{2}(b)} \overline{g\left(\frac{t}{\sigma(b)}\right) e^{-i 2 \pi \mu t}} d t  \tag{2.2.1}\\
\hat{\psi}_{\sigma}^{1}(\xi)=\frac{i}{2 \pi}(\hat{g})^{\prime}(\sigma(\xi-\mu)) .
\end{gather*}
$$

If $\psi_{\sigma}$ is Morlet's wavelet defined by

$$
\hat{\psi}_{\sigma}(\xi)=e^{-2 \pi^{2} \sigma^{2}(\xi-\mu)^{2}}
$$

then

$$
\hat{\psi}_{\sigma}^{1}(\xi)=-i 2 \pi \sigma(\xi-\mu) e^{-2 \pi^{2} \sigma^{2}(\xi-\mu)^{2}}
$$

For a similar calculation presented in the case of 2nd-order WSST non adaptive, the phase transformation $\widetilde{\omega}^{2 n d, a d p}$ is defined by

$$
\widetilde{\omega}_{x}^{2 n d, a d p}= \begin{cases}\operatorname{Re}\left\{\frac{\partial_{b} \widetilde{W}_{x}(b, a)}{i 2 \pi \widetilde{W}_{x}(b, a)}\right\}+\frac{\sigma^{\prime}(b)}{\sigma(b)} R e\left\{\frac{\widetilde{W}_{x}^{\psi^{2}}(b, a)}{i 2 \pi \widetilde{W}_{x}(b, a)}\right\}- & a \operatorname{Re}\left\{\frac{\widetilde{W}_{x}^{\psi^{1}}(b, a)}{i 2 \pi W_{x}(b, a)} R_{0}(a, b)\right\}, \\ & \text { if } \frac{\partial}{\partial a}\left(a \frac{W^{\psi} \psi^{1}(b, a)}{\widetilde{W}_{x}(b, a)}\right) \neq 0, \widetilde{W}_{x}(b, a) \neq 0 \\ \operatorname{Re}\left\{\frac{\partial_{b} \widetilde{W}_{x}(b, a)}{i 2 \pi \widetilde{W}_{x}(b, a)}\right\}+\frac{\sigma^{\prime}(b)}{\sigma(b)} R e\left\{\frac{\widetilde{W}_{x}^{\psi^{2}}(b, a)}{i 2 \pi \widetilde{W}_{x}(b, a)}\right\}, & \text { if } \frac{\partial}{\partial a}\left(a \frac{\widetilde{W}_{x}^{\psi^{1}}(b, a)}{\widetilde{W}_{x}(b, a)}\right)=0, \widetilde{W}_{x}(b, a) \neq 0 .\end{cases}
$$

where

$$
R_{0}(a, b)=\frac{1}{\frac{\partial}{\partial a}\left(a \frac{\widetilde{W}^{\psi^{1}}(b, a)}{\widetilde{W}_{x}(b, a)}\right)}\left(\frac{\partial}{\partial a}\left(\frac{\frac{\partial}{\partial b} \widetilde{W}_{x}(b, a)}{\widetilde{W}_{x}(b, a)}\right)+\frac{\sigma^{\prime}(b)}{\sigma(b)} \frac{\partial}{\partial a}\left(\frac{\widetilde{W}_{x}{ }^{2}(b, a)}{\widetilde{W}_{x}(b, a)}\right)\right)
$$

The 2nd-order adaptive WSST of a signal $x(t)$ is defined by

$$
\begin{equation*}
T_{s}^{2 n d, a d p}(\xi, b)=\int_{\left\{a \in \mathbb{R}_{+}: \widetilde{W}_{s} \neq 0\right\}} \widetilde{W}_{x}(a, b) \delta\left(\widetilde{\omega}_{x}^{2 n d, a d p}(a, b)-\xi\right) \frac{d a}{a} \tag{2.2.2}
\end{equation*}
$$

where $\xi$ is the frequency variable.
The input signal $x(t)$ can be recovered from its adaptive WSST for analytic $x(t) \in L_{2}(\mathbb{R})$, and we have

$$
x(b)=\frac{1}{C_{\psi}(b)} \int_{0}^{\infty} T_{x}^{2 n d, a d p}(\xi, b) d \xi
$$

and for a real valued $x(t) \in L_{2}(\mathbb{R})$

$$
x(b)=\operatorname{Re}\left(\frac{2}{C_{\psi}(b)} \int_{0}^{\infty} T_{x}^{2 n d, a d p}(\xi, b) d \xi\right)
$$

where $c_{\psi}(b)$ is defined by

$$
C_{\psi}(b)=\int_{0}^{\infty} \overline{\hat{\psi}_{\sigma(b)}(\xi)} \frac{d \xi}{\xi}=\int_{0}^{\infty} \overline{\hat{g}(\xi-\sigma(b) \mu)} \frac{d \xi}{\xi} .
$$

For a multicomponent signal $x(t)$, each component $x_{k}(b)$ can be recovered from adaptive WSST

$$
\begin{equation*}
x_{k}(b)=\operatorname{Re}\left(\frac{2}{C_{\psi}(b)} \int_{\left|\xi-\phi^{\prime}(b)\right|<\Gamma_{1}} T_{x}^{2 n d, a d p}(\xi, b) d \xi\right) \tag{2.2.3}
\end{equation*}
$$

for certain $\Gamma_{1}>0$.

### 2.3 Short-time Fourier transform-based synchrosqueezed transform (FSST)

Definition 2.3.1. The short-time Fourier transform (STFT) of $x(t) \in$ $L_{2}(\mathbb{R})$ with a window function $g(t) \in L_{2}(\mathbb{R})$ is defined by

$$
\begin{align*}
V_{x}(t, \eta) & =\int_{-\infty}^{\infty} x(\tau) g(\tau-t) e^{-i 2 \pi \eta(\tau-t)} d \tau  \tag{2.3.1}\\
& =\int_{-\infty}^{\infty}(x+\tau) g(\tau) e^{-i 2 \pi \eta \tau} d \tau \tag{2.3.2}
\end{align*}
$$

where $t$ is the time variable and $\eta$ is the frequency variable.
The STFT is written as follows

$$
V_{x}(t, \eta)=\int_{-\infty}^{\infty} x(\tau) g(\tau-t) e^{-i 2 \pi \eta(\tau-t)} d \tau=\int_{-\infty}^{\infty} \hat{x}(\xi) \hat{g}(\eta-\xi) e^{i 2 \pi t \xi} d \xi(2.3 .3)
$$

To prove the result in equation (2.3.3) we start writing the short-time Fourier transform $V_{x}$ in the form

$$
V_{x}(t, \eta)=<x, y_{t, \eta}>_{L_{2}(\mathbb{R})} \quad \text { with } \quad y_{t, \eta}(\tau)=g(\tau-t) e^{-i 2 \pi \eta(\tau-t)}
$$

which implies

$$
\begin{equation*}
V_{x}(t, \eta)=<\hat{x}, \hat{y}_{t, \eta}>_{L_{2}(\mathbb{R})}=\int_{-\infty}^{\infty} \hat{x}(\xi) \overline{\hat{y}_{t, \eta}(\xi)} d \xi \tag{2.3.4}
\end{equation*}
$$

and the calculation $\hat{y}_{t, \eta}(\xi)$ denotes

$$
\begin{gathered}
\hat{y}_{t, \eta}(\xi)=\int_{-\infty}^{\infty} g(\tau-t) e^{i 2 \pi \eta(\tau-t)} e^{-i 2 \pi \xi \tau} d \tau \\
\hat{y}_{t, \eta}(\xi)=\int_{-\infty}^{\infty} g(y) e^{i 2 \pi \eta y} e^{-i 2 \pi \xi(y-t)} d y=\underbrace{\int_{-\infty}^{\infty} g(y) e^{i 2 \pi(\eta-\xi)} d y}_{\hat{g}(\eta-\xi)} e^{i 2 \pi \xi t}=\hat{g}(\eta-\xi) e^{i 2 \pi \xi t} .
\end{gathered}
$$

Therefore, we have

$$
\begin{equation*}
V_{x}(t, \eta)=<\hat{x}, \hat{y}_{t, \eta}>_{L_{2}(\mathbb{R})}=\int_{-\infty}^{\infty} \hat{x}(\xi) \hat{g}(\eta-\xi) e^{i 2 \pi \xi t} d \xi \tag{2.3.6}
\end{equation*}
$$

We can recover the original signal $x(t)$ from its STFT:

$$
\begin{equation*}
x(t)=\frac{1}{\|g\|_{2}^{2}} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} V_{x}(t, \eta) \overline{g(t-\tau)} e^{-i 2 \pi \eta(\tau-t)} d \tau d \eta \tag{2.3.7}
\end{equation*}
$$

Theorem 2.3.2. (Inverse STFT ).
Involving $\xi$ suppose $g(t) \in L_{2}(\mathbb{R})$ with $g(0) \neq 0$ Then for $x(t) \in L_{2}(\mathbb{R})$, we have

$$
\begin{equation*}
x(t)=\frac{1}{g(0)} \int_{-\infty}^{\infty} V_{x}(t, \eta) d \eta \tag{2.3.8}
\end{equation*}
$$

Proof. From (2.3.3), we have

$$
\begin{aligned}
\int_{-\infty}^{\infty} V_{x}(t, \eta) d \eta & =\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \hat{x}(\xi) \hat{g}(\eta-\xi) e^{i 2 \pi t \xi} d \xi d \eta \\
& =\int_{-\infty}^{\infty} \hat{x}(\xi) e^{i 2 \pi t \xi} \int_{-\infty}^{\infty} \hat{g}(\eta) d \eta d \xi \\
& =\int_{-\infty}^{\infty} \hat{g}(\eta) e^{i 2 \pi \times 0 \times \eta} d \eta \int_{-\infty}^{\infty} \hat{x}(\xi) e^{i 2 \pi t \xi} d \xi \\
& =g(0) x(t)
\end{aligned}
$$

Theorem 2.3.3. (Inverse STFT).
Involving $\eta$ for real signal. Suppose the window function $g(t) \in L_{2}(\mathbb{R})$ is real with $g(0) \neq 0$. Then for a real-valued $x(t) \in L_{2}(\mathbb{R})$, we have

$$
\begin{equation*}
x(t)=\frac{2}{g(0)} \operatorname{Re}\left(\int_{0}^{\infty} V_{x}(t, \eta) d \eta\right) . \tag{2.3.9}
\end{equation*}
$$

Proof. Let $g(t)$ and $x(t)$ be two real functions, we have

$$
\begin{aligned}
V_{x}(t,-\eta) & =\int_{-\infty}^{\infty} x(\tau) g(\tau-t) e^{i 2 \pi \eta(\tau-t)} d \tau \\
& =\int_{-\infty}^{\infty} x(\tau) g(\tau-t) \overline{e^{-i 2 \pi \eta(\tau-t)}} d \tau=\overline{V_{x}(t, \eta)}
\end{aligned}
$$

Hence,

$$
\begin{aligned}
x(t) & =\frac{1}{g(0)} \int_{0}^{\infty} V_{x}(t, \eta) d \eta+\frac{1}{g(0)} \int_{-\infty}^{0} V_{x}(t, \eta) d \eta \\
& =\frac{1}{g(0)} \int_{0}^{\infty} V_{x}(t, \eta) d \eta+\frac{1}{g(0)} \int_{0}^{\infty} V_{x}(t,-\eta) d \eta \\
& =\frac{1}{g(0)} \int_{0}^{\infty} V_{x}(t, \eta) d \eta+\frac{1}{g(0)} \int_{0}^{\infty} V_{x}(t, \eta) d \eta=\frac{2}{g(0)} \operatorname{Re}\left\{\int_{0}^{\infty} V_{x}(t, \eta) d \eta\right\} .
\end{aligned}
$$

### 2.3.1 1st-order FSST

The 1st-order FSST was proposed in $[12,15]$. We derive the FSST through the STFT. We begin with the STFT of $x(t)=A e^{i 2 \pi c t}$ (for more detail, see [12]) where $A$ and $c$ are constants and $c>0$. We have

$$
\begin{aligned}
V_{x}(t, \eta) & =\int_{-\infty}^{\infty} A e^{i 2 \pi c(t+\tau)} g(\tau) e^{-i 2 \pi \eta \tau} d \tau \\
& =A e^{i 2 \pi t c} \hat{g}(\eta-c)
\end{aligned}
$$

Thus, we can obtain the instantaneous frequency (IF), defined here as $c$, of $x(t)$ by

$$
\frac{\frac{\partial}{\partial t} V_{x}(t, \eta)}{2 i \pi V_{x}(t, \eta)}=c
$$

For a general signal $x(t)$, at $(t, \eta)$ for which $V_{x}(t, \eta) \neq 0$ and where $c=$ $\omega_{x}(t, \eta)$, a good candidate for the instantaneous frequency (IF) of $x(t)$ is

$$
\omega_{x}^{1 s t}(t, \eta)=\frac{\frac{\partial}{\partial t} V_{x}(t, \eta)}{2 i \pi V_{x}(t, \eta)}
$$

This is also called the "reference IF function" or the "phase transformation."

The FSST reassigns the frequency variable $\xi$ by transforming the STFT $V_{x}(t, \xi)$ of $x(t)$ to a quantity, denoted by $R_{x}^{1 s t}(t, \eta)$, on the time-frequency plane:

$$
R_{x}^{1 s t}(t, \xi)=\int_{\left\{\zeta: V_{x}(t, \zeta) \neq 0\right\}} V_{x}(t, \zeta) \delta\left(\omega_{x}^{1 s t}(t, \zeta)-\xi\right) d \zeta
$$

where $\xi$ is the frequency variable.
We can recover the input signal $x(t)$ from its FSST. If we say $g(t) \in L_{2}(\mathbb{R})$ with a window function where $g(0) \neq 0$, then for $x(t) \in L_{2}(\mathbb{R})$,

$$
\begin{equation*}
x(t)=\frac{1}{g(0)} \int_{-\infty}^{\infty} R_{x}^{1 s t}(t, \xi) d \xi \tag{2.3.1}
\end{equation*}
$$

If $g(t)$ and $x(t)$ are also real-valued, then

$$
\begin{equation*}
x(t)=\frac{2}{g(0)} \operatorname{Re}\left(\int_{0}^{\infty} R_{x}^{1 s t}(t, \xi) d \xi\right) . \tag{2.3.2}
\end{equation*}
$$

We denote a multicomponent signal $x(t)$ as

$$
x(t)=\sum_{k=1}^{K} x_{k}(t)=\sum_{k=1}^{K} A_{k}(t) e^{i 2 \pi \phi_{k}(t)} .
$$

when $A_{k}(t)$ and $\phi_{k}(t)$ satisfy certain conditions, each component $x_{k}(t)$ can be recovered from its FSST (see e.g. [12])

$$
x_{k}(t) \approx \frac{1}{g(0)} \int_{\left|\xi-\phi_{k}^{\prime}(t)\right|<\Gamma} R_{x}^{1 s t}(t, \xi) d \xi
$$

for certain $\Gamma>0$.

### 2.3.2 2nd-order FSST

Daubechies, Lu , and Wu introduce the 2nd-order SST in [7]. The focus of this paper to define a new phase transformation $\omega_{x}^{2 n d}$. This new phase transformation is associated with 2nd order derivatives of the STFT of $x(t)$ such that when $x(t)$ is a linear frequncy modulation (LFM) signal (linear chirp), then $\omega_{x}^{2 n d}$ is exactly the IF of $x(t)$. We say $x(t)$ is a LFM signal if

$$
x(t)=A(t) e^{i 2 \pi \phi(t)}=A e^{p t+\frac{q}{2} t^{2}} e^{i 2 \pi\left(c t+\frac{1}{2} r t^{2}\right)}
$$

with phase function $\phi(t)=c t+\frac{1}{2} r t^{2}$, IF $\phi^{\prime}(t)=c+r t$ and chirp rate $\phi^{\prime \prime}(t)=r$, and IA $A(t)=A e^{p t+\frac{q}{2} t^{2}}$, where $p, q$ are eral numbers and $|p|,|q|$ are much smaller than c. For $u_{1}(t)=t g(t), V_{x}^{u_{1}}(t, \eta)$ denotes the STFT of $x(t)$ with $u_{1}(t)$ (see e.g. [15])

$$
x^{\prime}(t)=(p+q t+i 2 \pi(c+r t)) x(t) .
$$

From

$$
V_{x}(t, \eta)=\int_{-\infty}^{\infty} x(t+\tau) g(\tau) e^{-i 2 \pi \eta \tau} d \tau
$$

we can say

$$
\begin{aligned}
\frac{\partial}{\partial t} V_{x}(t, \eta) & =\int_{-\infty}^{\infty} x^{\prime}(t+\tau) g(\tau) e^{-i 2 \pi \eta \tau} d \tau \\
& =\int_{-\infty}^{\infty}(p+q(t+\tau)+i 2 \pi(c+r t+t \tau)) x(t+\tau) g(\tau) e^{-i 2 \pi \eta \tau} d \tau \\
& =(p+q t+i 2 \pi(c+r t)) V_{x}(t, \eta)+(q+i 2 \pi r) V_{x}^{u_{1}}(t, \eta)
\end{aligned}
$$

Hence at $(t, \eta)$ on which $V_{x}(t, \eta) \neq 0$, we have

$$
\frac{\frac{\partial}{\partial t} V_{x}(t, \eta)}{V_{x}(t, \eta)}=p+q t+i 2 \pi(c+r t)+(q+i 2 \pi r) \frac{V_{x}^{u_{1}}(t, \eta)(t, \eta)}{V_{x}(t, \eta)}
$$

$$
\frac{\partial}{\partial \eta}\left(\frac{\frac{\partial}{\partial t} V_{x}(t, \eta)}{V_{x}(t, \eta)}\right)=(q+i 2 \pi r) U(t, \eta)
$$

where we use $U(t, \eta)$ to denote

$$
U(t, \eta)=\frac{\partial}{\partial \eta}\left(\frac{V_{x}^{u_{1}}(t, \eta)(t, \eta)}{V_{x}(t, \eta)}\right) .
$$

Thus if $U(t, \eta) \neq 0$, then

$$
\begin{gathered}
q+i 2 \pi r=\frac{1}{U(t, \eta)} \frac{\partial}{\partial \eta}\left(\frac{V_{x}^{u_{1}}(t, \eta)}{V_{x}(t, \eta)}\right) \\
\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}(t, \eta)}=(p+q t)+i 2 \pi(c+r t)+\frac{1}{U(t, \eta)} \frac{\partial}{\partial \eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}(t, \eta)}\right)\left(\frac{V_{x}^{u_{1}}(t, \eta)}{V_{x}(t, \eta)}\right) .
\end{gathered}
$$

This implies

$$
c+r t=\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}-\frac{p+q t}{i 2 \pi}-\frac{V_{x}^{u_{1}}(t, \eta)}{V_{x}(t, \eta) U(t, \eta)} \cdot \frac{\partial}{\partial \eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}\right),
$$

So we have

$$
\phi^{\prime}(t)=c+r t=\operatorname{Re}\left\{\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}\right\}-\operatorname{Re}\left\{\frac{V_{x}^{u_{1}}(t, \eta)}{V_{x}(t, \eta) U(t, \eta)} \cdot \frac{\partial}{\partial \eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}\right)\right\} .
$$

Hence, the phase transformation $\omega_{x}^{2 n d}$ is defined as

$$
\omega_{x}^{2 n d}(t, \eta)= \begin{cases}\operatorname{Re}\left\{\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}\right\}-\operatorname{Re}\left\{\frac{V_{x}^{u_{1}}(t, \eta)}{V_{x}(t, \eta) U(t, \eta)} \cdot \frac{\partial}{\partial \eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}\right)\right\} . & \text { if } U(t, \eta) \neq 0, \quad V_{x}(t, \eta) \neq 0 \\ \operatorname{Re}\left\{\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}\right\}, & \text { if } U(t, \eta)=0, \quad V_{x}(t, \eta) \neq 0\end{cases}
$$

The FSST reassigns the frequency variable $\xi$ by transforming the STFT
$V_{x}(t, \eta)$ of $x(t)$ to a quantity, denoted by by $R_{x}(t, \xi)$, on the time-frequency plane:

$$
R_{x}^{2 n d}(t, \xi)=\int_{\left\{\zeta: V_{x}(t, \zeta) \neq 0\right\}} V_{x}(t, \zeta) \delta\left(\omega_{x}^{2 n d}(t, \zeta)-\xi\right) d \zeta
$$

where $\xi$ is the frequency variable.
We can recover the input signal $x(t)$ from its FSST. If $g(t) \in L_{2}(\mathbb{R})$ and is window function with $g(0) \neq 0$. Then for $x(t) \in L_{2}(\mathbb{R})$,

$$
\begin{equation*}
x(t)=\frac{1}{g(0)} \int_{-\infty}^{\infty} R_{x}^{2 n d}(t, \xi) d \xi \tag{2.3.1}
\end{equation*}
$$

If in addition, $g(t)$ and $x(t)$ are real-valued, then

$$
\begin{equation*}
x(t)=\frac{2}{g(0)} \operatorname{Re}\left(\int_{0}^{\infty} R_{x}^{2 n d}(t, \xi) d \xi\right) . \tag{2.3.2}
\end{equation*}
$$

For a multicomponent signal $x(t)$ given by

$$
x(t)=\sum_{k=1}^{K} x_{k}(t)=\sum_{k=1}^{K} A_{k}(t) e^{i 2 \pi \phi_{k}(t)}
$$

each component $x_{k}(t)$ can be recovered from its FSST

$$
x_{k}(t) \approx \frac{1}{g(0)} \int_{\left|\xi-\phi_{k}^{\prime}(t)\right|<\Gamma} R_{x}^{2 n d}(t, \xi) d \xi
$$

for certain $\Gamma>0$.

### 2.4 FSST with a time varying parameter

the window function given by $g_{\sigma(t)}(t)=\frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right)$ where $\sigma>0$ is a parameter, and $g \in L^{2}(\mathbb{R})$ is a function with $g(0) \neq 0$. If $g(t)=\frac{1}{\sqrt{2 \pi}} e^{-\frac{t^{2}}{2}}$, the $g_{\sigma(t)}(\tau)$ is the Gaussion window function. (see e.g. $[15,13]$ )

For a signal $x(t)$, we define the STFT of $x(t)$ with a $g_{\sigma(t)}$ parameter as

$$
\begin{aligned}
\widetilde{V}_{x}(t, \eta) & =\int_{-\infty}^{\infty} x(\tau) g_{\sigma(t)}(\tau-t) e^{-i 2 \pi \eta(\tau-t)} d \tau \\
& =\int_{-\infty}^{\infty} x(\tau+t) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta(\tau)} d \tau \\
& =\int_{-\infty}^{\infty} \hat{x}(\xi) \hat{g}_{\sigma(t)}(x-\xi) e^{i 2 \pi t \xi} d \xi \\
& =\int_{-\infty}^{\infty} \hat{x}(\xi) \hat{g}(\sigma(t)(\eta-\xi)) e^{i 2 \pi t \xi} d \xi
\end{aligned}
$$

A signal $x(t)$ can be recovered from its adaptive STFT

$$
x(t)=\frac{\sigma(t)}{g(0)} \int_{-\infty}^{\infty} \widetilde{V}_{x}(t, \eta) d \eta .
$$

If in addition $g(t)$ is real-valued, then for real-valued $x(t)$, we have

$$
x(t)=\frac{2 \sigma(t)}{g(0)} \operatorname{Re}\left(\int_{-\infty}^{\infty} \widetilde{V}_{x}(t, \eta) d \eta\right) .
$$

### 2.4.1 Adaptive 1st-order FSST

In order to define the adaptive 1st-order FSST introduced in [14], we must define the phase transformation $\omega_{x}^{a d p}$ associated with the adaptive STFT. Let $g_{\sigma}(t)$ and $g_{\sigma}^{2}(t)=\frac{t}{\sigma^{2}} g^{\prime}$ be the continuous wavelet. $\widetilde{V}_{x}^{g^{2}}(t, \eta)$ is defined by

$$
\widetilde{V}_{x}^{g^{2}}(t, \eta)=\int_{-\infty}^{\infty} x(\tau+t) \frac{\tau}{\sigma^{2}(t)} g^{\prime}\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau
$$

To define the phase transformation $\omega_{x}^{a d p}$, we consider $x(t)=A e^{i 2 \pi c t}$, so that

$$
\widetilde{V}_{x}(t, \eta)=\int_{-\infty}^{\infty} x(t+\tau) g_{\sigma(t)}(\tau) e^{-i 2 \pi \eta \tau} d \tau=A \int_{-\infty}^{\infty} e^{i 2 \pi c(t+\tau)} \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau
$$

therefore, we have

$$
\begin{aligned}
\frac{\partial \widetilde{V}_{x}}{\partial t}(t, \eta) & =A \int_{-\infty}^{\infty}(i 2 \pi c) e^{i 2 \pi c(t+\tau)} \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& +A \int_{-\infty}^{\infty} e^{i 2 \pi c(t+\tau)}\left(\frac{-\sigma^{\prime}(t)}{\sigma^{2}(t)}\right) g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& +A \int_{-\infty}^{\infty} e^{i 2 \pi c(t+\tau)}\left(\frac{-\sigma^{\prime}(t)}{(\sigma(t))^{3}} \tau\right) g^{\prime}\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& =i 2 \pi c \widetilde{V}_{x}(t, \eta)-\frac{\sigma^{\prime}(t)}{\sigma(t)} \widetilde{V}_{x}(t, \eta)-\frac{\sigma^{\prime}(t)}{\sigma(t)} \widetilde{V}_{x}^{g^{2}}(t, \eta)
\end{aligned}
$$

If $\widetilde{V}_{x}(t, \eta) \neq 0$, we then have

$$
\frac{\frac{\partial}{\partial t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}=c-\frac{\sigma^{\prime}(t)}{2 i \pi \sigma(t)}-\frac{\sigma^{\prime}(t)}{\sigma(t)} \cdot \frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}
$$

The IF of $x(t)$ can therefore be obtained by

$$
c=\operatorname{Re}\left\{\frac{\frac{\partial}{\partial t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}+\frac{\sigma^{\prime}(t)}{2 i \pi \sigma(t)}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \cdot \frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}
$$

Moreover, if $\sigma$ is real function $\operatorname{Re}\left\{\frac{\sigma^{\prime}(t)}{2 i \pi \sigma(t)}\right\}=0$, we have

$$
c=\operatorname{Re}\left\{\frac{\frac{\partial}{\partial t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \operatorname{Re}\left\{\frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}
$$

This quantity is also called the phase transformation or the reference IF function, and we denote it by $\tilde{\omega}_{x}(t, \eta)$ :

$$
\tilde{\omega}_{x}^{1 s t, a d p}(t, \eta)=\operatorname{Re}\left\{\frac{\frac{\partial}{\partial t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \operatorname{Re}\left\{\frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}, \text { if } \widetilde{V}_{x}(t, \eta) \neq 0
$$

The adaptive FSST of $x(t)$ with a $\sigma(t)$ parameter is defined by

$$
\begin{equation*}
R_{x}^{1 s t, a d p}(t, \xi)=\int_{\left\{\eta \in \mathbb{R}: \widetilde{V}_{x}(t, \eta) \neq 0\right\}} \widetilde{V}_{x}(t, \eta) \delta\left(\tilde{\omega}_{x}^{1 s t, a d p}(t, \eta)-\xi\right) d \eta \tag{2.4.1}
\end{equation*}
$$

where $\xi$ is the frequency variable.
The input signal $x(t)$ can be recovered from its adaptive FSST:

$$
x(t)=\frac{\sigma(t)}{g(0)} \int_{-\infty}^{\infty} R_{x}^{1 s t, a d p}(t, \xi) d \xi
$$

If $x(t)$ is also real-valued, then for real-valued $x(t)$, we denote that

$$
x(t)=\frac{2 \sigma(t)}{g(0)} \operatorname{Re}\left(\int_{-\infty}^{\infty} R_{x}^{1 s t, a d p}(t, \xi) d \xi\right) .
$$

We may use the following formula to recover the $k$-th component $x_{k}(b)$ of a multicomponent signal from the adaptive FSST

$$
x_{k}(b)=\frac{2 \sigma(t)}{g(0)} \operatorname{Re}\left(\int_{\left|\xi-\phi^{\prime}(t)\right|<\Gamma} R_{x}^{1 s t, a d p}(\xi, b) d \eta\right)
$$

for certain $\Gamma>0$.

## Remark:

If $g$ is the Gaussion function defined by $g(t)=\frac{1}{\sqrt{2 \pi}} e^{-\frac{t^{2}}{2}}$ for a signal $x(t)=A e^{i 2 \pi c t}$, we have (refer to [15])

$$
\begin{aligned}
\tilde{V}_{x}(t, \eta) & =\int_{-\infty}^{\infty} A e^{i 2 \pi c(t+\tau)} g_{\sigma(t)}(\tau) e^{-i 2 \pi \eta \tau} d \tau \\
& =A e^{i 2 \pi c t} \hat{g}_{\sigma(t)}(\eta-c) \\
& =A e^{i 2 \pi c t} e^{-2 \pi^{2} \sigma(t)^{2}(\eta-c)^{2}}
\end{aligned}
$$

$$
\begin{aligned}
\partial_{t} \widetilde{V}_{x}(t, \eta) & =i 2 \pi c A e^{i 2 \pi c t} e^{-2 \pi^{2} \sigma(t)^{2}(\eta-c)^{2}}+A e^{i 2 \pi c t} e^{-2 \pi^{2} \sigma(t)^{2}(\eta-c)^{2}}\left(-4 \pi^{2}\right)(\eta-c)^{2} \sigma(t) \sigma^{\prime}(t) \\
& =i 2 \pi c \widetilde{V}_{x}(t, \eta)+\widetilde{V}_{x}(t, \eta)\left(-4 \pi^{2}\right)(\eta-c)^{2} \sigma(t) \sigma^{\prime}(t)
\end{aligned}
$$

Thus, we have

$$
\frac{\partial_{t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}=c+i 2 \pi(\eta-c)^{2} \sigma(t) \sigma^{\prime}(t)
$$

Both $c$ and $2 \pi(\eta-c)^{2} \sigma(t) \sigma^{\prime}(t)$ are real, so the phase transformation of $x(t)$ can be obtained by

$$
c=R e\left\{\frac{\partial_{t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\} .
$$

Thus if $g$ is the Gaussian function for a general signal $x(t)$, we may write the phase transformation as

$$
\begin{gathered}
\widetilde{\omega}_{x}^{1 s t, a d p}=\operatorname{Re}\left\{\frac{\partial_{t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)},\right\}, \text { for } \widetilde{V}_{x}(t, \eta) \neq 0 . \\
\widetilde{\omega}_{x}^{1 s t, a d p}(t, \eta)= \begin{cases}\operatorname{Re}\left\{\frac{\partial_{t} \widetilde{V}_{x}(t, \eta)}{i 2 \widetilde{\widetilde{T}}_{x}(t, \eta)}\right\} & \text { if } \mathrm{g} \text { (Gaussion function) } \\
\operatorname{Re}\left\{\frac{\partial_{t} \vec{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \operatorname{Re}\left\{\frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}, & \text { otherwise }\end{cases}
\end{gathered}
$$

### 2.4.2 Adaptive 2nd-order FSST

In order to define the adaptive 2nd-order FSST, we apply a timevarying parameter to the STFT. Let $g_{\sigma(t)}(t)=\frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right)$ be the window function. We define $g_{\sigma(t)}^{1}(t)=\frac{t}{\sigma^{2}(t)} g\left(\frac{t}{\sigma(t)}\right)$ and $g_{\sigma(t)}^{2}(t)=\frac{t}{\sigma^{2}(t)} g^{\prime}\left(\frac{t}{\sigma(t)}\right)$ (see e.g. [10]).

We use $\widetilde{V}_{x}^{g^{1}}(t, \eta)$ to denote the STFT defined by $\widetilde{V}_{x}^{g^{1}}(t, \eta)=\int_{-\infty}^{\infty} x(\tau) g_{\sigma(t)}^{\prime}(\tau-t) e^{-i 2 \pi \eta(\tau-t)} d \tau=\int_{-\infty}^{\infty} x(t+\tau) \frac{\tau}{\sigma^{2}(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta(\tau-t)} d \tau$ and $\widetilde{V}_{x}^{g^{2}}(t, \eta)$ to denote the STFT defined by

$$
\widetilde{V}_{x}^{g^{2}}(t, \eta)=\int_{-\infty}^{\infty} x(\tau+t) \frac{\tau}{\sigma^{2}(t)} g^{\prime}\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau
$$

we then can note that

$$
\begin{aligned}
\partial_{t} \widetilde{V}_{x}(t, \eta) & =\int_{-\infty}^{\infty} x^{\prime}(t+\tau) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau+\int_{-\infty}^{\infty} x(t+\tau)\left(-\frac{\sigma^{\prime}(t)}{\sigma(t)^{2}}\right) g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& +\int_{-\infty}^{\infty} x(t+\tau)\left(-\frac{\sigma^{\prime}(t)}{\sigma(t)^{3}}\right) g^{\prime}\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& =(p+q t+i 2 \pi(c+r t)) \widetilde{V}_{x}(t, \eta)+(q+i 2 \pi r) \int_{-\infty}^{\infty} \tau x(t+\tau) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& -\frac{\sigma^{\prime}(t)}{\sigma(t)} \widetilde{V}_{x}(t, \eta)-\frac{\sigma^{\prime}(t)}{\sigma(t)} \widetilde{V}_{x}^{g^{2}}(t, \eta) \\
& =\left(p+q t+i 2 \pi(c+r t)-\frac{\sigma^{\prime}(t)}{\sigma(t)}\right) \widetilde{V}_{x}(t, \eta)+(q+i 2 \pi r) \sigma(t) \widetilde{V}_{x}^{g^{2}}(t, \eta)-\frac{\sigma^{\prime}(t)}{\sigma(t)} \widetilde{V}_{x}^{g^{2}}(t, \eta),
\end{aligned}
$$

and if $\widetilde{V}_{x}(t, \eta) \neq 0$, we have

$$
\begin{aligned}
& \frac{\partial_{t} \widetilde{V}_{x}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}=p+q t+\frac{\sigma^{\prime}(t)}{\sigma(t)}+i 2 \pi(c+r t)+(q+i 2 \pi r) \sigma(t) \frac{\widetilde{V}_{x}^{g^{1}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}-\frac{\sigma^{\prime}(t)}{\sigma(t)} \frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)} \\
& \quad \frac{\partial}{\partial \eta}\left(\frac{\partial_{t} \widetilde{V}_{x}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right)=(q+i 2 \pi r) \sigma(t) \frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{x}^{g^{1}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right)-\frac{\sigma^{\prime}(t)}{\sigma(t)}\left(\frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right) .
\end{aligned}
$$

Therefore, if in addition, $\frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{s}^{g^{1}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right) \neq 0$, then $(q+i 2 \pi r) \sigma(t)=R_{0}(t, \eta)$ with

$$
\begin{aligned}
& q+i 2 \pi r=\frac{R_{0}(t, \eta)}{\sigma(t)} \\
& \frac{\partial_{t} \widetilde{V}_{x}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}=p+q t-\frac{\sigma^{\prime}(t)}{\sigma(t)}+i 2 \pi(c+r t)+R_{0}(t, \eta) \frac{\widetilde{V}_{x}^{g^{1}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}-\frac{\sigma^{\prime}(t)}{\sigma(t)}\left(\frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right) .
\end{aligned}
$$

Thus
$\phi^{\prime}(t)=c+r t=\operatorname{Re}\left\{\frac{\partial_{t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}-\operatorname{Re}\left\{\frac{\widetilde{V}_{x}^{g^{1}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)} R_{0}(t, \eta)\right\}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \operatorname{Re}\left\{\frac{\widetilde{V}_{x}^{g^{2}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right\}$.
For a signal $x(t)$ in the following, we define the phase transformation $\widetilde{\omega}_{x}^{2 n d, a d p}$ as

The adaptive 2nd-orded FSST of a signal $x(t)$ is defined by

$$
\begin{equation*}
R_{x}^{2 n d, a d p}(t, \xi)=\int_{\left\{\eta \in \mathbb{R}: \tilde{V}_{x}(t, \eta) \neq 0\right\}} \widetilde{V}_{x}(t, \eta) \delta\left(\tilde{\omega}_{x}^{2 n d, a d p}(t, \eta)-\xi\right) d \eta \tag{2.4.2}
\end{equation*}
$$

where $\xi$ is the frequency variable.
To recover the input signal $x(t)$ from its adaptive FSST, we have

$$
x(t)=\frac{\sigma(t)}{g(0)} \int_{-\infty}^{\infty} R_{x}^{2 n d, a d p}(t, \xi) d \xi .
$$

If in addition $x(t)$ is real-valued, then for real-valued $x(t)$, we have

$$
x(t)=\frac{2 \sigma(t)}{g(0)} \operatorname{Re}\left(\int_{-\infty}^{\infty} R_{x}^{2 n d, a d p}(t, \xi) d \xi\right)
$$

We may use the following formula to recover the $k$-th component $x_{k}(b)$ of a multicomponent signal from the adaptive FSST

$$
x_{k}(b)=\frac{2 \sigma(t)}{g(0)} \operatorname{Re}\left(\int_{\left|\xi-\phi^{\prime}(t)\right|<\Gamma} R_{x}^{2 n d, a d p}(\xi, b) d \eta\right)
$$

for certain $\Gamma>0$.

### 2.5 Analysis of FSST

In this section, many pre-established concepts are reiterated and then expanded upon. Refrences to these materials can be found in [19].

### 2.5.1 STFT-based synchrosqueezing transform

The short-time Fourier transform (STFT) of $x(t) \in L_{2}(\mathbb{R})$ with a window function $g(t) \in L_{2}(\mathbb{R})$ given by (2.3.3). For a signal $x(t)$, at $(t, \eta)$ for which $V_{x}(t, \eta) \neq 0$, note that

$$
\omega_{x}(t, \eta)=\operatorname{Re}\left(\frac{\partial_{t} V_{x}(t, \eta)}{2 i \pi V_{x}(t, \eta)}\right)
$$

The FSST reassigns the frequency variable $\tau$ by transforming $\operatorname{STFT} V_{x}(t, \eta)$ of $x(t)$ to a quantity, denoted by

$$
\begin{equation*}
R_{x, \gamma}(t, \xi)=\int_{\left|V_{x}(t, \eta)\right| \neq \gamma} V_{x}(t, \eta) \frac{1}{\lambda} h\left(\frac{\left.\xi-\omega_{x}(t, \eta)\right)}{\lambda}\right) d \eta \tag{2.5.1}
\end{equation*}
$$

where $h(t)$ is a compactly supported function with certain smoothness and
$\int_{\mathbb{R}} h(t) d t=1 . \quad \int_{\left|V_{x}(t, \eta)\right|>\gamma}$ means the integral $\int_{\left\{\eta:\left|V_{x}(t, \eta)\right|>\gamma\right\}}$ with $\eta$ over the set $\left\{\eta:\left|V_{x}(t, \eta)\right|>\gamma\right\}$. We consider multicomponent signals $x(t)$, defined as follows

$$
\begin{equation*}
x(t)=\sum_{k=1}^{K} x_{k}(t), \quad \text { with } x_{k}(t)=A_{k}(t) e^{i 2 \pi \phi_{k}(t)} \tag{2.5.2}
\end{equation*}
$$

where $A_{k}(t)$ and $\phi_{k}(t)$ satisfy

$$
\begin{align*}
& A_{k}(t) \in C^{1}(\mathbb{R}) \cap L_{\infty}(\mathbb{R}), \phi_{k}(t) \in C^{2}(\mathbb{R})  \tag{2.5.3}\\
& A_{k}(t)>0, \inf _{t \in R} \phi_{k}^{\prime}(t)>0, \sup _{t \in R} \phi_{k}^{\prime}(t)<\infty \tag{2.5.4}
\end{align*}
$$

Let $\varepsilon>0$ and $\Delta>0$, and let $\beta_{\varepsilon, \Delta}$ denote the set of multicomponent signals satisfying (2.5.3), (2.5.4), and the following condition:

$$
\begin{array}{r}
\left|A_{k}(t)\right| \leq \varepsilon \phi_{k}^{\prime}(t),\left|\phi_{k}(t)\right| \leq \varepsilon \phi_{k}^{\prime \prime}(t), t \in \mathbb{R}, M_{k}^{\prime \prime}=\sup _{t \in \mathbb{R}}\left|\phi_{k}(t)\right|<\infty \\
\phi_{k}^{\prime}(t)-\phi_{k-1}^{\prime}(t) \geq 2 \Delta, \quad 2 \leq k \leq K, t \in \mathbb{R} . \tag{2.5.6}
\end{array}
$$

The condition (2.5.6) is called the well-separated condition with resolution $\Delta$. For the well-separated condition, [9] uses a stronger condition than that in (2.5.6):

$$
\begin{equation*}
\inf _{t \in \mathbb{R}} \phi_{k}^{\prime}(t)-\sup _{t \in \mathbb{R}} \phi_{k-1}^{\prime}(t) \geq 2 \Delta, \quad 2 \leq k \leq K \tag{2.5.7}
\end{equation*}
$$

The condition (2.5.5) considered in [12] implies that $A_{k}(t)$ and IF $\phi_{k}^{\prime}(t)$ change slowly as compared with $\phi_{k}^{\prime}(t)$. The Fourier-based synchrosqueezing transform, [25], uses another condition for the change of $A_{k}(t)$ and IF $\phi_{k}^{\prime}(t)$ :

$$
\begin{equation*}
\left|A_{k}^{\prime}(t)\right| \leq \varepsilon, \quad \text { and } \quad\left|\phi_{k}^{\prime}(t)\right| \leq \varepsilon, \quad \text { for } t \in \mathbb{R} \tag{2.5.8}
\end{equation*}
$$

Let $\beta_{\varepsilon, \Delta}$ denote the set of multicomponent signals that satisfy (2.5.3), (2.5.4), (2.5.8), and well-separated condition (2.5.6). For $1 \leq k \leq K$ let

$$
\begin{equation*}
Z_{k}=\left\{\eta:\left|\eta-\phi_{k}^{\prime}(t)\right|<\Delta\right\} . \tag{2.5.9}
\end{equation*}
$$

Hence, the well-separated condition (2.5.6) implies that $Z_{k}$ are not overlapping. We denote

$$
\begin{equation*}
\Gamma_{0}(t)=M I_{1}+\pi I_{2} \sum_{k=1}^{K} A_{k}(t), \text { and } \widetilde{\Gamma_{0}}(t)=M \widetilde{I}_{1}+\pi \widetilde{I}_{2} \sum_{k=1}^{K} A_{k}(t), \tag{2.5.10}
\end{equation*}
$$

where

$$
\begin{equation*}
I_{n}=\int_{\mathbb{R}}\left|\tau^{n} g(\tau)\right| d \tau, \text { and } \widetilde{I}_{n}=\int_{\mathbb{R}}\left|\tau^{n} g^{\prime}(\tau)\right| d \tau, \text { for } n=1,2, \ldots \tag{2.5.11}
\end{equation*}
$$

Theorem 2.5.1. As stated in [19], let $x(t) \in B_{\varepsilon, \Delta}$ and $g$ be a function in the Schwartz class with $\operatorname{supp}(\hat{g}) \subseteq[-\Delta, \Delta]$. Let $\Gamma_{0}(t), \widetilde{\Gamma_{0}}(t)$ be defined by (2.5.10). Then we have the following.
(a)Suppose $\widetilde{\varepsilon}$ satisfies $\widetilde{\varepsilon} \geq \varepsilon \Gamma_{0}(t)$. Then for any $\eta$ with $\left|V_{x}(t, \eta)\right|>\widetilde{\varepsilon}$, there exists a unique $k \in\{1,2, \ldots, K\}$ such that $(t, \eta) \in Z_{k}$.
(b) Suppose $(t, \eta)$ satisfies $\left|V_{x}(t, \eta)\right|>\widetilde{\varepsilon}$ and $(t, \eta) \in Z_{k}$. Then

$$
\begin{equation*}
\left|\omega_{x}(t, \eta)-\phi_{k}^{\prime}(t)\right|<\frac{\varepsilon}{\widetilde{\varepsilon}}\left(\Gamma_{0}(t) \Delta+\frac{1}{2 \pi} \widetilde{\Gamma_{0}}(t)\right) \tag{2.5.12}
\end{equation*}
$$

(c) Suppose that $\widetilde{\varepsilon}$ satisfies $\left(\Gamma_{0}(t) \Delta+\frac{1}{2 \pi} \widetilde{\Gamma_{0}}(t)\right) \frac{\varepsilon}{\tilde{\varepsilon}} \leq \widetilde{\varepsilon}_{3} \leq \Delta$, we have

$$
\begin{equation*}
\left|\lim _{\lambda \rightarrow 0} \frac{1}{g(0)} \int_{\left|\varepsilon-\phi_{k}^{\prime}\right|<\widetilde{\varepsilon}_{3}} R_{x, \widetilde{\varepsilon}}^{\lambda}(t, \xi) d \xi-x_{k}(t)\right| \leq \frac{2 \Delta\left(\varepsilon \Gamma_{0}(t)+\widetilde{\varepsilon}\right)}{|g(0)|} . \tag{2.5.13}
\end{equation*}
$$

(d) If $x(t) \in B_{\varepsilon, \Delta}$, then the above statements (a)-(c) hold with $\Gamma_{0}(t)$ and $\widetilde{\Gamma_{0}}(t)$ in (2.5.10) replaced by

$$
\begin{align*}
& \Gamma_{0}(t)=\sum_{k=1}^{K}\left\{\phi_{k}^{\prime}(t) I_{1}+\frac{1}{2} M_{k}^{\prime \prime} I_{2}+\pi A_{k}(t)\left(\phi_{k}^{\prime}(t) I_{2}+\frac{1}{3} M_{k}^{\prime \prime} I_{k}(2) .5 .1\right.\right. \\
& \widetilde{\Gamma_{0}}(t)=\sum_{k=1}^{K}\left\{\phi_{k}^{\prime}(t) \widetilde{I}_{1}+\frac{1}{2} M_{k}^{\prime} \widetilde{I}_{2}+\pi A_{k}(t)\left(\phi_{k}^{\prime}(t) \widetilde{I}_{2}+\frac{1}{3} M_{k}^{\prime} \widetilde{I_{3}}(2)\right\} .5 .1\right.
\end{align*}
$$

we may note that $\widetilde{\varepsilon}$ and $\widetilde{\varepsilon}_{3}$ in Theorem (2.5.1) could be a function of $t$. If we choose $\widetilde{\varepsilon}=\varepsilon^{\frac{1}{3}}$, and if $\varepsilon$ is small enough, such that

$$
\begin{equation*}
\widetilde{\varepsilon} \leq \min \left\{\Delta, \frac{1}{\left\|\Gamma_{0}(t) \Delta+\frac{1}{2 \pi} \widetilde{\Gamma_{0}}(t)\right\|_{\infty}}\right\} \tag{2.5.16}
\end{equation*}
$$

then $\widetilde{\varepsilon}\left(\Gamma_{0}(t) \Delta+\frac{1}{2 \pi} \widetilde{\Gamma_{0}}(t)\right) \leq 1$. Hence,

$$
\begin{equation*}
\left(\Gamma_{0}(t) \Delta+\frac{1}{2 \pi} \widetilde{\Gamma_{0}}(t)\right) \frac{\varepsilon}{\widetilde{\varepsilon}} \leq \widetilde{\varepsilon} \leq \Delta . \tag{2.5.17}
\end{equation*}
$$

Thus, the conditions in the Theorem (2.5.1) are satisfied, and Theorem (2.5.1) (with $\widetilde{\varepsilon}_{3}=\widetilde{\varepsilon}$ ) can be defined in theorem (2.5.2).

Theorem 2.5.2. As stated in [12, 19, 25], let $x(t) \in B_{\varepsilon, \Delta}$, and $\widetilde{\varepsilon}=\varepsilon^{\frac{1}{3}}$. Let $g$ be a function in he Schwartz class with $\operatorname{supp}(\hat{g}) \subseteq[-\Delta, \Delta]$. If $\varepsilon$ is small enough, then the following statements hold.
(a) For $(t, \eta)$ satisfying $\left|V_{x}(t, \eta)\right|>\widetilde{\varepsilon}$, there exists a unique $k \in\{1,2, \ldots, K\}$ such that $(t, \eta) \in Z_{k}$
(b) Suppose $(t, \eta)$ satisfies $\left|V_{x}(t, \eta)\right|>\widetilde{\varepsilon}$ and $(t, \eta) \in Z_{k}$ Then

$$
\begin{equation*}
\left|\omega_{x}(t, \eta)-\phi_{k}^{\prime}(t)\right|<\widetilde{\varepsilon} \tag{2.5.18}
\end{equation*}
$$

(c) For any $k \in\{1,2, \ldots, K\}$,

$$
\begin{equation*}
\left|\lim _{\lambda \rightarrow 0} \frac{1}{g(0)} \int_{\left|\varepsilon-\phi_{k}^{\prime}\right|<\widetilde{\varepsilon}} R_{x, \widetilde{\varepsilon}}^{\lambda}(t, \xi) d \xi-x_{k}(t)\right| \leq \frac{4 \Delta}{g(0)} \widetilde{\varepsilon} \tag{2.5.19}
\end{equation*}
$$

$" \varepsilon$ is small enough" in Theorem (2.5.2) implies that $\widetilde{\varepsilon}$ defined by $\widetilde{\varepsilon}=\varepsilon^{\frac{1}{3}}$ satisfies some inequalities like (2.5.16). Most theorems on the WSST and FSST analysis are stated in the form of Theorem 2.5.2, see e.g. [7, 9, 12, 18, 25]. Part (b) and part (c) in theorem 2.5.1 denote more direct bounds of the estimates. The quantity on the left-hand side (LHS) of (2.5.12) the IF estimate error. Additionally, we call that on LHS of (2.5.13) the error of component recovery or component separation. The statements in Theorem 2.5.1 can be found in $[12,18,25]$ with some different IF estimate errors. For example, $[18,25]$ gave IF estimate error $\frac{\varepsilon}{\tilde{\varepsilon}}\left(\Gamma_{0}(t)\left(\Delta+2 \phi_{k}^{\prime}(t)\right)+\frac{1}{2 \pi} \widetilde{\Gamma_{0}}(t)\right)$ instead of $\frac{\varepsilon}{\tilde{\varepsilon}}\left(\Gamma_{0}(t) \Delta+\frac{1}{2 \pi} \widetilde{\Gamma_{0}}(t)\right)$ in (2.5.12). One can also find that Theorem (2.5.1) is a special case of Theorem (2.6.1).

Observe that the condition (2.5.5) or (2.5.8) requires the slow change of the IF $\phi_{k}^{\prime}(t)$ of each component $x_{k}(t)$. There is no mathematical guarantee that the IF estimate and the component separation for a multicomponent signal $x(t)$ with a component $x_{k}(t)$ will have a fast-changing frequency. For example, the changing rate of IF of $x_{k}(t)$ is not very small in the second derivative $\phi_{k}^{\prime \prime}(t)$. To address this, the 2nd-order FSST was introduced in [10], and the 2nd-order WSST was proposed in [11]. The theoretical analysis of the 2nd-order FSST is established in [18].

If $V_{x}(t, \eta) \neq 0$ and $\partial_{t}\left(\frac{\partial_{\eta} V_{x}(t, \eta)}{V_{x}(t, \eta)}\right) \neq i 2 \pi$, then

$$
\begin{equation*}
\tilde{q}(t, \eta)=\frac{\partial_{t}\left(\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}(t, \eta)}\right)}{i 2 \pi-\partial_{t}\left(\frac{\partial_{\eta} V_{x}(t, \eta)}{V_{x}(t, \eta)}\right)} . \tag{2.5.20}
\end{equation*}
$$

The 2nd-order FSST in [18] is defined as

$$
\begin{equation*}
R_{x, \gamma}^{2 n d, \lambda}(t, \xi)=\int_{\left|V_{x}(t, \eta)\right|>\gamma} V_{x}(t, \eta) \frac{1}{\lambda} h\left(\frac{\xi-\omega_{x}^{2 n d}(t, \eta)}{\lambda}\right) d \eta \tag{2.5.21}
\end{equation*}
$$

where $\omega_{x}^{2 n d}(t, \eta)$ is the phase transformation for the 2nd-order FSST. For $(t, \eta)$ with $V_{x}(t, \eta) \neq 0$,
$\omega_{x}^{2 n d}(t, \eta)= \begin{cases}\operatorname{Re}\left\{\frac{\partial_{t} V_{x}(t, \eta)}{2 \pi i x_{x}(t, n)}\right\}+\operatorname{Re}\left\{\tilde{q}(t, \eta) \partial_{t}\left(\frac{\partial_{\eta} V_{x}(t, \eta)}{V_{x}(t, \eta)}\right)\right\} & \text { if } \quad \partial_{t}\left(\frac{\partial_{\eta} V_{x}(t, \eta)}{V_{x}(t, \eta)}\right) \neq i 2 \pi \\ \operatorname{Re}\left\{\frac{\partial_{t} V_{x}(t, \eta)}{2 \pi i V_{x}(t, \eta)}\right\} & \text { Otherwise. }\end{cases}$

Let $\varepsilon>0$ and $\Delta>0 . B_{\varepsilon, \Delta}^{(2)}$ is the set of multicomponent signals satisfying (2.5.4), the well-separated condition (2.5.6), and the condition:

$$
\begin{array}{r}
A_{k}(t) \in C^{2}(\mathbb{R}) \cap L_{\infty}(\mathbb{R}), \phi_{k}(t) \in C^{3}(\mathbb{R}), \phi_{k}^{\prime}(t) \in L_{\infty}(\mathbb{R}) \\
\left|A_{k}^{\prime}(t)\right| \leq \varepsilon,\left|A_{k}^{\prime \prime}(t)\right| \leq \varepsilon,\left|\phi_{k}^{(3)}(t)\right| \leq \varepsilon, t \in(\mathbb{R}) \tag{2.5.24}
\end{array}
$$

Then, when $x(t) \in B_{\varepsilon, \Delta}^{(2)}$, statements for the 2nd-order FSST that are similar to (2.5.2) hold under certain conditions that are more complicated than (2.5.16) because no hand-limited window functions and the 2nd-order phase transformation $\omega_{x}^{2 n d}(t, \eta)$ are involved. The deffinition of $B_{\varepsilon, \Delta}^{(2)}$ has no direct boundedness restriction on $\phi_{k}^{\prime \prime}(t)$. See [18] for the details.

### 2.6 Analysis of Adaptive FSST

In this section, many pre-established concepts are reiterated and then expanded upon. Refrences to these materials can be found in [19].

### 2.6.1 Adaptive FSST with a time-varying parameter

We begin with the window function given by

$$
\begin{equation*}
g_{\sigma}(t)=\frac{1}{\sigma} g\left(\frac{t}{\sigma}\right), \tag{2.6.1}
\end{equation*}
$$

where $\sigma>0$ is a parameter and $g(t) \in L_{2}(\mathbb{R})$ with $g(0) \neq 0$. In addition, $g(t)$ has a decaying order as $t \rightarrow \infty$. If

$$
\begin{equation*}
g(t)=\frac{1}{\sqrt{2 \pi}} e^{-\frac{t^{2}}{2}} \tag{2.6.2}
\end{equation*}
$$

then $g_{\sigma}(t)$ is the Gaussian window function. The parameter $\sigma$ is also the window width in the time-domain of the window function $g_{\sigma}(t)$ because the
time duration $\triangle_{g_{\sigma}}$ of $g_{\sigma}(t)$ is $\sigma$ (up to a constant ). $\triangle_{g_{\sigma}}=\sigma \triangle_{g}$, where $\triangle_{g}$ is the time duration of $g$.

For a signal $x(t)$, the STDT of $x(t)$ with a time-varying parameter, as defined in [15] is
\left.${\widetilde{V_{x}}}_{x}(t, \eta)=\int_{\mathbb{R}} x(\tau) g_{\sigma(t)}(\tau-t) e^{-i 2 \pi \eta(\tau-t)} d \tau=\int_{\mathbb{R}} x(t+\tau) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \mathcal{Z}, 6.3\right)$
where $\sigma=\sigma(t)$ is positive function of $\mathrm{t} . \widetilde{V}_{x}(t, \eta)$ is the adaptive STFT of $x(t)$ with $g_{\sigma}$.

Before review the SST associated with the adaptive STFT, we must introduce some notations used in this and the next two subsections. Note that

$$
g_{1}(\tau)=\tau g(\tau), g_{2}(\tau)=\tau^{2} g(\tau), g_{3}(\tau)=\tau g^{\prime}(\tau)
$$

Therefore,

$$
g_{1, \sigma}(\tau)=\frac{\tau}{\sigma^{2}} g\left(\frac{\tau}{\sigma}\right), g_{2, \sigma}(\tau)=\frac{\tau^{2}}{\sigma^{3}} g\left(\frac{\tau}{\sigma}\right), g_{3, \sigma}(\tau)=\frac{\tau}{\sigma^{2}} g^{\prime}\left(\frac{\tau}{\sigma}\right)
$$

We use $\widetilde{V}_{x}^{g_{i}}(t, \eta)$ and $\widetilde{V}_{x}^{g^{\prime}}(t, \eta)$ to denote the adaptive STFT is defined by (2.6.3). $g_{j, \sigma}$ replaces $g_{\sigma}$ and $g_{\sigma}^{\prime}(\tau)=\frac{1}{\sigma} g^{\prime}\left(\frac{\tau}{\sigma}\right)$ where $1 \leq j \leq 3$.

For $x(t)=A e^{i 2 \pi c t}$, we can prove that, if $\widetilde{V}_{x}(t, \eta)$, then

$$
\begin{equation*}
\omega_{x}^{a d p, c}(t, \eta)=\frac{\frac{\partial}{\partial t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}+\frac{\sigma^{\prime}(t)}{i 2 \pi \sigma(t)}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \frac{\widetilde{V}_{x}^{g_{3}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)} \tag{2.6.4}
\end{equation*}
$$

Furthermore, $\omega_{x}^{a d p, c}(t, \eta)$ is $c$, the IF of $x(t)$. For a general $x(t)$ at $(t, \eta)$, we must then define the real part of the quantity of $\omega_{x}^{a d p, c}(t, \eta)$ in the above equation. This real part denoted by $\omega_{x}^{a d p}(t, \eta)$ and as the phase transformation of the adaptive FSST (see[15]):
$\omega_{x}^{a d p}(t, \eta)=\operatorname{Re}\left\{\frac{\partial_{t}\left(\widetilde{V}_{x}(t, \eta)\right)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \operatorname{Re}\left\{\frac{\widetilde{V}_{x}^{g_{3}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}\right\}, \quad$ for $\quad \widetilde{V}_{x}(t, \eta) \neq(0.6 .5)$
Then, the 1st-order adaptive FSST, or $R_{x, \gamma}^{a d p, \lambda}$, is defined by

$$
\begin{equation*}
R_{x, \gamma}^{a d p, \lambda}(t, \xi)=\int_{\left|\widetilde{V}_{x}(t, \eta)\right|>\gamma} \tilde{V}_{x}(t, \eta) \frac{1}{\lambda} h\left(\frac{\xi-\omega_{x}^{a d p}(t, \eta)}{\lambda}\right) d \eta \tag{2.6.6}
\end{equation*}
$$

We must now consider the 2nd-order adaptive FSST. For a linear chirp signal,

$$
\begin{equation*}
x(t)=A e^{i 2 \pi \phi(t)}=A e^{i 2 \pi\left(c t+\frac{1}{2} r t^{2}\right)} . \tag{2.6.7}
\end{equation*}
$$

Adaptive short-time Fourier transform and synchrosqueezing transform for non-stationary signal separation [15] shows $\omega_{x}^{\text {adp,2nd,c }}$, defined below as $c+r t$, as the IF of $x(t)$ :
$\omega_{x}^{a d p, 2 n d, c}=\frac{\sigma^{\prime}(t)}{i 2 \pi \sigma(t)}+\frac{\frac{\partial}{\partial t} \widetilde{V}_{x}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}-\frac{\widetilde{V}_{x}^{g_{1}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)} P_{0}(t, \eta)+\frac{\sigma^{\prime}(t)}{\sigma(t)} \frac{\widetilde{V}_{x}^{g_{3}}(t, \eta)}{i 2 \pi \widetilde{V}_{x}(t, \eta)}(2,6.8)$
for $(t, \eta)$ satisfying $\frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{x}^{g_{1}(t, \eta)}}{\widetilde{V}_{x}(t, \eta)}\right) \neq 0$ and $\widetilde{V}_{x}(t, \eta) \neq 0$, where

$$
\begin{equation*}
P_{0}(t, \eta)=\frac{1}{\frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{x}^{g_{1}}(t, \eta}{\widetilde{V}_{x}(t, \eta)}\right)}\left\{\frac{\partial}{\partial \eta}\left(\frac{\frac{\partial}{\partial t} \widetilde{V}_{x}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right)+\frac{\sigma^{\prime}(t)}{\sigma(t)} \frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{x}^{g_{3}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right)\right\} \tag{2.6.9}
\end{equation*}
$$

Li, Cai, Han, Jiang, Ji define the real part of $\omega_{x}^{a d p, 2 n d, c}$ as the phase transformation for the 2nd-order adaptive FSST (see [15]). The phase transformation $\omega_{x}^{a d p, 2 n d, c}$ is defined by

The adaptive FSST with $\widetilde{V}_{x}(t, \eta) \neq 0$ and $\frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{x}^{g_{1}}}{\widetilde{V}_{x}(t, \eta)}\right) \neq 0$ described by thresholds $\gamma_{1}>0, \gamma_{2}>0$. More precisely:
$\omega_{x \gamma_{1}, \gamma_{2}}^{\text {adn }, c}(t, \eta)=\left\{\begin{array}{lll}\text { quantity } & \text { in }(2.6 .8), & \text { if }\left|\widetilde{V}_{x}(t, \eta)\right|>\gamma_{1} \\ \text { and } \\ \text { quantity } & \text { in }(2.6 .4), & \text { if }\left|\widetilde{V}_{x}(t, \eta)\right|>\gamma_{1}\end{array} \quad\right.$ and $\left\lvert\, \begin{array}{ll}\left.\frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{x}^{g_{1}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right) \right\rvert\,>\gamma_{2} \\ \left.\frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{x}^{g_{1}}(t, \eta)}{\widetilde{V}_{x}(t, \eta)}\right) \right\rvert\, \leq \gamma_{2}\end{array}\right.$

Again, let $h(t)$ be a compactly supported function with certain smoothness and $\int_{\mathbb{R}} h(t) d t=1$.
The 2nd-order adaptive FSST, $R_{x \gamma_{1}, \gamma_{2}}^{a d p, \lambda, \lambda}$, is defined as

### 2.6.2 Analysis of adaptive FSST

The analysis of adaptive FSST was studied in [19]. A sinusoidal signal locally approximates each component of $x_{k}(t)=A_{k}(t) e^{i 2 \pi \phi_{k}(t)}$. We assume $A_{k}^{\prime}(t)$ and $\phi_{k}^{\prime}(t)$ are small.

$$
\begin{equation*}
\left|A_{k}^{\prime}(t)\right| \leq \varepsilon_{1}, \quad\left|\phi_{k}^{\prime \prime}(t)\right| \leq \varepsilon_{2}, \quad t \in \mathbb{R}, \quad 1 \leq k \leq K \tag{2.6.1}
\end{equation*}
$$

$$
\begin{equation*}
\tilde{V}_{x}(t, \eta)=\int_{\mathbb{R}} x(\tau) g(\tau-t) e^{-i 2 \pi \eta(\tau-t)} d \tau \tag{2.6.2}
\end{equation*}
$$

for some positive number $\varepsilon_{1}, \varepsilon_{2}$. Let $D_{\varepsilon_{1}, \varepsilon_{2}}$ denote the set of the set of multicomponent signals.

Let $x(t) \in D_{\varepsilon_{1}, \varepsilon_{2}}$. We write $x_{k}(t+\tau)$ as

$$
\begin{aligned}
x_{k}(t+\tau)= & x_{k}(t) e^{i 2 \pi \phi_{k}^{\prime}(t) \tau}+\left(A_{k}(t+\tau)-A_{k}(t)\right) e^{i 2 \pi \phi_{k}(t+\tau)} \\
& +x_{k}(t) e^{i 2 \pi \phi_{k}^{\prime}(t) \tau}\left(e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\phi_{k}^{\prime}(t) \tau\right)}-1\right) .
\end{aligned}
$$

If we simplify the expansion, we obtain

$$
x_{k}(t+\tau)=\left(A_{k}(t+\tau)-A_{k}(t)\right) e^{i 2 \pi \phi_{k}(t+\tau)}+x_{k}(t) e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)\right)} .
$$

Then, we have

$$
\begin{align*}
\widetilde{V}_{x}(t, \eta) & =\sum_{k=1}^{K} \int_{\mathbb{R}} x_{k}(t+\tau) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& =\sum_{k=1}^{K} \int_{\mathbb{R}} x_{k}(t) e^{i 2 \pi \phi_{k}^{\prime}(t) \tau} \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau+\text { rem }_{0} \\
\widetilde{V}_{x}(t, \eta) & =\sum_{k=1}^{K} x_{k}(t) \int_{\mathbb{R}} e^{i 2 \pi \phi_{k}^{\prime}(t) \tau} \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau+\operatorname{rem}_{0}  \tag{2.6.3}\\
& =\sum_{k=1}^{K} x_{k}(t) \hat{g}\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right)\right)+r e m_{0} \tag{2.6.4}
\end{align*}
$$

where rem $_{0}$ is the remainder for the expansion of $V_{x}(t, \eta)$ in (2.6.4) given by

$$
\begin{aligned}
\operatorname{rem}_{0}= & \sum_{k=1}^{K} \int_{\mathbb{R}}\left(A_{k}(t+\tau)-A_{k}(t)\right) e^{i 2 \pi \phi_{k}(t+\tau)} \cdot \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& +\sum_{k=1}^{K} \int_{\mathbb{R}} x_{k}(t) e^{i 2 \pi \phi_{k}^{\prime}(t) \tau}\left(e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\phi_{k}^{\prime}(t) \tau\right)}-1\right) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau
\end{aligned}
$$

with $\left|A_{k}(t+\tau)-A_{k}(t)\right| \leq \varepsilon_{1}|\tau|$ and
$\left|e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\phi_{k}^{\prime}(t) \tau\right)}-1\right| \leq 2 \pi \mid\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\left.\phi_{k}^{\prime}(t) \tau\left|\leq \pi \varepsilon_{2}\right| \tau\right|^{2}\right.$,
we have

$$
\begin{gathered}
\left|r e m_{0}\right| \leq \sum_{k=1}^{K} \int_{\mathbb{R}} \varepsilon_{1}|\tau| \frac{1}{\sigma(t)}\left|g\left(\frac{\tau}{\sigma(t)}\right)\right| d \tau+\sum_{k=1}^{K} A_{k}(t) \int_{\mathbb{R}} \pi \varepsilon_{2}|\tau|^{2} \frac{1}{\sigma(t)}\left|g\left(\frac{\tau}{\sigma(t)}\right)\right| d \tau \\
=K \varepsilon_{1} I_{1} \sigma(t)+\pi \varepsilon_{2} I_{2} \sigma^{2}(t) \sum_{k=1}^{K} A_{k}(t)
\end{gathered}
$$

where $I_{n}$ is defined in (2.5.11). Hence we have

$$
\begin{equation*}
\left|r e m_{0}\right| \leq \sigma(t) \Lambda_{0}(t) \tag{2.6.5}
\end{equation*}
$$

where

$$
\begin{equation*}
\Lambda_{0}(t)=K \varepsilon_{1} I_{1}+\pi \varepsilon_{2} I_{2} \sigma(t) \sum_{k=1}^{K} A_{k}(t) \tag{2.6.6}
\end{equation*}
$$

We can extend $\widetilde{V}_{x}^{g^{\prime}}(t, \eta)$ as (2.6.4) with remainder rem ${ }_{0}^{\prime}$. In (2.6.5), rem ${ }_{0}^{\prime}$ is defined as $r e m_{0}$ in (2.6.5) with $g(\tau)$ replaced by $g^{\prime}(\tau)$. Then we have the estimate for the remainder similar to (2.6.5). More precisely, we have

$$
\begin{equation*}
\left|r e m_{0}^{\prime}\right| \leq \sigma(t) \widetilde{\Lambda}_{0}(t) \tag{2.6.7}
\end{equation*}
$$

where

$$
\begin{equation*}
\widetilde{\Lambda}_{0}(t)=K \varepsilon_{1} \widetilde{I}_{1}+\pi \varepsilon_{2} \widetilde{I}_{2} \sigma(t) \sum_{k=1}^{K} A_{k}(t) \tag{2.6.8}
\end{equation*}
$$

with $\widetilde{I}_{n}$ defined in (2.5.11).
If the remainder $\operatorname{rem}_{0}$ in (2.6.4) is small, then the term $x_{k}(t) \hat{g}\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right)\right.$ in (2.6.4) gives the time-frequency zone of the STFT $\widetilde{V}_{x_{k}}$ of the $k$ th component $x_{k}(t)$ of $x(t)$. Additionally, if g is band-limited, that is $\hat{g}$ is compactly supported. Therefore, if $(\hat{g}) \subset[-\triangle, \triangle]$, then $x_{k}(t) \hat{g}\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right)\right.$ lies within the zone:

$$
\left\{(t, \eta):\left|\eta-\phi_{k}^{\prime}(t)\right|<\frac{\triangle}{\sigma(t)}, t \in \mathbb{R}\right\} .
$$

The multicomponent signal $x(t)$ is well-separated (that is $Z_{k} \cap Z_{l}=\varnothing, k \neq l$ ), provided that $\sigma(t)$ satisfies

$$
\begin{equation*}
\sigma(t) \geq \frac{2 \triangle}{\phi_{k}^{\prime}(t)-\phi_{k-1}^{\prime}(t)}, t \in \mathbb{R}, k=2, \ldots, K \tag{2.6.9}
\end{equation*}
$$

If $\hat{g}$ is not compactly supported, we must define the"support" of $\hat{g}$ outside of $\hat{g}(\xi) \approx 0$. Specifically, for a given threshold $0<\tau_{0}<1$, if $|\hat{g}(\xi)| \leq \tau_{0}$ for $|X| \geq \alpha$, then we say $\hat{g}(\xi)$ is "supported" in $[-\alpha, \alpha]$. When $|\hat{g}(\xi)|$ is even and decreasing for $\xi \geq 0$, then $\alpha$ can be obtained by solving

$$
\begin{equation*}
|\hat{g}(\alpha)|=\tau_{0} . \tag{2.6.10}
\end{equation*}
$$

For instance, when $g$ is the Gaussian function with $\hat{g}(\xi)=e^{-2 \pi^{2} \xi^{2}}$, the corresponding $\alpha$ is given by

$$
\begin{equation*}
\alpha=\frac{1}{2 \pi} \sqrt{2 \ln \left(\frac{1}{\tau_{0}}\right)} . \tag{2.6.11}
\end{equation*}
$$

Thus, $g$ with $\hat{g}(\xi)$ is "supported" in $[-\alpha, \alpha]$, and we define the time-frequency zone $Z_{k}$ of the $k$ th-component $x_{k}(t)$ of $x(t)$ by
$Z_{k}=\left\{(t, \eta): \left\lvert\, \hat{g}\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right) \mid>\tau_{0}, t \in \mathbb{R}\right\}=\left\{(t, \eta):\left|\eta-\phi_{k}^{\prime}(t)\right|<\frac{\alpha}{\sigma(t)}, t \in\{\mathbb{2}\} 6.12\right)\right.\right.$

If $\sigma(t)$ satisfies

$$
\begin{equation*}
\sigma(t) \geq \frac{2 \triangle}{\phi_{k}^{\prime}(t)-\phi_{k-1}^{\prime}(t)}, t \in \mathbb{R}, k=2, \ldots, K \tag{2.6.13}
\end{equation*}
$$

the multicomponent signal $x(t)$ is well-separated.
In this case $Z_{k} \cap Z_{l}=\varnothing, k \neq l$. When $\sigma(t)$ satisfies (2.6.13), since $\phi_{k}^{\prime}(t)$ is bounded, we can say

$$
\left\|\frac{1}{\sigma(t)}\right\|_{\infty}
$$

In this case, we may also say

$$
\begin{equation*}
\left.\sigma(t) \mid \phi_{k}^{\prime}(t)-\phi_{l}^{\prime} t\right)|\geq 2 \sigma| k-l \mid \tag{2.6.14}
\end{equation*}
$$

Theorem 2.6.1. [19] Let $x(t) \in D_{\varepsilon_{1}, \varepsilon_{2}}$ for small $\varepsilon_{1}, \varepsilon_{2}>0$. Then we have the following
(a) Suppose $\widetilde{\varepsilon_{1}}$ satisfies $\widetilde{\varepsilon_{1}} \geq \sigma(t) \Lambda_{0}(t)+\tau_{0} \sum_{k=1}^{K} A_{k}(t)$. Then for $(t, \eta)$ with $\left|\widetilde{V}_{x}(t, \eta)\right|>\widetilde{\varepsilon_{1}}$, there exists $k \in\{1,2,3, . . K\}$ such that $(t, \eta) \in Z_{k}$
(b) For $(t, \eta)$ with $\left|\widetilde{V}_{x}(t, \eta)\right| \neq 0$, we have

$$
\omega_{x}^{a p d, c}(t, \eta)-\phi_{k}^{\prime}(t)=\frac{R e m_{1}}{i 2 \pi \widetilde{V}_{x}(t, \eta)}
$$

where

$$
\operatorname{Rem}_{1}=i 2 \pi\left(\eta-\phi_{k}^{\prime}(t)\right) r e m_{0}-\frac{r e m_{0}^{\prime}}{\sigma(t)}+i 2 \pi \sum_{k \neq l} x_{l}(t)\left(\phi_{l}^{\prime}(t)-\phi_{k}^{\prime}(t)\right) \hat{g}\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right)\right)
$$

Hence, for $(t, \eta)$ satisfying $\left|\widetilde{V}_{x}(t, \eta)\right|>\widetilde{\varepsilon_{1}}$ and $(t, \eta) \in Z_{k}$, we have

$$
\begin{gathered}
\left|\omega_{x}^{a p d, c}(t, \eta)-\phi_{k}^{\prime}(t)\right|<b d_{1} \\
b d_{1}=\frac{1}{\widetilde{\varepsilon_{1}}}\left(\alpha \Lambda_{0}(t)+\frac{1}{2 \pi} \widetilde{\Lambda}_{0}(t)\right) \\
\\
\quad+\frac{1}{\widetilde{\varepsilon_{1}}} \max _{k \in\{1, \ldots, K\}}\left\{\sum_{k \neq l} A_{l}(t)\left|\phi_{l}^{\prime}(t)-\phi_{k}^{\prime}(t)\right| \sup _{u:|u|<\alpha}\left|\hat{g}\left(u+\sigma(t)\left(\phi_{k}^{\prime}(t)-\phi_{l}^{\prime}(t)\right)\right)\right|\right\}
\end{gathered}
$$

(c) Suppose that $\widetilde{\varepsilon_{1}}$ satisfies the condition in part (a) and that bd $d_{1}$ in part (b) satisfies $b d_{1} \leq \frac{\alpha}{\sigma(t)}$. Then for $\widetilde{\varepsilon_{3}}$ satisfying $b d_{1} \leq \widetilde{\varepsilon_{3}} \leq \frac{\alpha}{\sigma(t)}$ we have

$$
\begin{equation*}
\left|\lim _{\lambda \rightarrow 0} \frac{\sigma(t)}{g(0)} \int_{\left|\xi-\phi_{k}^{\prime}(t)\right|<\widetilde{\varepsilon_{3}}} R_{x, \widetilde{\varepsilon_{1}}}^{a d p, \lambda}(t, \xi) d \xi-x_{k}(t)\right| \leq b d_{2} \tag{2.6.15}
\end{equation*}
$$

where
$\left.b d_{2}=\frac{1}{|g(0)|}\left\{2 \alpha(\sigma(t)) \Lambda_{0}(t)+\widetilde{\varepsilon_{1}}\right)+A_{k}(t)\left|\int_{|u| \geq \alpha} \hat{g}(u) d u\right|+\sum_{l \neq k} A_{l}(t) m_{l, k}((t).) . \hat{\beta} .16\right)$
with

$$
m_{l, k}(t)=\left|\int_{|u| \geq \alpha} \hat{g}\left(u+\sigma(t)\left(\phi_{k}^{\prime}(t)-\phi_{l}^{\prime}(t)\right)\right) d u\right|
$$

### 2.6.3 Analysis of 2nd-order adaptive FSST

For a given $t, G_{k}(\xi)$ denotes the Fourier transform of $e^{i \pi \sigma(t) \phi_{k}^{\prime \prime}(t) \tau^{2}} g(\tau)$, namely (see [19]),

$$
\begin{gathered}
G_{k}(\xi)=\mathcal{F}\left(e^{i \pi \sigma(t) \phi_{k}^{\prime \prime}(t) \tau^{2}} g(\tau)\right)(\xi) \\
=\int_{\mathbb{R}} e^{i \pi \sigma(t) \phi_{k}^{\prime \prime}(t) \tau^{2}} g(\tau) e^{-i 2 \pi \xi \tau} d \tau \\
x_{k}(t+\tau)=x_{k}(t) e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}+\left(A_{k}(t+\tau)-A_{k}(t)\right) e^{i 2 \pi \phi_{k}(t+\tau)} \\
+x_{k}(t) e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}\left(e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\phi_{k}^{\prime}(t) \tau-\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}-1\right) .
\end{gathered}
$$

Then, we have

$$
\begin{aligned}
\widetilde{V}_{x}(t, \eta) & =\sum_{k=1}^{K} \int_{\mathbb{R}} x_{k}(t+\tau) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& =\sum_{k=1}^{K} \int_{\mathbb{R}} x_{k}(t) e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)} \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau+r e s_{0} .
\end{aligned}
$$

where

$$
\begin{gather*}
r e s_{0}=\sum_{k=1}^{K} \int_{\mathbb{R}}\left\{\left(A_{k}(t+\tau)-A_{k}(t)\right) e^{i 2 \pi \phi_{k}(t+\tau)}\right.  \tag{2.6.1}\\
\left.+x_{k}(t) e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}\left(e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\phi_{k}^{\prime}(t) \tau-\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}-1\right)\right\} \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i \nmid \widetilde{2} 2!6 d \not Z z)} \\
\widetilde{V}_{x}(t, \eta)=\sum_{k=1}^{K} x_{k}(t) G_{k}\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right)\right)+\text { res }_{0} \tag{2.6.3}
\end{gather*}
$$

with

$$
\left|A_{k}(t+\tau)-A_{k}(t)\right| \leq \varepsilon_{1}|\tau|
$$

and

$$
\left|e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\phi_{k}^{\prime}(t) \tau-\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}-1\right| \leq 2 \pi \frac{1}{6} \sup _{\eta \in \mathbb{R}}\left|\phi_{k}^{(3)}(\eta) \tau^{3}\right| \leq \frac{\pi}{3} \varepsilon_{2}|\tau|^{3}
$$

we have

$$
\begin{aligned}
\left|r e s_{0}\right| & \leq \sum_{k=1}^{K} \int_{\mathbb{R}} \varepsilon_{1}|\tau| \frac{1}{\sigma(t)}\left|g\left(\frac{\tau}{\sigma(t)}\right)\right| d \tau+\sum_{k=1}^{K} A_{k}(t) \int_{\mathbb{R}} \frac{\pi}{3} \varepsilon_{3}|\tau|^{3} \frac{1}{\sigma(t)}\left|g\left(\frac{\tau}{\sigma(t)}\right)\right| d \tau \\
& =K \varepsilon_{1} I_{1} \sigma^{3}(t) \sum_{k=1}^{K} A_{k}(t)
\end{aligned}
$$

where $I_{n}$ is defined in (2.5.11). Hence,

$$
\begin{equation*}
\left|r e s_{0}\right| \leq \sigma(t) \prod_{0}(t) \tag{2.6.4}
\end{equation*}
$$

where

$$
\begin{equation*}
\prod_{0}(t)=K \varepsilon_{1} I_{1}+\frac{\pi}{3} \varepsilon_{3} I_{3} \sigma^{2}(t) \sum_{k=1}^{K} A_{k}(t) . \tag{2.6.5}
\end{equation*}
$$

Therefore, if $\varepsilon_{1}, \varepsilon_{2}$ are small enough, then $\left|r e s_{0}\right|$ is small as well. $G_{k}(\sigma(t)(\eta-$ $\left.\left.\phi_{k}^{\prime}(t)\right)\right)$ also provides the time-frequency zone for $\widetilde{V}_{x_{k}}(t, \eta)$. To describe those time-frequency zones mathematically, let $0<\tau_{0}<1$ be a given small number and the threshold:

$$
\begin{equation*}
O_{k}=\left\{(t, \eta): \mid G_{k}\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right) \mid>\tau_{0}, t \in \mathbb{R}\right\}\right. \tag{2.6.6}
\end{equation*}
$$

Assuming $\left|G_{k}(\xi)\right|$ is even and decreasing for $\xi \geq 0$, then we may write $O_{k}$ as

$$
\begin{equation*}
O_{k}=\left\{(t, \eta):\left|\eta-\phi_{k}^{\prime}(t)\right|<\frac{\alpha_{k}}{\sigma(t)}, t \in \mathbb{R}\right\} \tag{2.6.7}
\end{equation*}
$$

where $\alpha_{k}=\alpha_{k}(t)$ is obtained by solving $\left|G_{k}(\xi)\right|=\tau_{0}$. In this instance, we will say that multicomponent signal $x(t)$ is well-separated and there is $\sigma(t)$ such that

$$
O_{k} \cap O_{l}=\emptyset, k \neq l
$$

We can use a Gaussian function defined in (2.5.24) as an example. For this g , we can obtain (see [15]),

$$
G_{k}(u)=\frac{1}{\sqrt{1-i 2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)}} e^{\left.-\frac{2 \pi^{2} u^{2}}{1+\left(2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)^{2}}\left(1+i 2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)\right)}
$$

Thus,

$$
\left|G_{k}(u)\right|=\frac{1}{\left(1+\left(2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)^{2}\right)^{\frac{1}{4}}} e^{-\frac{2 \pi^{2}}{\left.1+\left(2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)^{2}\right)}} u^{2}
$$

The solution of $\left|G_{k}(u)=\tau_{0}\right| \Leftrightarrow e^{-\frac{2 \pi^{2} u^{2}}{1+\left(2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)^{2}}}=\tau_{0}\left(\left(1+\left(2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)^{2}\right)^{\frac{1}{4}}\right.$
Therefore, in this case, assume $\tau_{0}\left(\left(1+\left(2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)^{2}\right)^{\frac{1}{4}} \leq 1\right.$,

$$
\alpha_{k}=\sqrt{\left.1+\left(2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)^{2}\right)} \frac{1}{2 \pi} \sqrt{\left.2 \ln \left(\frac{1}{\tau_{0}}\right)-\frac{1}{2} \ln \left(1+\left(2 \pi \phi_{k}^{\prime \prime}(t) \sigma^{2}(t)\right)^{2}\right)\right)} .
$$

The main theorem on the 2nd-order adptive FSST can be written in more notations observe:

$$
\begin{align*}
G_{j, k}(t, \eta) & =\int_{\mathbb{R}} e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)} \frac{\tau^{j}}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau .  \tag{2.6.8}\\
& =\mathcal{F}\left(e^{i \pi\left(\phi_{k}^{\prime \prime}(t) \tau^{2}\right)} \tau^{j} g(\tau)\right)\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right)\right) . \tag{2.6.9}
\end{align*}
$$

Clearly,

$$
G_{0, k}(t, \eta)=G_{k}\left(\sigma\left(\eta-\phi_{k}^{\prime}(t)\right)\right),
$$

and, when $j \geq 1$, we can see that

$$
\begin{equation*}
G_{j, k}=\frac{1}{(-i 2 \pi)^{j}} G_{k}^{(j)}\left(\sigma(t)\left(\eta-\phi_{k}^{\prime}(t)\right)\right) \tag{2.6.10}
\end{equation*}
$$

Let res ${ }_{1}$, res ${ }_{2}$, res ${ }_{0}^{\prime}$, and $r e s_{1}^{\prime}$ be the residuals defined as res ${ }_{0}$ in (2.6.2) with $g(\tau)$ replaced respectively by $g_{1}(\tau), g_{2}(\tau), g^{\prime}(\tau)$, and $g_{3}(\tau)=\tau g^{\prime}(\tau)$. We therefore have the estimates for these residuals that are similar to (2.6.4).

$$
\left|r e s_{1}\right| \leq \sigma(t) \prod_{1}(t), \quad\left|r e s_{2}\right| \leq \sigma(t) \prod_{2}(t), \quad\left|r e s_{0}^{\prime}\right| \leq \sigma(t) \widetilde{\prod}_{0}(t),\left|r e s_{1}^{\prime}\right| \leq \sigma(t) \widetilde{\prod}_{1}(t)
$$

where

$$
\begin{array}{ll}
\prod_{1}(t)=K \varepsilon_{1} I_{2}+\frac{\pi}{3} \varepsilon_{3} I_{4} \sigma^{2}(t) \sum_{k=1}^{K} A_{k}(t), & \prod_{2}(t)=K \varepsilon_{1} I_{3}+\frac{\pi}{3} \varepsilon_{3} I_{5} \sigma^{2}(t) \sum_{k=1}^{K} A_{k}(t), \\
\widetilde{\prod}_{0}(t)=K \varepsilon_{1} \widetilde{I}_{1}+\frac{\pi}{3} \varepsilon_{3} \widetilde{I}_{3} \sigma^{2}(t) \sum_{k=1}^{K} A_{k}(t), & \widetilde{\prod}_{1}(t)=K \varepsilon_{1} \widetilde{I}_{2}+\frac{\pi}{3} \varepsilon_{3} \widetilde{I}_{4} \sigma^{2}(t) \sum_{k=1}^{K} A_{k}(t),
\end{array}
$$

with $I_{n}, \widetilde{I}_{n}$ defined in (2.5.11).
Denote:

$$
\begin{aligned}
B_{k}(t, \eta)=\sum_{l \neq k} x_{l}(t)\left(\phi_{l}^{\prime}(t)-\phi_{k}^{\prime}(t)\right) G_{0, l}(t, \eta), & D_{k}(t, \eta)=\sum_{l \neq k} x_{l}(t)\left(\phi_{l}^{\prime \prime}(t)-\phi_{k}^{\prime \prime}(t)\right) G_{1, l}(t, \eta), \\
E_{k}(t, \eta)=\sum_{l \neq k} x_{l}(t)\left(\phi_{l}^{\prime}(t)-\phi_{k}^{\prime}(t)\right) G_{1, l}(t, \eta), & F_{k}(t, \eta)=\sum_{l \neq k} x_{l}(t)\left(\phi_{l}^{\prime \prime}(t)-\phi_{k}^{\prime \prime}(t)\right) G_{2, l}(t, \eta)
\end{aligned}
$$

and

$$
\begin{align*}
\text { Res }_{1}= & i 2 \pi B_{k}(t, \eta)+i 2 \pi \sigma D_{k}(t, \eta)+i 2 \pi\left(\eta-\phi_{k}^{\prime}(t)\right) r e s_{0}-\frac{r e s^{\prime} 0}{\sigma(t)}-i 2 \pi \phi_{k}^{\prime \prime}(t) \sigma(t) r\left(\mathbf{z}_{\mathbf{1}} 6.11\right) \\
\text { Res }_{2} & =4 \pi^{2} \sigma(t) E_{k}(t, \eta)+4 \pi^{2} \sigma^{2}(t) F_{k}(t)  \tag{2.6.12}\\
& +i 2 \pi r e s_{0}+4 \pi^{2}\left(\eta-\phi_{k}^{\prime}(t)\right) \sigma(t) r e s_{1}+i 2 \pi r e s_{1}^{\prime}-4 \pi^{2} \phi_{k}^{\prime \prime}(t) \sigma^{2}((t 2) \gamma 6.12)
\end{align*}
$$

Lemma 2.6.2. [19] Let Res ${ }_{1}$ be the quantity defined by (2.6.11). Then,
$\left.\partial_{t} \widetilde{V}_{x}(t, \eta)=\left(i 2 \pi \phi_{k}^{\prime}(t)-\frac{\sigma^{\prime}(t)}{\sigma(t)}\right) \widetilde{V}_{x}(t, \eta)+i 2 \pi \phi_{k}^{\prime \prime}(t) \sigma(t) \widetilde{V}_{x}^{g 1}(t, \eta)-\frac{\sigma^{\prime}(t)}{\sigma(t)}\right) \widetilde{V}_{x}^{g_{3}}(t, \eta)+$ R(2s $\left.\varphi .14\right)$
Lemma 2.6.3. [19] For $(t, \eta)$ satisfying $\widetilde{V}_{x}(t, \eta) \neq 0$ and $\frac{\partial}{\partial \eta}\left(\frac{\widetilde{V}_{x}^{g_{1}}(t, \eta}{\widetilde{V}_{x}(t, \eta)}\right) \neq 0$, we have

$$
\begin{equation*}
P_{0}(t, \eta)=i 2 \pi \sigma \phi_{k}^{\prime \prime}(t)+\operatorname{Res}_{3} \tag{2.6.15}
\end{equation*}
$$

where

$$
\begin{equation*}
\operatorname{Res}_{3}=\frac{\widetilde{V}_{x}(t, \eta) \operatorname{Res}_{2}-\partial_{\eta} \widetilde{V}_{x}(t, \eta) \operatorname{Res}_{1}}{\widetilde{V}_{x}(t, \eta) \partial_{\eta} \widetilde{V}_{x}^{g_{1}}(t, \eta)-\widetilde{V}_{x}^{g_{1}}(t, \eta) \partial_{\eta} \widetilde{V}_{x}(t, \eta)}, \tag{2.6.16}
\end{equation*}
$$

with Res $1_{1}$ and Res $2_{2}$ defined by (2.6.11)and (2.6.13) respectively.

Theorem 2.6.4. [19] Suppose $x(t) \in D_{\varepsilon_{1}, \varepsilon_{2}}^{(2)}$ with a window function $g(t)$ for some small $\varepsilon_{1}, \varepsilon_{2}>0$. Then, we have the following:
(a) Suppose $\widetilde{\varepsilon_{1}}$ satisfies $\widetilde{\varepsilon_{1}} \geq \varepsilon_{0} \sum_{k=1}^{K} A_{k}(t)+\sigma(t) \Pi_{0}(t)$. Then for $(t, \eta)$ with $\left|\widetilde{V}_{x}(t, \eta)\right|>\widetilde{\varepsilon_{1}}$, there exists $k \in\{1,2, \ldots, K\}$ such that $(t, \eta) \in O_{k}$.
(b) Suppose $(t, \eta)$ satisfies $\left|V_{x}(t, \eta)\right|>\widetilde{\varepsilon_{1}},\left|\partial_{\eta}\left(\widetilde{V}_{x}^{g_{1}}(t, \eta) / \widetilde{V}_{x}(t, \eta)\right)\right|>\widetilde{\varepsilon}_{2}$, and $(t, \eta) \in O_{k}$. Then

$$
\begin{equation*}
\omega^{a p d, 2 n d, c}(t, \eta)-\phi x_{k}^{\prime}(t)=\operatorname{Res}_{4} \tag{2.6.17}
\end{equation*}
$$

where

$$
\begin{equation*}
R e s_{4}=\frac{R e s_{1}}{i 2 \pi \widetilde{V}_{x}(t, \eta)}-\frac{\widetilde{V}_{x}^{g_{1}}(t, \eta) R e s_{3}}{i 2 \pi \widetilde{V}_{x}(t, \eta)} . \tag{2.6.18}
\end{equation*}
$$

Furthermore,

$$
\begin{equation*}
\left|\omega^{a p d, 2 n d}(t, \eta)-\phi_{k}^{\prime}(t)\right|<B d_{1} \tag{2.6.19}
\end{equation*}
$$

where

$$
B d_{1}=\max _{1 \leq k \leq K} \sup _{\eta \in O_{k}}\left\{\frac{\left|R e s_{1}\right|}{2 \pi \widetilde{\varepsilon}_{1}}+\frac{1}{2 \pi \widetilde{\varepsilon}_{1}^{3} \widetilde{\varepsilon}_{2}}\left|\widetilde{V}_{x}^{g_{1}}(t, \eta)\right|\left(\left|\partial_{\eta} \widetilde{V}_{x}(t, \eta)\right|\left|\operatorname{Res}_{1}\right|+\widetilde{\varepsilon}_{1}\left|\operatorname{Res}_{2}\right|\right)\right\}
$$

(c) Suppose that $\varepsilon_{1}$ satisfies the condition in part (a) and $B d_{1} \leq \frac{1}{2} L_{k}(t)$, where

$$
\begin{equation*}
L_{k}(t)=\frac{1}{\sigma(t)} \min \left\{\alpha_{k}+\alpha_{k-1}, \alpha_{k}+\alpha_{k+1}\right\} . \tag{2.6.20}
\end{equation*}
$$

Then for any $\widetilde{\varepsilon}_{3}=\widetilde{\varepsilon}_{3}(t)>0$ satisfying $B d_{1} \leq \widetilde{\varepsilon}_{3} \leq \frac{1}{2} L_{k}(t)$,

$$
\begin{equation*}
\left|\lim _{\lambda \rightarrow 0} \frac{\sigma(t)}{g(0)} \int_{\left|\xi-\phi_{k}^{\prime}(t)\right|<\widetilde{\varepsilon}_{3}} R_{x, \tilde{\varepsilon}_{1}, \tilde{\varepsilon}_{2}}^{a d p, 2 n d, \lambda}(t, \xi) d \xi-x_{k}(t)\right| \leq B d_{2} \tag{2.6.21}
\end{equation*}
$$

where $B d_{2}=B d_{2}^{\prime}+B d_{2}^{\prime \prime}$ with

$$
\begin{aligned}
B d_{2}^{\prime} & =\frac{1}{|g(0)|}\left\{2 \alpha_{k}\left(\widetilde{\varepsilon}_{1}+\sigma(t) \Pi_{0}(t)\right)+A_{k}(t)\left|\int_{|u| \geq \alpha_{k}} G_{k}(u) d u\right|+\sum_{l \neq k} A_{l}(t) M_{l, k}(t)\right\}, \\
B d_{2}^{\prime \prime} & \left.=\frac{1}{|g(0)|}\left\{2 \Pi_{0}(t)\right)+A_{k}(t)\|g\|_{1}\left|Z_{t}\right|+\sum_{l \neq k} A_{l}(t) M_{l, k}(t)\right\}
\end{aligned}
$$

and $\left|Z_{t}\right|$ represents the Lebesgue measure of the set $Z_{t}$ :
$Z_{t}=\left\{\eta:(t, \eta) \in O_{k},\left|\widetilde{V}_{x}(t, \eta)\right|>\widetilde{\varepsilon}_{1},,\left|\partial\left(V_{x}^{g_{1}}(t, \eta) / \widetilde{V}_{x}(t, \eta)\right)\right| \leq \widetilde{\varepsilon}_{2}\right\}$.

## Chapter 3

## Higher-order SST

### 3.1 High-order Synchrosqueezing transform

### 3.1.1 The higher-order wavelet synchrosqueezing trans-

 form
## Higher-order WSST

Consider the Nth-order polynomial-phase signal

$$
\begin{equation*}
x(t)=A e^{i 2 \pi \phi(t)} \tag{3.1.1}
\end{equation*}
$$

with $\phi(t)=r_{1} t+\frac{1}{2} r_{2} t^{2}+\cdots+\frac{1}{N} r_{N} t^{N}$, and $\phi^{\prime}(t)=r_{1}+\cdots+r_{N} t^{N-1}$. We have $x^{\prime}(t)=i 2 \pi \phi^{\prime}(t) x(t)$ and

$$
\begin{equation*}
\phi^{\prime}(b+a t)=\phi^{\prime}(b)+\frac{\phi^{\prime \prime}(b) a t}{1!}+\cdots+\frac{\phi^{(N)}(b) a^{N-1} t^{N-1}}{(N-1)!} . \tag{3.1.2}
\end{equation*}
$$

We recall the CWT definition of the signal $x(t)$

$$
\begin{equation*}
W_{x}(a, b)=\int_{-\infty}^{\infty} x(b+a t) \overline{\psi(t)} d t . \tag{3.1.3}
\end{equation*}
$$

Thus, we have

$$
\begin{align*}
\partial_{b} W_{x}(a, b) & =\int_{-\infty}^{\infty} x^{\prime}(b+a t) \overline{\psi(t)} d t=\int_{-\infty}^{\infty} i 2 \pi \phi^{\prime}(b+a) x(b+a t) \overline{\psi(t)} d t \\
& =\int_{-\infty}^{\infty} i 2 \pi\left(r_{1}+r_{2}(b+a t)+\cdots+r_{N}(b+a t)^{N-1}\right) x(b+a t) \overline{\psi(t)} d t \\
& =i 2 \pi \int_{-\infty}^{\infty}\left(\phi^{\prime}(b)+\frac{\phi^{\prime \prime}(b) a t}{1!}+\cdots+\frac{\phi^{(N)}(b) a^{N-1} t^{N-1}}{(N-1)!}\right) \cdot x(b+a t) \overline{\psi(t)} d t  \tag{3.1.4}\\
& =i 2 \pi\left(\phi^{\prime}(b) W_{x}(b, a)+\frac{\phi^{(2)}(b) a}{1!} W_{x}^{\psi_{1}}(b, a)+\cdots+\frac{a^{N-1} \phi^{(N)}(b)}{(N-1)!} W_{x}^{\psi_{N-1}}(b, a)\right)
\end{align*}
$$

where

$$
\begin{equation*}
W_{x}^{\psi_{k}}(b, a)=\int_{-\infty}^{\infty} x(b+a t) t^{k} \overline{\psi(t)} d t \tag{3.1.5}
\end{equation*}
$$

We can then write $\frac{\partial}{\partial b} W_{x}(b, a)=\sum_{k=1}^{N} \frac{i 2 \pi a^{k-1}}{(k-1)!} \phi^{(k)}(b) W_{x}^{\psi_{k-1}}(b, a)$.
The goal is to determine $\phi^{(1)}(b)$ according to $W_{x}(b, a), W_{x}^{\psi_{1}}(b, a), \ldots, W_{x}^{\psi_{N-1}}(b, a)$

$$
\begin{equation*}
\frac{\partial_{b} W_{x}(b, a)}{i 2 \pi W_{x}(b, a)}=\phi^{(1)}(b)+\sum_{k=2}^{N} \frac{a^{k-1}}{(k-1)!} \frac{W_{x}^{\psi_{k-1}}(b, a)}{W_{x}(b, a)} \phi^{(k)}(b) . \tag{3.1.6}
\end{equation*}
$$

We can write the equation in the form of a scalar product:

$$
w_{x}(a, b)=\left[x_{1}(b), \ldots, x_{N}(b)\right]\left[1, V_{2,1}(b, a), \ldots, V_{N, 1}(b, a)\right]^{T}
$$

where $w_{x}(a, b)=\frac{\partial_{b} W_{x}(b, a)}{i 2 \pi W_{x}(b, a)}$ and $x_{k}(b)=\phi^{(k)}(b)$, for $k=1, \ldots, N$.

To solve the problem, we pass through successive derivatives of equation
(3.1.6) according to the variable $a$.

To get sequence $\left(x_{k}(b)\right)_{1 \leq k \leq N}$, we create up a system of $N$ equations with variables $x_{k}(b)$. Let us denote

$$
\begin{equation*}
y_{1}=V_{N} \cdot X_{N}^{T} \tag{3.1.7}
\end{equation*}
$$

with $V_{N}=\left[1, V_{2,1}(b, a), \ldots, V_{N, 1}(b, a)\right]$ and $X_{N}=\left[x_{1}(b), \ldots, x_{N}(b)\right]$.

Computing the partial derivatives:

$$
\begin{equation*}
y_{2}(b, a)=\frac{\partial a y_{1}(b, a)}{\partial a V_{2,1}(b, a)} \text { and } V_{k, 2}(b, a)=\frac{\partial a V_{k, 1}(b, a)}{\partial a V_{2,1}(b, a)}, \tag{3.1.8}
\end{equation*}
$$

which implies the following expression:

$$
\begin{align*}
& y_{2}(b, a)=\left[0,1, V_{3,2}(b, a), \ldots, V_{N, 2}(b, a)\right] X_{N}^{T}  \tag{3.1.9}\\
& V_{k, 2}(b, a)=\frac{\partial a V_{k, 1}(b, a)}{\partial a V_{2,1}(b, a)}, \text { for } k=3, \ldots, N .
\end{align*}
$$

To get the $j^{\text {th }}$ equation. We repeat the same process iteratively. We define the new parameter for the $A_{N}$ matrix for $j=2, \ldots, N$ and $k=j, \ldots, N$ by:

$$
\begin{equation*}
y_{j}(b, a)=\frac{\partial a y_{j-1}(b, a)}{\partial a V_{j, j-1}(b, a)}, \text { and } V_{k, j}(b, a)=\frac{\partial a V_{k, j-1}(b, a)}{\partial a V_{j, j-1}(b, a)} . \tag{3.1.10}
\end{equation*}
$$

Then,

$$
y_{j}(b, a)=\left[0,0, \ldots, 1, V_{j+1, j}, \ldots, V_{N, j}\right] X_{N}^{T}
$$

We group the $N$ equations and get a good linear system

$$
\left(\begin{array}{c}
y_{1}  \tag{3.1.11}\\
\vdots \\
y_{N}
\end{array}\right)=A_{N}\left(\begin{array}{c}
x_{1} \\
\vdots \\
x_{N}
\end{array}\right)
$$

where

$$
A_{N}=\left(\begin{array}{ccccc}
1 & V_{2,1} & \ldots & \ldots & V_{N, 1}  \tag{3.1.12}\\
0 & 1 & V_{3,2} & \ldots & V_{N, 2} \\
\vdots & & & & \\
0 & 0 & \ldots & 1 & V_{N, N-1} \\
0 & 0 & \ldots & 0 & 1
\end{array}\right)
$$

Since the $A_{N}$ is an upper triangular matrix with a nonzero diagonal, the solution of the linear system is given by

$$
\begin{align*}
& x_{N}(b)=y_{N}(b, a) \\
& x_{j}(b)=y_{j}(b, a)-\sum_{k=j+1}^{N} V_{k, j}(b, a) x_{k}(b), \text { for } j=N-1, \ldots, 1 . \tag{3.1.13}
\end{align*}
$$

We can write this idea in the form of an algorithm.
Determination of the Nth-order local phase transformation
Step 1. We define the matrix $A_{N}$ by (3.1.12) with $V_{k, j}$ obtained by the following formula:

- $V_{k, 1}(b, a)=\frac{a^{k-1}}{(k-1)!} \frac{W_{x}^{\psi_{k-1}}(b, a)}{W_{x}(b, a)}$, for $k=2, \ldots, N$
- and $V_{k, j}(b, a)=\frac{\partial_{a} V_{k, j-1}(b, a)}{\partial_{a} V_{j, j-1}(b, a)}$, for $j=2, \ldots, N, k=j, \ldots, N$
- The matrix $A_{N}$ is defined by :

$$
A_{N}=\left(\begin{array}{ccccc}
1 & V_{2,1} & \ldots & \ldots & V_{N, 1} \\
0 & 1 & V_{3,2} & \ldots & V_{N, 2} \\
\vdots & & & & \\
0 & 0 & \ldots & 1 & V_{N, N-1} \\
0 & 0 & \ldots & 0 & 1
\end{array}\right)
$$

Step 2. We compute $y_{1}, y_{2}, \ldots, y_{N}$ by

$$
\begin{aligned}
& y_{1}(b, a)=w_{x}(a, b)=\frac{\partial_{b} W_{x}(b, a)}{i 2 \pi W_{x}(b, a)} \\
& y_{j}(b, a)=\frac{\partial_{a} y_{j-1}(b, a)}{\partial_{a} V_{j, j-1}(b, a)}, \text { for } j=2, \ldots, N
\end{aligned}
$$

Step 3. We solve (3.1.11) and obtain

$$
\begin{aligned}
& x_{N}(b)=y_{N}(b, a) \\
& x_{j}(b)=y_{j}(b, a)-\sum_{k=j+1}^{N} V_{k, j}(b, a) x_{k}(b), \text { for } j=N-1, \ldots, 1
\end{aligned}
$$

Step 4. The Nth-order phase transformation $\omega_{x}^{N}$ is defined by

$$
\omega_{x}^{N}(a, b)= \begin{cases}w_{x}(b, a)-\sum_{k=2}^{N} V_{k, 1}(b, a) x_{k}(b), & \text { if } W_{x}(b, a) \neq 0 \text { and } \partial a V_{j, j-1}(b, a) \neq 0 \\ w_{x}(b, a), & \text { if } W_{x}(b, a) \neq 0\end{cases}
$$

## Adaptive higher-order WSST

For a signal with Nth-order polynomial-phase, we define adaptive higher order synchrosqueezing transform ${\widetilde{\omega_{x}}}^{N}$. We start with the CWT with a time-varying parameter $\sigma(t)$.
$\widetilde{W}_{x}(a, b)=\int_{-\infty}^{\infty} x(b+a t) \overline{\psi_{\sigma(b)}(t)} d t=\int_{-\infty}^{\infty} x(b+a t) \frac{1}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu \nmid \partial d t .14)}$

The partial derivative of $W(a, b)$ is given by

$$
\begin{aligned}
\frac{\partial}{\partial b} \widetilde{W}_{x}(a, b) & =\int_{-\infty}^{\infty} x^{\prime}(b+a t) \frac{1}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu t} d t \\
& +\int_{-\infty}^{\infty} x(b+a t) \frac{-\sigma^{\prime}(b)}{\sigma(b)^{2}} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu t} d t \\
& +\int_{-\infty}^{\infty} x(b+a t) \frac{-\sigma^{\prime}(b) t}{\sigma(b)^{3}} g^{\prime}\left(\frac{t}{\sigma(b)}\right) e^{-i 2 \pi \mu t} d t .
\end{aligned}
$$

We can simplify the expression

$$
\partial_{b} \widetilde{W}_{x}(a, b)=\int_{-\infty}^{\infty} x^{\prime}(b+a t) \frac{1}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu t} d t-\frac{\sigma^{\prime}(b)}{\sigma(b)} \widetilde{W}_{x}(a, b)-\frac{\sigma^{\prime}(b)}{\sigma(b)} \widetilde{W}_{x}^{\psi^{2}}(a, b)
$$

where

$$
\begin{aligned}
\widetilde{W}_{x}(a, b) & =\int_{-\infty}^{\infty} x(b+a t) \frac{1}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu t} d t \\
\widetilde{W}_{x}^{\psi^{2}}(a, b) & =\int_{-\infty}^{\infty} x(b+a t) \frac{t}{\sigma(b)^{2}} \overline{g^{\prime}\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu t} d t
\end{aligned}
$$

Observe that

$$
\partial_{b} \widetilde{W}_{x}(a, b)+\frac{\sigma^{\prime}(b)}{\sigma(b)}\left(\widetilde{W}_{x}(a, b)+\widetilde{W}_{x}^{\psi^{2}}(a, b)\right)=\int_{-\infty}^{\infty} x^{\prime}(b+a t) \frac{1}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu t} d t .
$$

we denote: $\widetilde{\beta}(a, b)=\partial_{b} \widetilde{W}_{x}(a, b)+\frac{\sigma^{\prime}(b)}{\sigma(b)}\left(\widetilde{W}_{x}(a, b)+\widetilde{W}_{x}^{\psi^{2}}(a, b)\right)$.
Therefore,

$$
\begin{equation*}
\widetilde{\beta}(a, b)=\int_{-\infty}^{\infty} x^{\prime}(b+a t) \frac{1}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \eta t} d t \tag{3.1.15}
\end{equation*}
$$

In addition for a signal $x$ defined in (3.1.1), $\widetilde{\beta}(a, b)$ satisfies

$$
\begin{aligned}
\widetilde{\beta}(a, b) & =\int_{-\infty}^{\infty} i 2 \pi\left(\phi^{\prime}(b)+\cdots+\frac{(a t)^{N-1}}{(N-1)!} \phi^{(N)}\right) x(b+a t) \frac{1}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu t} d t \\
& =i 2 \pi\left(\widetilde{W}_{x}(a, b) \phi^{\prime}(b)+\cdots+\frac{a^{N-1}}{(N-1)!} \widetilde{W}_{x}^{\psi_{N-1}}(a, b) \phi^{(N)}(b)\right)
\end{aligned}
$$

where for $k=1, \ldots, N-1$

$$
\widetilde{W}_{x}^{\psi_{k}}(a, b)=\int_{-\infty}^{\infty} x(b+a t) \frac{t^{k}}{\sigma(b)} \overline{g\left(\frac{t}{\sigma(b)}\right)} e^{-i 2 \pi \mu t} d t .
$$

Thus, we have

$$
\begin{gathered}
\frac{\widetilde{\beta}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}=\phi^{\prime}(b)+a \frac{\widetilde{W}_{x}^{\psi_{1}}(a, b)}{\tilde{W}_{x}(a, b)} \phi^{(2)}(b)+\cdots+\frac{a^{N-1}}{(N-1)!} \frac{\widetilde{W}_{x}^{\psi_{N-1}}(a, b)}{\widetilde{W}_{x}(a, b)} \phi^{(N)}(b) \\
\phi^{\prime}(b)=\operatorname{Re}\left\{\frac{\widetilde{\beta}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}-\sum_{k=2}^{N} \phi^{(k)}(b) T_{k, 1}(a, b)\right\},
\end{gathered}
$$

with $T_{k, 1}(a, b)=\frac{\widetilde{W}_{x}^{\psi_{k-1}}(a, b)}{\widetilde{W}_{x}(a, b)} \frac{a^{k-1}}{(k-1)!}$. It remains only to determine $\phi^{(k)}(b)$ for $k=1, \ldots, N$

$$
\frac{\widetilde{\beta}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}=\phi^{\prime}(b)+\sum_{k=2}^{N} T_{k, 1}(a, b) \phi^{(k)}(b) .
$$

We denote

$$
\begin{aligned}
& y_{1}(a, b)=\frac{\widetilde{\beta}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}=\frac{\sigma(b) \partial_{b} \widetilde{W}_{x}(a, b)+\sigma^{\prime}(b)\left(\widetilde{W}_{x}(a, b)+\widetilde{W}_{x}^{\psi^{2}}(a, b)\right)}{i 2 \pi \sigma(b) \widetilde{W}_{x}(a, b)} \\
& y_{1}(a, b)=\chi_{1}(b)+\sum_{k=2}^{N} T_{k, 1}(a, b) \chi_{k}(b), \text { with } \chi_{k}(b)=\phi^{(k)}(b), k=1, \ldots, N .
\end{aligned}
$$

We can also put $y_{1}$ in the form of a scalar product defined as follows

$$
y_{1}(a, b)=\left[1, T_{2,1}(a, b), \ldots, T_{N, 1}(a, b)\right]\left(\begin{array}{c}
\chi_{1}(b) \\
\chi_{2}(b) \\
\vdots \\
\chi_{N}(b)
\end{array}\right)=<\left(\begin{array}{c}
1 \\
T_{2,1} \\
\vdots \\
T_{N, 1}
\end{array}\right),\left(\begin{array}{c}
\chi_{1}(b) \\
\chi_{2}(b) \\
\vdots \\
\chi_{N}(b)
\end{array}\right)>.
$$

Similarly, applying the algorithm (3.1.1) in the case of $\sigma$ is constant. The Nth-order phase transformation or the reference IF function estimate ${\widetilde{\omega_{x}}}^{N}$ is defined by
${\widetilde{\omega_{x}}}^{N}(a, b)= \begin{cases}\frac{\widetilde{\beta}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}-\sum_{k=2}^{N} T_{k, 1}(b, a) \chi_{k}(b), & \text { if } \widetilde{W}_{x}(b, a) \neq 0 \text { and } \partial_{a} T_{j, j-1}(b, a) \neq 0 \\ \frac{\widetilde{\beta}(a, b)}{i 2 \pi \widetilde{W}_{x}(a, b)}, & \text { if } \widetilde{W}_{x}(a, b) \neq 0\end{cases}$

### 3.1.2 The higher-order short time Fourier synchrosqueezing transform

## Higher-order FSST

Definition 3.1.1. Given a signal $x(\tau)=A(\tau) e^{i 2 \pi \phi(\tau)}$ in $L^{2}(\mathbb{R})$ with $A(\tau)$ and $\phi(\tau)$ are equal to their $L t h$-order and $N$-order respectively, the Taylor expansion for $\tau$ close to $t$ (see[17]):

$$
\begin{align*}
& A(\tau)=e^{\log (A(\tau))}=\exp \left(\sum_{k=0}^{L} \frac{(\log (A))^{(k)}(t)}{k!}(\tau-t)^{k}\right) \\
& \phi(\tau)=\sum_{k=0}^{N} \frac{\phi^{(k)}(t)}{k!}(\tau-t)^{k} . \tag{3.1.1}
\end{align*}
$$

The signal $x$ is defined as above, with $L \leq N$, can be written as:

$$
\begin{aligned}
x(\tau) & =\exp \left(\sum_{k=0}^{L} \frac{(\log (A))^{(k)}(t)}{k!}(\tau-t)^{k}\right) \exp \left(i 2 \pi \sum_{k=0}^{N} \frac{\phi^{(k)}(t)}{k!}(\tau-t)^{k}\right) \\
& =\exp \left(\sum_{k=0}^{N} \frac{1}{k!}\left([\log (A)]^{(k)}(t)+i 2 \pi \phi^{(k)}(t)\right)(\tau-t)^{k}\right)
\end{aligned}
$$

Since $(\log (t))^{(k)}(t)=0$ if $L+1 \leq K \leq N$, we define the STFT for a signal $x$ by:

$$
\begin{equation*}
V_{x}^{g}(t, \eta)=\int_{\mathbb{R}} x(\tau+t) g(\tau) e^{-i 2 \pi \eta \tau} d \tau \tag{3.1.2}
\end{equation*}
$$

Applying the derivative of STFT, we have:

$$
\begin{align*}
& \partial t V_{x}^{g}(t, \eta)=\int_{\mathbb{R}} \partial_{t}\left[\exp \left(\sum_{k=0}^{N} \frac{1}{k!}\left([\log (A)]^{(k)}(t)+i 2 \pi \phi^{(k)}(t)\right)(\tau)^{k}\right) g(\tau) e^{-i 2 \pi \eta \tau}\right] d \tau \\
&= \sum_{k=0}^{N} \frac{1}{k!}\left([\log (A)]^{(k+1)}(t)+i 2 \pi \phi^{(k+1)}(t)\right) V_{x}^{t^{k-1} g}(t, \eta)  \tag{3.1.3}\\
&= \sum_{k=1}^{N} \frac{1}{(k-1)!}\left(\frac{1}{i 2 \pi}[\log (A)]^{(k)}(t)+\phi^{(k)}(t)\right) \cdot i 2 \pi V_{x}^{t^{k-1} g}(t, \eta) \\
& w_{x}(t, \eta)=\frac{\partial_{t} V_{x}^{g}(t, \eta)}{2 i \pi V_{x}^{g}(t, \eta)}=\sum_{k=1}^{N} \frac{V_{x}^{t^{k-1} g}(t, \eta)}{V_{x}^{g}(t, \eta)} r_{k}(t) \\
&=\frac{(\log (A))^{\prime}(t)}{i 2 \pi}+\phi^{\prime}(t)+\sum_{k=2}^{N} \frac{V_{x}^{t^{k-1} g}(t, \eta)}{V_{x}^{g}(t, \eta)} r_{k}(t)
\end{align*}
$$

where $r_{k}(t)=\frac{1}{(k-1)!}\left(\frac{1}{i 2 \pi}[\log (A)]^{(k)}(t)+\phi^{(k)}(t)\right)$, for $k=1, \ldots, N$

$$
w_{x}(t, \eta)-\sum_{k=2}^{N} \frac{V_{x}^{t^{k-1} g}(t, \eta)}{V_{x}^{g}(t, \eta)} r_{k}(t)=\frac{(\log (A))^{\prime}(t)}{i 2 \pi}+\phi^{\prime}(t) .
$$

The Nth-order IF estimate follows:

$$
\begin{equation*}
\operatorname{Re}\left\{w_{x}(t, \eta)+\sum_{k=2}^{N} r_{k}(t)\left(-P_{k, 1}(t, \eta)\right)\right\}=\phi^{\prime}(t) \tag{3.1.6}
\end{equation*}
$$

where $P_{k, 1}(t, \eta)=\frac{V_{x}^{t^{k-1} g}(t, \tau)}{V_{x}^{g}(t, \tau)}$, for $k=2, \ldots, N$.
Moreover, we can put equation (3.1.5) in the form

$$
w_{x}(t, \eta)=\left[1, P_{2,1}(t, \eta), \ldots, P_{N, 1}(t, \eta)\right]\left(\begin{array}{c}
r_{1}(t) \\
r_{2}(t) \\
\vdots \\
r_{N}(t)
\end{array}\right)=P_{N} \cdot R_{N}^{T} .
$$

In the same way, we can present the previous cases to the algorithm 3.1.1 to provide the parameters $r_{1}, \ldots, r_{N}$.

Definition 3.1.2. If $s \in L^{2}(\mathbb{R})$, the Nth-order local complex IF estimate or phase transformation $w_{x}^{N}$ is defined by (see [17]):

$$
w_{x}^{N}(t, \eta)= \begin{cases}w_{x}(t, \eta)+\sum_{k=2}^{N} r_{k}(t)\left(-P_{k, 1}(t, \eta)\right) & \text { if } V_{x}^{g}(t, \eta) \neq 0 \text { and } \partial_{\eta} P_{j, j-1}(t, \eta) \neq q_{(3.1 .7)} \\ w_{x}(t, \eta) & \text { if } V_{x}^{g}(t, \eta) \neq 0 \text { and } \partial_{\eta} P_{j, j-1}(t, \eta) \neq 0\end{cases}
$$

## Adaptive higher-order FSST

We recall the STFT of the signal $x(t)$ denoted by

$$
\widetilde{V_{x}^{g}}(t, \eta)=\int_{\mathbb{R}} x(t+\tau) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau
$$

Taking the derivative by $t$

$$
\begin{aligned}
\frac{\partial}{\partial t} \widetilde{V_{x}^{g}}(t, \eta) & =\int_{\mathbb{R}} x^{\prime}(t+\tau) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& +\int_{\mathbb{R}} x(t+\tau)\left(\frac{-\sigma^{\prime}(t)}{\sigma^{2}(t)}\right) g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& +\int_{\mathbb{R}} x(t+\tau)\left(\frac{-\sigma^{\prime}(t)}{\sigma^{3}(t)} \tau\right) g^{\prime}\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau .
\end{aligned}
$$

To simplify the calculation, we note the following expression

$$
\begin{gather*}
\frac{\sigma^{\prime}(t)}{\sigma(t)} \widetilde{V_{x}^{g}}(t, \eta) \quad=\int_{\mathbb{R}} x(t+\tau)\left(\frac{-\sigma^{\prime}(t)}{\sigma^{2}(t)}\right) g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
\frac{-\sigma^{\prime}(t)}{\sigma(t)}{\widetilde{V_{x}}}^{g^{2}}(t, \eta)=\int_{\mathbb{R}} x(t+\tau)\left(\frac{-\sigma^{\prime}(t)}{\sigma^{3}(t)} \tau\right) g^{\prime}\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
\frac{\partial_{t} \widetilde{V_{x}^{g}}(t, \eta)}{\widetilde{V_{x}^{g}}(t, \eta)}=I-\frac{\sigma^{\prime}(t)}{\sigma(t)}-\frac{\sigma^{\prime}(t)}{\sigma(t)} \frac{\widetilde{V_{x}^{g^{g^{2}}}}(t, \eta)}{\widetilde{V_{x}^{g}}(t, \eta)} \tag{3.1.8}
\end{gather*}
$$

where

$$
\begin{aligned}
I & =\frac{1}{\widetilde{V_{x}^{g}}(t, \eta)}\left(\int_{\mathbb{R}} x^{\prime}(t+\tau) \frac{1}{\sigma(t)} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau\right) \\
& =\frac{i 2 \pi}{\widetilde{V_{x}^{g}}(t, \eta)} \cdot \sum_{k=1}^{N} \frac{1}{(k-1)!}\left(\frac{1}{i 2 \pi}[\log (A)]^{(k)}(t)+\phi^{(k)}(t)\right) \int_{\mathbb{R}} x(t+\tau) \tau^{k} g\left(\frac{\tau}{\sigma(t)}\right) e^{-i 2 \pi \eta \tau} d \tau \\
& =\frac{i 2 \pi}{\widetilde{V_{x}^{g}}(t, \eta)} \sum_{k=1}^{N} r_{k}(t){\widetilde{V_{x}}}^{t^{k} g}(\eta, t)
\end{aligned}
$$

with $r_{k}(t)=\frac{1}{(k-1)!}\left(\frac{1}{i 2 \pi}[\log (A)]^{(k)}(t)+\phi^{(k)}(t)\right)$, for $k=1, \ldots, N$.

Thus, we have

$$
\frac{1}{i 2 \pi}\left(\frac{\partial_{t} \widetilde{V_{x}^{g}}(t, \eta)}{\widetilde{V_{x}^{g}}(t, \eta)}+\frac{\sigma^{\prime}(t)}{\sigma(t)}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \frac{{\widetilde{V_{x}}}^{g^{2}}(t, \eta)}{\widetilde{V_{x}^{g}}(t, \eta)}\right)=\frac{\log ^{(1)}(A)(t)}{i 2 \pi}+\phi^{(1)}(t)+\sum_{k=2}^{N} r_{k}(t) \frac{{\widetilde{V_{x}}}^{t^{k} g}(\eta, t)}{\widetilde{V_{x}^{g}}(\eta, t)}
$$

we denote:

$$
\widetilde{w}_{x}(t, \eta)=\frac{1}{i 2 \pi}\left(\frac{\partial_{t} \widetilde{V_{x}^{g}}(t, \eta)}{\widetilde{V_{x}^{g}}(t, \eta)}+\frac{\sigma^{\prime}(t)}{\sigma(t)}+\frac{\sigma^{\prime}(t)}{\sigma(t)} \frac{{\widetilde{V_{x}}}^{g^{2}}(t, \eta)}{\widetilde{V_{x}^{g}}(t, \eta)}\right)
$$

Then,

$$
\begin{equation*}
\frac{\log (A)^{(1)}(t)}{i 2 \pi}+\phi^{(1)}(t)=\widetilde{w}_{x}(t, \eta)-\sum_{k=2}^{N} r_{k}(t) \frac{{\widetilde{V_{x}}}^{k} g(\eta, t)}{\widetilde{V_{x}^{g}}(\eta, t)} \tag{3.1.9}
\end{equation*}
$$

The IF function is obtained by

$$
\phi^{(1)}(t)=\operatorname{Re}\left\{\widetilde{w}_{x}(t, \eta)-\sum_{k=2}^{N} r_{k}(t) \frac{{\widetilde{V_{x}}}^{t^{k} g}(t, \eta)}{\widetilde{V_{x}^{g}}(t, \eta)}\right\} .
$$

We can use the equation (3.3.2) to apply the unknowns $r_{1}(t), \ldots, r_{N}(t)$ in the form of a scalar product previously solved by the algorithm (3.1.1):

$$
\widetilde{w}_{x}(t, \eta)=\left[1, S_{2,1}(t, \eta), \ldots, S_{N, 1}(t, \eta)\right]\left(\begin{array}{c}
r_{1}(t) \\
r_{2}(t) \\
\vdots \\
r_{N}(t)
\end{array}\right)=S_{N} \cdot R_{N}^{T}
$$

where $S_{k, 1}(t, \eta)=\frac{\widetilde{V_{x}{ }^{k} g}(t, \eta)}{\widetilde{V_{x}^{g}}(t, \eta)}$, for $k=2, \ldots, N$.
Let us denote $y_{1}(t, \eta)=\widetilde{w}_{x}(t, \eta), y_{j}(t, \eta)=\frac{\partial_{\eta} y_{j-1}(t, \eta)}{\partial \eta S_{j, j-1}(t, \eta)}$, and $S_{k, j}(t, \eta)=$ $\frac{\partial \eta S_{k, j-1}(t, \eta)}{\partial \eta S_{j, j-1}(t, \eta)}$. In the same way, algorithm (3.1.1) solves the problem.

### 3.2 New FSST transform

A new phase transformation for the 2nd-order adaptive FSST was proposed in [26]. We consider the new second order FSST associated with STFT. For a signal $x(t)$ defined by

$$
x(t)=A e^{p t+\frac{q}{2} t^{2}} e^{i 2 \pi\left(c t+\frac{1}{2} r t^{2}\right)}
$$

The STFT of $x(t) \in \mathrm{L}_{2}(\mathbb{R})$ with a window function $g(t) \in L_{2}(\mathbb{R})$ is defined as

$$
\begin{equation*}
V_{x}(t, \eta)=\int_{\mathbb{R}} x(\tau) g(\tau-t) e^{-i 2 \pi \eta(\tau-t)} d \tau \tag{3.2.1}
\end{equation*}
$$

where $t$ is the time variable and $\eta$ are is the frequency variable.
As in previous cases, we have

$$
\partial_{t} V_{x}(t, \eta)=(p+q t+i 2 \pi(c+r t)) V_{x}(t, \eta)+(q+i 2 \pi r) V_{x}^{g_{1}}(t, \eta)
$$

Thus at $(t, \eta)$ on which $V_{x}(t, \eta) \neq 0$

$$
\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}=(p+q t+i 2 \pi(c+r t)) \frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}+(q+i 2 \pi r)
$$

Taking partial derivative $\partial_{\eta}$, then we have

$$
\begin{equation*}
\frac{\partial}{\partial \eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)=(p+q t+i 2 \pi(c+r t)) P_{0}(t, \eta) \tag{3.2.2}
\end{equation*}
$$

where we use $P_{0}(t, \eta)$ to denote
$P_{0}(t, \eta)=\frac{\partial}{\partial \eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right) \Rightarrow(p+q t+i 2 \pi(c+r t))=\frac{1}{P_{0}(t, \eta)} \frac{\partial}{\partial \eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)$.
Thus,

$$
c+r t=-\frac{p+q t}{i 2 \pi}+\frac{1}{i 2 \pi P_{0}(t, \eta)} \frac{\partial}{\partial \eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right) .
$$

Thus for a general $x(t)$, we define a new phase transformation for the 2ndorder FSST, denoted by $\omega_{x}^{N e w, 2 n d}$, as

$$
\omega_{x}^{N e w, 2 n d}(t, \eta)= \begin{cases}\operatorname{Re}\left\{\frac{1}{\left.i 2 \pi \frac{\partial}{\partial \eta}\left(\frac{V_{x}(t, \eta)}{V_{V}^{x} 1,(t, \eta)}\right)^{\frac{\partial}{\partial \eta}}\left(\frac{V_{x}(t, \eta)}{V_{x}^{9} 1(t, \eta)}\right)\right\} .}\right. & \text { if } \frac{\partial}{\partial \eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{91}(t, \eta)}\right) \neq 0, \quad V_{x}(t, \eta) \neq 0 \\ \operatorname{Re}\left\{\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}\right\}, & \text { if } \frac{\partial}{\partial \eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{91}(t, \eta)}\right)=0, \quad V_{x}(t, \eta) \neq 0\end{cases}
$$

### 3.2.1 New higher-order FSST

Definition 3.2.1. Given a signal $x(\tau)=A(\tau) e^{i 2 \pi \phi(\tau)}$ in $L^{2}(\mathbb{R})$ with $A(\tau)$ and $\phi(\tau)$ are equal to their Lth-order and $N$-order respectively, the Taylor expansion for $\tau$ close to $t$ :

$$
\begin{align*}
& A(\tau)=e^{\log (A(\tau))}=\exp \left(\sum_{k=0}^{L} \frac{(\log (A))^{(k)}(t)}{k!}(\tau-t)^{k}\right)  \tag{3.2.1}\\
& \phi(\tau)=\sum_{k=0}^{N} \frac{\phi^{(k)}(t)}{k!}(\tau-t)^{k},
\end{align*}
$$

the signal $x$ is defined as above, with $L \leq N$, can be written as:

$$
\begin{aligned}
x(\tau) & =\exp \left(\sum_{k=0}^{L} \frac{(\log (A))^{(k)}(t)}{k!}(\tau-t)^{k}\right) \exp \left(i 2 \pi \sum_{k=0}^{N} \frac{\phi^{(k)}(t)}{k!}(\tau-t)^{k}\right) \\
& =\exp \left(\sum_{k=0}^{N} \frac{1}{k!}\left([\log (A)]^{(k)}(t)+i 2 \pi \phi^{(k)}(t)\right)(\tau-t)^{k}\right) .
\end{aligned}
$$

Since $(\log (t))^{(k)}(t)=0$ if $L+1 \leq K \leq N$. Next we define the STFT for a signal $x$ by:

$$
\begin{equation*}
V_{x}^{g}(t, \eta)=\int_{\mathbb{R}} x(\tau+t) g(\tau) e^{-i 2 \pi \eta \tau} d \tau \tag{3.2.2}
\end{equation*}
$$

Applying the derivative of STFT, we have:

$$
\begin{align*}
\partial_{t} V_{x}^{g}(t, \eta) & =\int_{\mathbb{R}} \partial_{t}\left[\exp \left(\sum_{k=0}^{N}\left(\frac{(\log (A))^{(k)}(t)}{k!}+i 2 \pi \phi^{(k)}(t)\right) \tau^{k}\right) g(\tau) e^{-i 2 \pi \eta \tau}\right] d \tau \\
& =\sum_{k=0}^{N}\left(\frac{(\log (A))^{(k+1)}(t)}{k!}+i 2 \pi \phi^{(k+1)}(t)\right) V_{x}^{t^{k-1} g}(t, \eta)  \tag{3.2.3}\\
& =\sum_{k=1}^{N}\left(\frac{(\log (A))^{(k)}(t)}{i 2 \pi(k-1)!}+\phi^{(k)}(t)\right) i 2 \pi V_{x}^{t^{k-1} g}(t, \eta) \\
w_{x}(t, \eta) & =\frac{\partial_{t} t_{x}^{g}(t, \eta)}{2 i \pi V_{x}^{t g}(t, \eta)}=\sum_{k=1}^{N} \frac{V_{x}^{t^{k-1} g}(t, \eta)}{V_{x}^{g}(t, \eta)} r_{k}(t) \\
& =\left(\frac{(\log (A))^{\prime}(t)}{i 2 \pi}+\phi^{\prime}(t)\right) \frac{V_{x}^{g}(t, \eta)}{V_{x}^{t g}(t, \eta)}+\frac{(\log (A))^{\prime \prime}(t)}{i 2 \pi}+\phi^{\prime \prime}(t)+\sum_{k=3}^{N} \frac{V_{x}^{t^{k-1} g}(t, \eta)}{V_{x}^{t g}(t, \eta)} r_{k}(t)
\end{align*}
$$

where $r_{k}(t)=\frac{(\log (A))^{(k)}(t)}{i 2 \pi(k-1)!}+\phi^{(k)}(t)$, for $k=1, \ldots, N$.
Taking the derivative by $\eta$ for the equation (3.2.4), we have:

$$
\partial_{\eta} w_{x}(t, \eta)=\left(\frac{(\log (A))^{\prime}(t)}{i 2 \pi}+\phi^{\prime}(t)\right) \partial_{\eta}\left(\frac{V_{x}^{g}(t, \eta)}{V_{x}^{t g}(t, \eta)}\right)+\sum_{k=3}^{N} \partial_{\eta}\left(\frac{V_{x}^{t^{k-1} g}(t, \eta)}{V_{x}^{t g}(t, \eta)}\right) r_{k}(t)
$$

Therefore, if in addition, $\partial_{\eta}\left(\frac{V_{x}^{g}(t, \eta)}{V_{x}^{g}(t, \eta)}\right) \neq 0$, then

$$
\left.w_{x}^{\text {New }}(t, \eta)=\left(\frac{(\log (A))^{\prime}(t)}{i 2 \pi}+\phi^{\prime}(t)\right)+\sum_{k=3}^{N} \frac{1}{\partial_{\eta}\left(\frac{V_{0}^{g}(t, \eta)}{V_{x}^{t g}(t, \eta)}\right)} \partial_{\eta}\left(\frac{V_{x}^{t^{k-1} g}(t, \eta)}{V_{x}^{t g}(t, \eta)}\right) r_{k}(B) 2.4\right)
$$

where the function $w_{x}^{\text {New }}$ is defined by

$$
\begin{equation*}
w_{x}^{N e w}(t, \eta)=\frac{1}{\partial_{\eta}\left(\frac{V_{x}^{g}(t, \eta)}{V_{x}^{t g}(t, \eta)}\right)} \partial_{\eta}\left(\frac{\partial_{t} V_{x}^{g}(t, \eta)}{2 i \pi V_{x}^{t g}(t, \eta)}\right) \tag{3.2.5}
\end{equation*}
$$

In addition, we can write equation (3.2.4) in the form

$$
w_{x}^{N e w}(t, \eta)-\sum_{k=3}^{N} W_{k, 1}(t, \eta) r_{k}(t)=\frac{(\log (A))^{\prime}(t)}{i 2 \pi}+\phi^{\prime}(t)
$$

where $W_{k, 1}(t, \eta)=\frac{1}{\partial_{\eta}\left(\frac{V_{t}^{g}(t, \eta)}{V_{x}^{t}(t, \eta)}\right)} \partial_{\eta}\left(\frac{V_{x}^{k^{k-1} g}(t, \eta)}{V_{x}^{t g}(t, \eta)}\right)$ for $k=3, \ldots, N$.
The new version for the Nth-order IF estimate is defined by the STFT:

$$
\begin{equation*}
\Re\left\{w_{x}^{N e w}(t, \eta)+\sum_{k=3}^{N} r_{k}(t)\left(-W_{k, 1}(t, \eta)\right)\right\}=\phi^{\prime}(t) . \tag{3.2.6}
\end{equation*}
$$

Moreover, we can put the equation (3.2.4) in the form:

$$
w_{x}^{N e w}(t, \eta)=\left[1, W_{3,1}(t, \eta), W_{4,1}(t, \eta), \ldots, W_{N, 1}(t, \eta)\right]\left(\begin{array}{c}
r_{1}(t) \\
r_{3}(t) \\
r_{4}(t) \\
\vdots \\
r_{N}(t)
\end{array}\right)=W_{N-1} \cdot R_{N-1}^{T}
$$

In the same way the previous cases used the algorithm (3.1.1) to provide the parameters $r_{1}, r_{3}, r_{4}, \ldots, r_{N}$, we can denote $k=3, \ldots, N$ by computing the partial derivatives:

$$
\begin{equation*}
y_{2}(t, \eta)=\frac{\partial_{\eta} \omega_{x}^{N e w}(t, \eta)}{\partial_{\eta} W_{3,1}(t, \eta)} \text { and } W_{k, 2}(t, \eta)=\frac{\partial_{\eta} W_{k, 1}(t, \eta)}{\partial_{\eta} W_{3,1}(t, \eta)} \tag{3.2.7}
\end{equation*}
$$

which implies the following expression

$$
\begin{equation*}
y_{2}(t, \eta)=\left[0,1, W_{4,2}(t, \eta), \ldots, W_{N, 2}(t, \eta)\right] R_{N-1}^{T} \tag{3.2.8}
\end{equation*}
$$

To find the $j^{\text {th }}$ equation, we do the same process iteratively. We define the new parameter for the $A_{N-1}$ matrix for $j=2, \ldots, N-1$ and $k=j+1, \ldots N-1$ by:

$$
\begin{equation*}
y_{j}(t, \eta)=\frac{\partial_{\eta} y_{j-1}(t, \eta)}{\partial_{\eta} W_{j+1, j-1}(t, \eta)}, \text { and } W_{k, j}(t, \eta)=\frac{\partial_{\eta} W_{k, j-1}(t, \eta)}{\partial_{\eta} W_{j+1, j-1}(t, \eta)} \tag{3.2.9}
\end{equation*}
$$

Then,

$$
y_{j}(t, \eta)=\left[0,0, \ldots, 1, W_{j+2, j}, \ldots, W_{N, j}\right] R_{N-1}^{T}
$$

We group the $N-1$ equations and get a good linear system:

$$
\begin{gather*}
\left(\begin{array}{c}
y_{1} \\
y_{2} \\
y_{3} \\
\vdots \\
y_{N-1}
\end{array}\right)=\left(\begin{array}{ccccc}
1 & W_{3,1} & \ldots & \ldots & W_{N, 1} \\
0 & 1 & W_{4,2} & \ldots & W_{N, 2} \\
\vdots & & & & \\
0 & 0 & \ldots & 1 & W_{N, N-2} \\
0 & 0 & \ldots & 0 & 1
\end{array}\right)\left(\begin{array}{c}
r_{1} \\
r_{3} \\
r_{4} \\
\vdots \\
r_{N}
\end{array}\right)  \tag{3.2.10}\\
Y_{N-1}=A_{N-1} \cdot\left[R_{N-1}\right]^{T} \tag{3.2.11}
\end{gather*}
$$

Since the $A_{N-1}$ is an upper triangular matrix with a nonzero diagonal, the solution of the linear system is given by

$$
\begin{align*}
& r_{N}(t)=y_{N-1}(t, \eta) \\
& r_{j}(t)=y_{j-1}(t, \eta)-\sum_{k=j+1}^{N} W_{k, j-2}(t, \eta) r_{k}(t), \text { for } j=N-1, \ldots, 3 \tag{3.2.12}
\end{align*}
$$

Definition 3.2.2. Let $x \in L^{2}(\mathbb{R})$. The New version for Nth-order local complex IF estimate or phase transformation $\omega_{x}^{N, N e w}$ is defined by
$\omega_{x}^{N, N e w}(t, \eta)= \begin{cases}w_{x}^{N e w}(t, \eta)+\sum_{k=3}^{N} r_{k}(t)\left(-W_{k, 1}(t, \eta)\right) & \text { if } \partial_{\eta} W_{j, j-1}(t, \eta) \neq 0 \\ w_{x}^{N e w}(t, \eta) & \text { if } \partial_{\eta} W_{j, j-1}(t, \eta)=0 .\end{cases}$

## Chapter 4

## Numerical simulation

### 4.1 Numerical Simulation

In this section, we present some experimental results for the new second order of the phase transformation $\omega_{x}^{2 n d}$. Let $x(t)$ be a signal with two linear chirps:

$$
\begin{equation*}
x(t)=x_{1}(t)+x_{2}(t)=\cos \left(2 \pi\left(c_{1}+\frac{1}{2} b_{1} t\right) t\right)+\cos \left(2 \pi\left(c_{2}+\frac{1}{2} b_{2} t\right) t\right), t \in[0,1] \tag{4.1.1}
\end{equation*}
$$

where the reference frequencies are $c_{1}=12, c_{2}=34$, and the chip rates are $b_{1}=50, b_{2}=64$. Here, $x(t)$ is sampled uniformly with $N=256$ sample points.


We proceed with two representative signal types.
Example 1: Our first signal is a signal with three components, given by $s(t)=s_{1}(t)+s_{2}(t)+s_{3}(t)$, where

$$
\begin{align*}
& s_{1}(t)=\cos \left(118 \pi\left(t-\frac{1}{2}\right)+100 \pi\left(t-\frac{1}{2}\right)^{2}\right) 1_{\left[\frac{1}{2}, 1\right]} . \\
& s_{2}(t)=\cos \left(94 \pi t+110 \pi t^{2}+13 \cos \left(4 \pi t-\frac{p i}{2}\right)\right) .  \tag{4.1.2}\\
& s_{3}(t)=\cos \left(194 \pi t+112 \pi t .^{2}\right) 1_{\left[0, \frac{3}{4}\right]} .
\end{align*}
$$



Figure 4.1: The signal $s(t)$ and its components $s_{1}(t), s_{2}(t)$ and $s_{3}(t)$

We use the relative "root mean square error" (RMSE) to evaluate the separation performance, which is defined by

$$
R M S E=\frac{1}{K} \sum_{k=1}^{K} \frac{\left\|z_{k}-\hat{z}_{k}\right\|_{2}}{\left\|z_{k}\right\|},
$$

where $\hat{z}_{k}$ is the reconstruction result of $z_{k}, K$ is the number of components.
Here, we test some parameters of $\sigma$ from 0.001 to 0.1 . The best value for the reconstrucation signal and its components from FSST2 can be obtained by minimizing the RMSE for both approaches of FSST2. Then, $\sigma_{\text {Old }}=$ $\sigma_{\text {New }}=0.023$. We colculated the following results.

| $\sigma$ | 0.015 | 0.020 | 0.023 | 0.025 | 0.030 |
| :--- | :---: | :---: | :---: | :---: | :---: |
| RMSE for Old | 0.1375 | 0.0860 | 0.0798 | 0.0843 | 0.0922 |
| RMSE for New | 0.1091 | 0.0720 | 0.0717 | 0.0782 | 0.0888 |

Table 4.1: Some differents values of $\sigma$ and their RMSE.


Figure 4.2: RMSE for FSST2 Old and New with $\sigma \in[0.001,0.04]$

As we can see, the best value for minimizing the value of RMSE is $\sigma \approx$ 0.023. This value shows the difference between the original IFs and the reconstrucation using the FSST2 old and FSST2 new.


Figure 4.3: The original IF of the signal $\mathrm{s}(\mathrm{t})$


Figure 4.4: Difference of reconstructed IFs with original IFs by old 2nd-order and new 2nd-order FSST.


Figure 4.5: Difference for the reconstructed $s_{1}, s_{2}, s_{3}$ with original component $s_{1}(t), s_{2}(t), s_{3}(t)$ by old 2nd-order and new 2nd-order FSST.

Example 2: The second signal is a signal with two components, given by the signal $s(t)=s_{1}(t)+s_{2}(t)$, defined by

$$
\begin{equation*}
s_{1}(t)=\cos \left(2 \pi\left(12 t+25 t^{2}\right)\right) \text { and } s_{2}(t)=\cos \left(2 \pi\left(34 t+32 t^{2}\right)\right) . \tag{4.1.3}
\end{equation*}
$$



Figure 4.6: The signal $s(t)$


Figure 4.7: The components of the signal $s(t)$ one by one $s_{1}(t)(\mathrm{left})$, and $s_{2}(t)$ (right)

Here, we test some parameters of $\sigma$ from 0.001 to 0.1 . The best value for the reconstrucation signal and its components from FSST2 can be obtained by minimizing the RMSE for both approaches of FSST2. Then, $\sigma_{\text {Old }}=$ $\sigma_{\text {New }} \approx 0.05$. We colculated the following results.

| $\sigma$ | 0.04 | 0.042 | 0.045 | 0.047 | 0.050 |
| :--- | :---: | :---: | :---: | :---: | :---: |
| RMSE for Old | 0.0893 | 0.0897 | 0.0802 | 0.0827 | 0.0824 |
| RMSE for New | 0.0893 | 0.0893 | 0.0800 | 0.0823 | 0.0822 |

Table 4.2: Some Examples for differents parameter of $\sigma$.


Figure 4.8: RMSE for Old and New FSST2 with $\sigma \in[0.001,0.1]$

As we can see, the best value for minimizing the value of RMSE is $\sigma \approx$ 0.045. This value demonstrates the difference between the original IFs and the reconstrucations using the FSST2 old and FSST2 new.


Figure 4.9: The original IF of the signal $\mathrm{s}(\mathrm{t})$


Figure 4.10: Difference of reconstructed IFs with original IFs by old 2ndorder and new 2nd-order FSST.


Figure 4.11: The difference for the signal $s(t)$


Figure 4.12: Difference for the reconstructed $s_{1}(t), s_{2}(t)$ with original component $s_{1}(t), s_{2}(t)$ by old 2nd-order and new 2nd-order FSST.

## Chapter 5

## Analysis of Adaptive Shor-time Fourier Transform-based Synchrosqueezing Transform

### 5.1 Analysis for new approach of FSST2

The Fourier transform of $e^{i \pi \phi_{k}^{\prime \prime}(t) \tau^{2}} g(\tau)$, which we denote by $G_{k}(\xi)$, is defined by (refer to [19]):

$$
\begin{aligned}
G_{k}(\xi) & =\mathcal{F}\left(e^{i \pi) \phi_{k}^{\prime \prime}(t) \tau^{2}} g(\tau)\right)(\xi) \\
& =\int_{\mathbb{R}} e^{i \pi \phi_{k}^{\prime \prime}(t) \tau^{2}} g(\tau) e^{-i 2 \pi \xi \tau} d \tau
\end{aligned}
$$

$$
\begin{aligned}
x_{k}(t+\tau) & =x_{k}(t) e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}+\left(A_{k}(t+\tau)-A_{k}(t)\right) e^{i 2 \pi \phi_{k}(t+\tau)} \\
& +x_{k}(t) e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}\left(e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\phi_{k}^{\prime}(t) \tau-\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}-1\right) .
\end{aligned}
$$

Then, we have

$$
\begin{aligned}
V_{x}(t, \eta) & =\sum_{k=1}^{K} \int_{\mathbb{R}} x_{k}(t+\tau) g(\tau) e^{-i 2 \pi \eta \tau} d \tau \\
& =\sum_{k=1}^{K} \int_{\mathbb{R}} x_{k}(t) e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)} g(\tau) e^{-i 2 \pi \eta \tau} d \tau+\operatorname{res}_{0}
\end{aligned}
$$

where

$$
\begin{align*}
& \operatorname{res}_{0}=\sum_{k=1}^{K} \int_{\mathbb{R}}\left\{\left(A_{k}(t+\tau)-A_{k}(t)\right) e^{i 2 \pi \phi_{k}(t+\tau)}\right. \\
& \left.+x_{k}(t) e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}\left(e^{i 2 \pi\left(\phi_{k}(t+\tau)-\phi_{k}(t)-\phi_{k}^{\prime}(t) \tau-\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)}-1\right)\right\} g(\tau) e^{-i 2 \pi \eta \tau} d \tau  \tag{Lbll.1}\\
& \left|r e s_{0}\right| \leq \prod_{0}(t)
\end{align*}
$$

where

$$
\prod_{0}(t)=K \varepsilon_{1} I_{1}+\frac{\pi}{3} \varepsilon_{3} I_{3} \sum_{k=1}^{K} A_{k}(t) .
$$

We introduce more notations defined as follows

$$
\begin{aligned}
G_{j, k}(t, \eta) & =\int_{\mathbb{R}} e^{i 2 \pi\left(\phi_{k}^{\prime}(t) \tau+\frac{1}{2} \phi_{k}^{\prime \prime}(t) \tau^{2}\right)} \tau^{j} g(\tau) e^{i 2 \pi \eta \tau} d \tau . \\
& =\mathcal{F}\left(e^{i \pi\left(\phi_{k}^{\prime \prime}(t) \tau^{2}\right)} \tau^{j} g(\tau)\right)\left(\eta-\phi_{k}^{\prime}(t)\right)
\end{aligned}
$$

We also note

$$
\begin{array}{r}
G_{k}\left(\eta-\phi_{k}^{\prime}(t)\right)=G_{0, k}(t, \eta) \\
G_{j, k}=\frac{1}{(-i 2 \pi)^{j}} G_{k}^{(j)}\left(\eta-\phi_{k}^{\prime}(t)\right)
\end{array}
$$

Lemma 5.1.1. [19] Let $x(t)=A(t) e^{i 2 \pi \phi(t)} \in L^{2}(\mathbb{R})$, we have

$$
\begin{equation*}
\partial_{t} V_{x}(t, \eta)=i 2 \pi \phi_{k}^{\prime}(t) V_{x}(t, \eta)+i 2 \pi \phi_{k}^{\prime \prime}(t) V_{x}^{g 1}(t, \eta)+\operatorname{Res}_{1} \tag{5.1.3}
\end{equation*}
$$

where
$\operatorname{Res}_{1}=i 2 \pi B_{k}(t, \eta)+i 2 \pi D_{k}(t, \eta)+i 2 \pi\left(\eta-\phi_{k}^{\prime}(t)\right) r e s_{0}-r e s_{0}^{\prime}-i 2 \pi \phi_{k}^{\prime}(t) r e s_{1}$.

Proof. $V_{x}^{g^{\prime}}$ is defined by (see [19]):

$$
V_{x}^{g^{\prime}}=i 2 \pi \sum_{\ell=1}^{K} x_{\ell}(t)\left(\eta-\phi_{\ell}^{\prime}(t)\right) G_{0, \ell}(t, \eta)-i 2 \pi \sum_{\ell=1}^{K} x_{\ell}(t) \phi_{\ell}^{\prime \prime}(t) G_{1, \ell}(t, \eta)+\operatorname{res}_{0}^{\prime}
$$

Taking the derivatve by the time variable $t$, we obtain

$$
\begin{equation*}
\partial_{t} V_{x}(t, \eta)=i 2 \pi \eta V_{x}(t, \eta)-V_{x}^{g^{\prime}}(t, \eta) \tag{5.1.4}
\end{equation*}
$$

Using the derivative of the time variable (5.1.4), we have

$$
\begin{aligned}
& \partial_{t} V_{x}(t, \eta)-i 2 \pi \phi_{k}^{\prime}(t) V_{x}(t, \eta)-i 2 \pi \phi_{k}^{\prime \prime}(t) V_{x}^{g 1}(t, \eta) \\
= & i 2 \pi\left(\eta-\phi_{k}^{\prime}(t)\right) V_{x}(t, \eta)-V_{x}^{g^{\prime}}(t, \eta)-i 2 \pi \phi_{k}^{\prime \prime}(t) V_{x}^{g 1}(t, \eta) \\
= & i 2 \pi\left(\eta-\phi_{k}^{\prime}(t)\right) V_{x}(t, \eta)-i 2 \pi \sum_{\ell=1}^{K} x_{\ell}(t)\left(\eta-\phi_{\ell}^{\prime}(t)\right) G_{0, \ell}(t, \eta)-r e s_{0}^{\prime} \\
+ & i 2 \pi \sum_{\ell=1}^{K} x_{\ell}(t) \phi_{\ell}^{\prime \prime}(t) G_{1, \ell}(t, \eta)-i 2 \pi \phi_{k}^{\prime \prime}(t) V_{x}^{g 1}(t, \eta)
\end{aligned}
$$

In addition, we define the expression of $V_{x}(t, \eta)$ and $V_{x}^{g 1}(t, \eta)$ in the form

$$
\begin{aligned}
& V_{x}(t, \eta)=\sum_{\ell=1}^{K} x_{\ell}(t) G_{0, \ell}(t, \eta)+\operatorname{res}_{0} \\
& V_{x}^{g 1}(t, \eta)=\sum_{\ell=1}^{K} x_{\ell}(t) G_{1, \ell}(t, \eta)+\text { res }_{1} .
\end{aligned}
$$

Then, we have

$$
\begin{aligned}
& \partial_{t} V_{x}(t, \eta)-i 2 \pi \phi_{k}^{\prime}(t) V_{x}(t, \eta)-i 2 \pi \phi_{k}^{\prime \prime}(t) V_{x}^{g 1}(t, \eta) \\
= & i 2 \pi\left(\eta-\phi_{k}^{\prime}(t)\right)\left(\sum_{\ell=1}^{K} x_{\ell}(t) G_{0, \ell}(t, \eta)+r e s_{0}\right)-i 2 \pi \sum_{\ell=1}^{K} x_{\ell}(t)\left(\eta-\phi_{\ell}^{\prime}(t)\right) G_{0, \ell}(t, \eta)-r e s_{0}^{\prime} \\
+ & i 2 \pi \sum_{\ell=1}^{K} x_{\ell}(t) \phi_{\ell}^{\prime \prime}(t) G_{1, \ell}(t, \eta)-i 2 \pi \phi_{k}^{\prime \prime}(t)\left(\sum_{\ell=1}^{K} x_{\ell}(t) G_{1, \ell}(t, \eta)+r e s_{1}\right) \\
= & i 2 \pi \sum_{\ell \neq k} x_{\ell}(t)\left(\phi_{\ell}^{\prime}(t)-\phi_{k}^{\prime}(t)\right) G_{0, \ell}(t, \eta)+i 2 \pi \sum_{\ell \neq k} x_{\ell}(t)\left(\phi_{\ell}^{\prime \prime}(t)-\phi_{k}^{\prime \prime}(t)\right) G_{1, \ell}(t, \eta) \\
+ & i 2 \pi\left(\eta-\phi_{k}^{\prime}(t)\right) r e s_{0}-r e s_{0}^{\prime}-i 2 \pi \phi_{k}^{\prime \prime}(t) r e s_{1} \\
= & i 2 \pi B_{k}(t, \eta)+i 2 \pi D_{k}(t, \eta)+i 2 \pi\left(\eta-\phi_{k}^{\prime}(t)\right) r e s_{0}-r e s_{0}^{\prime}-i 2 \pi \phi_{k}^{\prime \prime}(t) r e s_{1} \\
= & \operatorname{Res}_{1}
\end{aligned}
$$

where

$$
\begin{aligned}
B_{k}(t, \eta) & =\sum_{\ell \neq k} x_{\ell}(t)\left(\phi_{\ell}^{\prime}(t)-\phi_{k}^{\prime}(t)\right) G_{0, \ell}(t, \eta) \\
D_{k}(t, \eta) & =\sum_{\ell \neq k} x_{\ell}(t)\left(\phi_{\ell}^{\prime \prime}(t)-\phi_{k}^{\prime \prime}(t)\right) G_{1, \ell}(t, \eta)
\end{aligned}
$$

We recall the definition of the new phase transformation for the 2nd-order STFT, which $V_{x}(t, \eta) \neq 0$.:

$$
\omega_{x}^{n e w, 2 n d}(t, \eta)= \begin{cases}\operatorname{Re}\left\{\frac { 1 } { i 2 \pi \frac { \partial } { \partial \eta } ( \frac { V _ { x } ( t , \eta ) } { V _ { x } ^ { g _ { 1 } } ( t , \eta ) } ) } \frac { \partial } { \partial \eta } \left(\frac{\partial_{t} V_{x}(t, \eta)}{\left.\left.V_{x}^{g_{1}(t, \eta)}\right)\right\}}\right.\right. & \text { if } \frac{\partial}{\partial \eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{\left.V_{x}^{g_{1}(t, \eta)}\right) \neq 0, \quad V_{x}^{g_{1}}(t, \eta) \neq 0}\right. \\ \operatorname{Re}\left\{\frac{\partial_{t} V_{x}(t, \eta)}{i 2 \pi V_{x}(t, \eta)}\right\}, & \text { if } \frac{\partial}{\partial \eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)=0, \quad V_{x}^{g_{1}}(t, \eta) \neq 0 .\end{cases}
$$

Then, the new second order complex IF is defined as follows

$$
\omega_{x}^{N e w, 2 n d, c}(t, \eta)=\frac{P_{1}(t, \eta)}{i 2 \pi Q_{1}(t, \eta)}
$$

where $P_{1}$ and $Q_{1}$ are defined by

$$
\begin{equation*}
P_{1}(t, \eta)=\frac{\partial}{\partial \eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right) \text { and } Q_{1}(t, \eta)=\frac{\partial}{\partial \eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right) \tag{5.1.5}
\end{equation*}
$$

Lemma 5.1.2. For $(t, \eta)$ such that $V_{x}^{g_{1}}(t, \eta) \neq 0$ and $P_{1}(t, \eta) \neq 0$, as defined in the equation (5.1.5), we have

$$
\begin{equation*}
P_{1}(t, \eta)=i 2 \pi \phi_{k}^{\prime}(t) Q_{1}(t, \eta)+\operatorname{Res}_{3} \tag{5.1.6}
\end{equation*}
$$

and we note that $\operatorname{Res}_{2}=\partial_{\eta} \operatorname{Res}_{1}($ see $[19])$, where

$$
\operatorname{Res}_{3}=\frac{\operatorname{Res}_{2} V_{x}^{g_{1}}(t, \eta)-\partial_{\eta} V_{x}^{g_{1}}(t, \eta) \operatorname{Res}_{1}}{\left(V_{x}^{g_{1}}(t, \eta)\right)^{2}}
$$

Proof. Using the result of lemma (5.1.1), we have

$$
\begin{equation*}
\partial_{t} V_{x}(t, \eta)=i 2 \pi \phi_{k}^{\prime}(t) V_{x}(t, \eta)+i 2 \pi \phi_{k}^{\prime \prime}(t) V_{x}^{g 1}(t, \eta)+\operatorname{Res}_{1} . \tag{5.1.7}
\end{equation*}
$$

Therefore, in the equation (5.1.7), therefore, thus

$$
\begin{equation*}
\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}=i 2 \pi \phi_{k}^{\prime}(t) \frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}+i 2 \pi \phi_{k}^{\prime \prime}(t)+\frac{R e s_{1}}{V_{x}^{g_{1}}(t, \eta)} \tag{5.1.8}
\end{equation*}
$$

Taking the derivative by $\eta$, then yields

$$
\begin{equation*}
\partial_{\eta}\left(\frac{\partial_{t} V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)=i 2 \pi \phi_{k}^{\prime}(t) \partial_{\eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)+\partial_{\eta}\left(\frac{R e s_{1}}{V_{x}^{g_{1}}(t, \eta)}\right) . \tag{5.1.9}
\end{equation*}
$$

We replace in the equation (5.1.9) with the expression of the $P_{1}$ and $Q_{1}$ :

$$
\begin{equation*}
P_{1}(t, \eta)=i 2 \pi \phi_{k}^{\prime}(t) Q_{1}(t, \eta)+\operatorname{Res}_{3} \tag{5.1.10}
\end{equation*}
$$

This completes the proof of lemma (5.1.2).

Theorem 5.1.3. Let $x(t) \in D_{\varepsilon_{1}, \varepsilon_{2}}$ for small $\varepsilon_{1}, \varepsilon_{2}>0$. We have the following results:
(a)Suppose $\varepsilon_{1}$ satisfies $\varepsilon_{1} \geq \Pi_{0}(t)+\tau_{0} \sum_{k=1}^{K} c_{k} A_{k}(t)$, and for $(t, \eta)$ with $\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1}$. Then, there exists $k \in\{1, \ldots, K\}$ such that $(t, \eta) \in O_{k}$
(b) If $(t, \eta)$ such that $\left|V_{x}(t, \eta)\right|>\varepsilon_{1},\left|\partial_{\eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{G_{1}(t, \eta)}}\right)\right|>\varepsilon_{2}$ and $(t, \eta) \in O_{k}$, then, we have

$$
\omega_{x}^{\text {New }, 2 n d, c}(t, \eta)-\phi_{k}^{\prime}(t)=\operatorname{Res}_{4}
$$

where

$$
\operatorname{Res}_{4}=\frac{\operatorname{Res}_{2} V_{x}^{g_{1}}(t, \eta)-\partial_{\eta} V_{x}^{g_{1}}(t, \eta) \operatorname{Res}_{1}}{i 2 \pi\left(\partial_{\eta} V_{x}(t, \eta) V_{x}^{g 1}(t, \eta)-\partial_{\eta} V_{x}^{g 1}(t, \eta) V_{x}(t, \eta)\right)}
$$

In addition, we have

$$
\left|\omega^{N e w, 2 n d}(t, \eta)-\phi_{k}^{\prime}(t)\right|<B d_{1}
$$

where

$$
B d_{1}=\max _{1 \leq k \leq K} \sup _{\eta \in O_{k}}\left\{\frac{1}{2 \pi \varepsilon_{1}^{2} \varepsilon_{2}}\left(\left|\operatorname{Res}_{2}\right|\left|V_{x}^{g_{1}}(t, \eta)\right|+\left|\operatorname{Res}_{1}\right|\left|\partial_{\eta} V_{x}^{g_{1}}(t, \eta)\right|\right)\right\}
$$

(c) If $\varepsilon_{1}$ satisfies the condition in part (a) and $B d_{1} \leq \frac{1}{2} L_{k}(t)$, then

$$
L_{k}(t)=\min \left\{\alpha_{k}+\alpha_{k-1}, \alpha_{k}+\alpha_{k+1}\right\} .
$$

Then for any $\varepsilon_{3}=\varepsilon_{3}(t)>0$ satisfying $B d_{1} \leq \varepsilon_{3} \leq \frac{1}{2} L_{k}(t)$,

$$
\left|\lim _{\lambda \rightarrow 0} \frac{1}{g(0)} \int_{\left|\xi-\phi_{k}^{\prime}(t)\right|<\varepsilon_{3}} R_{x, \varepsilon_{1}, \varepsilon_{2}}^{2 n d, \lambda}(t, \xi) d \xi-x_{k}(t)\right| \leq B d_{2}
$$

where $B d_{2}=B d_{2}^{\prime}+B d_{2}^{\prime \prime}$ with

$$
\begin{aligned}
B d_{2}^{\prime} & =\frac{1}{|g(0)|}\left\{2 \alpha_{k}\left(\prod_{0}(t)+\varepsilon_{1}\right)+A_{k}(t)\left|\int_{|u| \geq \alpha_{k}} G_{k}(u) d u\right|+\sum_{l \neq k} A_{l}(t) M_{l, k}(t)\right\}, \\
B d_{2}^{\prime \prime} & =\frac{1}{|g(0)|}\left\{2 \prod_{0}(t)+A_{k}(t)\|g\|_{1}\left|Z_{t}\right|+\sum_{l \neq k} A_{l}(t) M_{l, k}(t)\right\}
\end{aligned}
$$

and with $\left|Z_{t}\right|$ represents the Lebesgue measure of the set $Z_{t}$ :

$$
Z_{t}:=\left\{\eta:(t, \eta) \in O_{k},\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1},\left|\partial_{\eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)\right| \leq \varepsilon_{2}\right\} .
$$

Proof Part (a). Assume $(t, \eta) \notin \cup_{k=1}^{K} O_{k}$. Then for any k, by the definition of $O_{k}$ in (2.6.6) with $\sigma=1$, we have $\left|G_{k}\left(\eta-\phi_{k}^{\prime}(t)\right)\right| \leq \tau_{0}$. Hence, by (5.1.1) and (5.1.2), we have

$$
\begin{equation*}
\left|G_{1, k}\left(\eta-\phi_{k}^{\prime}(t)\right)\right| \leq c_{k}\left|G_{k}\left(\eta-\phi_{k}^{\prime}(t)\right)\right| \tag{5.1.11}
\end{equation*}
$$

where $c_{k}$ is a polynomial of $\left(\left|\phi_{k-1}(t)-\phi_{k}(t)\right|+\alpha_{k}\right)$, therefore, thus

$$
\begin{aligned}
\left|V_{x}^{g_{1}}(t, \eta)\right| & \leq \sum_{k=1}^{K}\left|x_{k}(t) G_{1, k}\left(\eta-\phi_{k}^{\prime}(t)\right)\right|+\left|r e s_{0}\right| \\
& \leq \tau_{0} \sum_{k=1}^{K} A_{k}(t) c_{k}+\Pi_{0}(t) \leq \varepsilon_{1}
\end{aligned}
$$

which contradiction the assumption $\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1}$. Therefore, (a) holds.
Proof Part (b). Using the result of Lemma (5.1.2),

$$
\begin{equation*}
P_{1}(t, \eta)=i 2 \pi \phi_{k}^{\prime}(t) Q_{1}(t, \eta)+\operatorname{Res}_{3} . \tag{5.1.12}
\end{equation*}
$$

Then we have

$$
\omega_{x}^{N e w, 2 n d, c}(t, \eta)-\phi_{k}^{\prime}(t)=\operatorname{Res}_{4}
$$

where

$$
\begin{aligned}
\operatorname{Res}_{4} & =\frac{\operatorname{Res}_{3}}{i 2 \pi Q_{1}(t, \eta)} \\
& =\frac{\left(V_{x}^{g 1}(t, \eta)\right)^{2} \operatorname{Res}_{3}}{i 2 \pi\left(\partial_{\eta} V_{x}(t, \eta) V_{x}^{g 1}(t, \eta)-\partial_{\eta} V_{x}^{g 1}(t, \eta) V_{x}(t, \eta)\right)}
\end{aligned}
$$

However, the formula of $\mathrm{Res}_{3}$ is defined in the lemma (5.1.2), and we obtain

$$
\left(V_{x}^{g_{1}}(t, \eta)\right)^{2} \operatorname{Res}_{3}=\operatorname{Res}_{2} V_{x}^{g_{1}}(t, \eta)-\partial_{\eta} V_{x}^{g_{1}}(t, \eta) \operatorname{Res}_{1}
$$

Then,

$$
\operatorname{Res}_{4}=\frac{\operatorname{Res}_{2} V_{x}^{g_{1}}(t, \eta)-\partial_{\eta} V_{x}^{g_{1}}(t, \eta) \operatorname{Res}_{1}}{i 2 \pi\left(\partial_{\eta} V_{x}(t, \eta) V_{x}^{g 1}(t, \eta)-\partial_{\eta} V_{x}^{g 1}(t, \eta) V_{x}(t, \eta)\right)}
$$

Next, we have

$$
\begin{aligned}
\left|R e s_{4}\right| & =\left|\frac{\operatorname{Res}_{2} V_{x}^{g_{1}}(t, \eta)-\partial_{\eta} V_{x}^{g_{1}}(t, \eta) \operatorname{Res}_{1}}{V_{x}^{g_{1}}(t, \eta)^{2}}\right|\left|\frac{V_{x}^{g_{1}}(t, \eta)^{2}}{i 2 \pi\left(\partial_{\eta} V_{x}(t, \eta) V_{x}^{g 1}(t, \eta)-\partial_{\eta} V_{x}^{g 1}(t, \eta) V_{x}(t, \eta)\right)}\right| \\
& \leq \frac{1}{2 \pi \varepsilon_{2}}\left|\frac{\operatorname{Res}_{2} V_{x}^{g_{1}}(t, \eta)-\partial_{\eta} V_{x}^{g_{1}}(t, \eta) \operatorname{Res}_{1}}{V_{x}^{g_{1}}(t, \eta)^{2}}\right| \\
& \leq \frac{1}{2 \pi \varepsilon_{2}}\left(\frac{\left|\operatorname{Res}_{2}\right|}{\left|V_{x}^{g_{1}}(t, \eta)\right|}+\left|\frac{\partial_{\eta} V_{x}^{g_{1}}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)^{2}}\right|\left|\operatorname{Res}_{1}\right|\right) \\
& \leq \frac{1}{2 \pi \varepsilon_{1}^{2} \varepsilon_{2}}\left(\left|\operatorname{Res}_{2}\right|\left|V_{x}^{g_{1}}(t, \eta)\right|+\left|\operatorname{Res}_{1}\right|\left|\partial_{\eta} V_{x}^{g_{1}}(t, \eta)\right|\right)
\end{aligned}
$$

where

$$
B d_{1}=\max _{1 \leq k \leq K} \sup _{\eta \in O_{k}}\left\{\frac{1}{2 \pi \varepsilon_{1}^{2} \varepsilon_{2}}\left(\left|\operatorname{Res}_{2}\right|\left|V_{x}^{g_{1}}(t, \eta)\right|+\left|\operatorname{Res}_{1}\right|\left|\partial_{\eta} V_{x}^{g_{1}}(t, \eta)\right|\right)\right\} .
$$

Proof Part (c). First, we have the following result from [7] on p. 254

$$
\begin{equation*}
\lim _{\lambda \rightarrow 0} \int_{\left|\xi-\phi_{k}^{\prime}(t)\right|<\varepsilon_{3}} R_{x, \varepsilon_{1}, \varepsilon_{2}}^{2 n d, \lambda}(t, \xi) d \xi=\int_{X_{t}} V_{x}(t, \eta) d \eta \tag{5.1.13}
\end{equation*}
$$

where

$$
X_{t}:=\left\{\eta:\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1},\left|\partial_{\eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)\right|>\varepsilon_{2} \text { and }\left|\omega^{2 n d}(t, \eta)-\phi_{k}^{\prime}(t)\right|<\varepsilon_{3}\right\} .
$$

Denote

$$
Y_{t}:=\left\{\eta:\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1},\left|\partial_{\eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)\right|>\varepsilon_{2} \text { and }(t, \eta) \in O_{k}\right\} .
$$

According to theorem (2.6.4) part(b), if $\eta \in Y_{t}$, then

$$
\left|\omega^{2 n d}(t, \eta)-\phi_{k}^{\prime}(t)\right|<B d_{1} \leq \varepsilon_{3} .
$$

Thus, $\eta \in X_{t}$. This which implies the first inclusion $Y_{t} \subset X_{t}$.

Using theorem (2.6.4) part(a), if $\eta \in X_{t}$,
then $\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1}$ and there exists $l \in\{1,2, \ldots, K\}$ such that $(t, \eta) \in O_{l}$. If $l \neq k$, then

$$
\left|\omega_{x, \varepsilon_{1}, \varepsilon_{2}}^{N e w d}(t, \eta)-\phi_{k}^{\prime}(t)\right| \geq\left|\phi_{k}^{\prime}(t)-\phi_{l}^{\prime}(t)\right|-\left|\phi_{l}^{\prime}(t)-\omega_{x, \varepsilon_{1}, \varepsilon_{2}}^{N e w, 2 n d}(t, \eta)\right| .
$$

We use the following inequalities

$$
\left|\phi_{k}^{\prime}(t)-\phi_{l}^{\prime}(t)\right|>L_{k} \text { and }\left|\phi_{l}^{\prime}(t)-\omega_{x, \varepsilon_{1}, \varepsilon_{2}}^{\text {New,2nd }}(t, \eta)\right|<B d_{1} \leq \varepsilon_{3} .
$$

Therefore,

$$
\left|\omega_{x, \varepsilon_{1}, \varepsilon_{2}}^{N e w, 2 n d}(t, \eta)-\phi_{k}^{\prime}(t)\right|>L_{k}(t)-\varepsilon_{3} \geq \varepsilon_{3}
$$

and contradicts the assumption $\eta \in X_{t}$.
Therefore, $l=k$ and $\eta \in Y_{t}$, implying the second inclusion $X_{t}=Y_{t}$. We recall

$$
Z_{t}:=\left\{\eta:(t, \eta) \in O_{k},\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1},\left|\partial_{\eta}\left(\frac{V_{x}(t, \eta)}{V_{x}^{g_{1}}(t, \eta)}\right)\right| \leq \varepsilon_{2}\right\}
$$

The fact that $X_{t}=Y_{t}$ and $Y_{t} \cap Z_{t}=\phi$, with the equation (5.1.13), imply that

$$
\begin{align*}
\int_{Y_{t}} V_{x}(t, \eta) d \eta & =\lim _{\lambda \rightarrow 0} \int_{\left|\xi-\phi_{k}^{\prime}(t)\right|<\varepsilon_{3}} R_{x, \varepsilon_{1}, \varepsilon_{2}}^{2 n d}(t, \eta) d \xi . \\
& =\int_{Y_{t} \cup Z_{t}} V_{x}(t, \eta) d \eta-\int_{Z_{t}} V_{x}(t, \eta) d \eta  \tag{5.1.14}\\
& =\int_{\left\{\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1}\right\} \cap\left\{\eta:(t, \eta) \in O_{k}\right\}} V_{x}(t, \eta) d \eta-\int_{Z_{t}} V_{x}(t, \eta) d \eta
\end{align*}
$$

Using this equalition

$$
\begin{aligned}
\int_{\left\{\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1}\right\} \cap\left\{\eta:(t, \eta) \in O_{k}\right\}} V_{x}(t, \eta) d \eta & =\int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} V_{x}(t, \eta) d \eta \\
& -\int_{\left\{\left|V_{x}^{g_{1}}(t, \eta)\right| \leq \varepsilon_{1}\right\} \cap\left\{\eta:(t, \eta) \in O_{k}\right\}} V_{x}(t, \eta) d \eta
\end{aligned}
$$

we have

$$
\begin{array}{rlrl}
\mid \int_{\left\{\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1}\right\} \cap\left\{\eta:(t, \eta) \in O_{k}\right\}} & V_{x}(t, \eta) & d \eta-g(0) x_{k}(t) \mid \\
& = & \mid \int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} V_{x}(t, \eta) d \eta-g(0) x_{k}(t) \\
& - & \underbrace{}_{\text {Term }} \int_{\left\{\left|V_{v}^{g_{1}}(t, \eta)\right| \leq \varepsilon_{1}\right\} \cap\left\{\eta:(t, \eta) \in O_{k}\right\}} V_{x}(t, \eta) d \eta \mid \\
& \leq \underbrace{}_{\text {Term }} \int_{\left\{\eta:(t, \eta) \in O_{k}\right\}}(t, \eta) d \eta-g(0) x_{k}(t) \mid \\
& +\underbrace{\left|\int_{x}(t, \eta) d \eta\right|}_{\left\{\left|V_{x}^{g_{1}}(t, \eta)\right| \leq \varepsilon_{1}\right\} \cap\left\{\eta:(t, \eta) \in O_{k}\right\}}
\end{array}
$$

We recall that $O_{k}=\left\{(t, \eta):\left|\eta-\phi_{k}^{\prime}(t)\right|<\alpha_{k}, \quad t \in \mathbb{R}\right\}$
For the first term, we get the following results:

$$
\begin{aligned}
& \text { Term }_{1}=\left|\int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} V_{x}(t, \eta) d \eta-g(0) x_{k}(t)\right| \\
& \leq\left|\int_{\left\{\eta:(t, \eta) \in O_{k}\right\}}\left(\sum_{l=1}^{k} x_{l}(t) G_{0, l}(t, \eta)+r e s_{0}\right) d \eta-g(0) x_{k}(t)\right| \\
& \leq \underbrace{\int_{\left\{\eta:(t, \eta) \in O_{k}\right\}}\left|r e s_{0}\right| d \eta}_{T_{1}}+\underbrace{\left|x_{k}(t) \int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} G_{0, k}(t, \eta)-g(0)\right|}_{T_{3}} \\
&+ \underbrace{\sum_{l \neq k} x_{l}(t)\left|\int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} G_{0, l}(t, \eta) d \eta\right|}_{T_{2}} \\
& T_{1}= \int_{\left\{\eta:(t, \eta) \in O_{k}\right\}}\left|r e s_{0}\right| d \eta=\left|r e s_{0}\right| \int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} 1 d \eta=2 \alpha_{k}\left|r e s_{0}\right| \\
& T_{2}=\left|x_{k}(t) \int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} G_{0, k}(t, \eta)-g(0)\right| \\
&=\left|x_{k}(t) \int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} G_{k}\left(\eta-\phi_{k}^{\prime}(t)\right)-g(0)\right|
\end{aligned}
$$

Using change of variable $u=\eta-\phi_{k}^{\prime}(t)$,

$$
\begin{aligned}
T_{2} & \left.=\mid x_{k}(t) \int_{|u|<\alpha_{k}} G_{k}(u)\right)-g(0) x_{k}(t) \mid \\
& \left.=\mid x_{k}(t) g(0)-g(0) x_{k}(t)-x_{k}(t) \int_{|u| \geq \alpha_{k}} G_{k}(u) d u\right) \mid \\
& =\left|x_{k}(t)\right|\left|\int_{|u| \geq \alpha_{k}} G_{k}(u) d u\right|=A_{k}(t)\left|\int_{|u| \geq \alpha_{k}} G_{k}(u) d u\right| \\
T_{3} & =\sum_{l \neq k}\left|x_{l}(t) \int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} G_{0, l}(t, \eta) d \eta\right| \\
= & \sum_{l \neq k}\left|x_{l}(t)\right|\left|\int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} G_{0, l}(t, \eta) d \eta\right|=\sum_{l \neq k} A_{l}(t) M_{l, k}(t)
\end{aligned}
$$

substituting the parameters estimated above we get

$$
\begin{gathered}
\operatorname{Term}_{1} \leq 2 \mid \text { res }_{0}\left|\alpha_{k}+A_{k}(t)\right| \int_{|u| \geq \alpha_{k}} G_{k}(u) d u \mid+\sum_{l \neq k} A_{l}(t) M_{l, k}(t) . \\
\text { Term }_{2} \leq \varepsilon_{0} \int_{(t, \eta) \in O_{k}} 1 d \eta=\varepsilon_{0} \int_{\phi_{k}^{\prime}(t)-\alpha_{k}}^{\phi_{k}^{\prime}(t)+\alpha_{k}} 1 d t \leq 2 \varepsilon_{0} \alpha_{k}
\end{gathered}
$$

We group all parameter estimates, we find the following results

$$
\begin{aligned}
& \left\lvert\, \int_{\left\{\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1}\right\} \cap\left\{\eta:(t, \eta) \in O_{k}\right\}} \begin{aligned}
& V_{x} \\
&(t, \eta) d \eta-g(0) x_{k}(t) \mid \\
& \leq 2\left(\left|r e s_{0}\right|+\varepsilon_{0}\right) \alpha_{k}+A_{k}(t)\left|\int_{|u| \geq \alpha_{k}} G_{k}(u) d u\right|+\sum_{l \neq k} A_{l}(t) M_{l, k}(t) \\
&\left|\frac{1}{g(0)} \int_{\left\{\left|V_{x}^{g_{1}}(t, \eta)\right|>\varepsilon_{1}\right\} \cap\left\{\eta:(t, \eta) \in O_{k}\right\}} V_{x}(t, \eta) d \eta-x_{k}(t)\right| \leq B d_{2}^{\prime} \\
& \int_{\mathbb{R}} G_{k}(u) d u=\int_{\mathbb{R}} \mathcal{F}\left(e^{i 2 \pi \phi_{k}^{\prime \prime}(t) \tau^{2}} g(\tau)\right)(u) d u=g(0)
\end{aligned} .\right.
\end{aligned}
$$

Hence, we have

$$
\begin{aligned}
\left|\int_{Z_{t}} V_{x}(t, \eta) d \eta\right| & =\left|\int_{Z_{t}}\left(\sum_{\ell=1}^{K} x_{\ell}(t) G_{0, \ell}(t, \eta)+r e s_{0}\right) d \eta\right| \\
& \leq\left|r e s_{0}\right| 2 \alpha_{k}+A_{k}(t) \sup _{\eta \in Z_{t}}\left|G_{k}\left(\eta-\phi_{k}^{\prime}(t)\right)\right|\left|Z_{t}\right| \\
& +\sum_{\ell \neq k} A_{\ell}(t)\left|\int_{\left\{\eta:(t, \eta) \in O_{k}\right\}} G_{0, \ell}(t, \eta) d \eta\right| \\
& \leq 2\left|r e s_{0}\right| \alpha_{k}+\left.A_{k}(t)| | g\right|_{1}\left|Z_{t}\right|+\sum_{\ell \neq k} A_{\ell}(t) M_{\ell, k}(t) \leq B d_{2}^{\prime \prime}
\end{aligned}
$$

Then we have the result

$$
\left|\lim _{\lambda \rightarrow 0} \frac{1}{g(0)} \int_{\left|\xi-\phi_{k}^{\prime}(t)\right|<\varepsilon_{3}} R_{x, \varepsilon_{1}, \varepsilon_{2}}^{a d p, 2 n d, \lambda}(t, \xi) d \xi-x_{k}(t)\right| \leq B d_{2},
$$

This completes proof of Theorem.

## Chapter 6

## Conclusion and future work

In this study, we introduced a generalization of the STFT-based SST (FSST) with time-varying, CWT-based SST (WSST), the WSST with time-varying, and the FSST with a new phase transformation by using higher order amplitude and phase approximations. This generalization allows us to better assess a wide variety of multicomponent signals containing very strongly modulated AM-FM modes.

We also studyed the theoretical analysis of the 2nd-order FSST with a new phase transformation. The new phase transformation is much simpler than the convectional one. The new FSST performance in IF estimation and component recovery is comparable with that of the conventional 2nd-order FSST. In some cses, the new FSST perfermed even better than its conventional counterpart.

Since the result showed a better concentration and reconstruction for a wider variety of AM-FM modes, we will continue working on the adaptive FSST with a new phase transformation by using higher order approximations both for the amplitude and phase.

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