



Sampling Rate Conversion in the Discrete Linear Canonical Transform Domain

Z. Zhuo^{*a}, N. Zhong^b, X. Zhan^a

^a School of Information Communication Engineering, Beijing Information Science and Technology University, Beijing, China

^b China Youth University of Political Sciences Beijing, China

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ABSTRACT

Sampling rate conversion (SRC) is one of important issues in modern sampling theory. It can be realized by up-sampling, filtering, and down-sampling operations, which need large complexity. Although some efficient algorithms have been presented to do the sampling rate conversion, they all need to compute the N -point original signal to obtain the up-sampling or the down-sampling signal in the time domain. Most of the published papers about the sampling rate conversion require the signal to be band limited in the Fourier transform domain, and there are few paper published related to the SRC in the linear canonical transform (LCT) domain. This paper investigates how to perform the SRC in the discrete linear canonical transform (DLCT) domain for integer and fractional rate conversion. The simulations are performed to verify the correctness of the proposed results.

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1. INTRODUCTION

Sampling rate conversion (SRC) is one of important concepts in modern signal processing community. It can be realized by up-sampling, filtering and down-sampling operations in the time domain [1, 2], which need large computational complexity. Some efficient algorithms have been presented to do the sampling rate conversion. However, they all need to do it in the time domain. To reduce the computational load as well as saving the storage space, a novel kind of SRC in the FT domain has been studied in detail recently. This method converts a discrete sequence into another discrete sequence with a different sampling rate in the spectrum, which is quite different from the process in [1, 2]. Usually, the SRC is performed by manipulating the FFT of a signal. We only need to formulate the sampling points after up-sampling or down-sampling, which largely reduced the computational costs. The method of [3] requires that the processed signal must be bandlimited in the FT domain. However, in real applications, for example in image zooming techniques, there are lots of signals which are non-bandlimited in the FT domain [3, 4], but bandlimited in the other transform domains. For these

signals, the above methods cannot be used, or it may leads to suboptimal conclusions. Therefore, it is interesting in theory and worthwhile in practice to perform the SRC in transform domains.

As a generalization of the classical Fourier transform and the fractional Fourier transform, the linear canonical transform is an integral transform with four parameters. Many transforms in signal processing community and applied mathematics fields, such as the Fourier transform, the fractional Fourier transform (FrFT), the Fresnel transform and scaling operations, are special cases of the LCT [5-8]. It has found many advantages and applications in radar signal processing, optical signal processing, image processing and many other areas [1-10]. Recently, the theories associated with sampling rate conversion in the LCT domain have also been presented by Zhao [11] as shown in Figure 1.

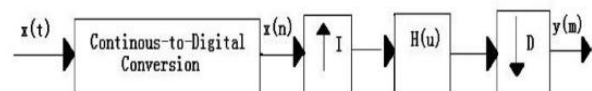


Figure 1. Sampling rate conversion in LCT domain

*Corresponding Author's Email: zhuozhihai@bistu.edu.cn (Z. Zhuo)

The method proposed in [11] discusses the SRC relation in the discrete time linear canonical transform domain (DTLCT). Unlike the method in [11], in this paper, we derive a new SRC operation related to the discrete linear canonical transform domain (DLCT) in this paper. This paper is organized as follows, in Section 2, we study the interpolation and decimation associating with LCT and the discrete LCT [12-15]. In Section 3, we propose the sampling rate conversion in the LCT domain. In Section 4, simulations are given to verify the achieved results. Finally, we make a conclusion in Section 5.

2. PRELIMINARY

2.1. The Linear Canonical Transform The LCT [1-7] of a signal $x(t)$ with parameter $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ is defined as:

$$L_A(x(t))(u) = \begin{cases} \sqrt{\frac{1}{j2\pi b}} e^{j(a/2b)u^2} \int_{-\infty}^{+\infty} x(t) e^{-j(d/2b)t^2} dt, & b \neq 0 \\ \sqrt{d} e^{j(cd/2)u^2} x(du), & b = 0 \end{cases} \quad (1)$$

where, parameters a, b, c, d are real numbers and satisfy the relation of $ad - bc = 1$. In this paper, we consider the case of $b > 0$ in the following sections.

The inverse transform of the LCT is proved to be a LCT with parameter $A^{-1} = \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$. Hence, we can obtain the original signal $x(t)$ from $L_A[x](u)$ by the following equation.

$$x(t) = L_{A^{-1}}[L_A[x](u)](t) = \begin{cases} \sqrt{\frac{1}{j2\pi b}} e^{-j(a/2b)t^2} \int_{-\infty}^{+\infty} L_A[x](u) e^{-j(d/2b)u^2} du, & b \neq 0, \\ \sqrt{a} e^{-j(ca/2)t^2} \int_{-\infty}^{+\infty} L_A[x](au), & b = 0. \end{cases} \quad (2)$$

Most of the concepts and theories in the classical Fourier domain are generalized to the LCT domain by different researchers. For example, the sampling theories [16-23], the Wigner-Ville distribution [10, 24], the ambiguity function [25], the convolution and product theories [26, 27], and the uncertainty principle

[28], the spectral analysis [29] are well studied in the LCT domain. The eigen functions, speech recovery and Instantaneous frequency estimation are also studied in the LCT domain [30-36].

2.2. Discrete LCT In order to investigate the SRC problems in the LCT domain, we need to define the discrete LCT (DLCT). Along with applications of the LCT in the signal processing community, the discrete LCT and its efficient algorithms are becoming hot research topics [12-15]. The exact relation between continuous and discrete LCTs is presented in [15]. We use the following discrete time LCT of $x(n)$ of Equation (1).

$$X_{(a,b,c,d)}(u) = \sqrt{\frac{1}{j2\pi b}} e^{j(d/2b)u^2} \sum_{n=-\infty}^{+\infty} x(n) e^{j(a/2b)n^2 \Delta t_x^2} e^{-j(1/b)un \Delta t_x}, \quad (3)$$

where, $x(n)$ is the sampled signal from the continuous signal $x(t)$ with the sampling period Δt_x . The DLCT can be defined as following:

$$X_{(a,b,c,d)}(m) = \sqrt{\frac{1}{j2\pi b}} e^{j(d/2b)m^2} \sum_{n=0}^{N-1} x(n) e^{j(a/2b)n^2 \Delta t_x^2} \times e^{-j(2\pi nm/N)} \quad (4)$$

3. MAIN RESULTS

We assume that the original signal $x(t)$ is band-limited in the LCT domain, i.e., $X_{(a,b,c,d)}(u) = 0$ for $|u| > u_c$, where $\Delta t_x \leq \pi|b|/|u_c|$. In the following part, we will discuss the SRC by performing the discrete sequence in the LCT domain.

3.1. Integer Sampling Rate Increase Let us define $X(m)$ to be the N point DLCT of the sequence $x(n)$ and $Y(m)$ to be the N_1 -point DLCT of the sequence $y(n)$, where $N_1 = IN$. $x(n)$ is obtained by sampling $x(t)$ at sampling frequencies F_x .

Theorem 1. To increase the sampling rate by an integer factor I , it can be achieved by passing the original $x(n)$ through an I -fold expander. The sequence $y(n)$ is:

$$y(n) = \begin{cases} x(n/I), & n = Ik, k \in \mathbb{Z}, \\ 0, & \text{otherwise.} \end{cases} \quad (5)$$

The sampling period of $y(n)$ is $\Delta t_y = \Delta t_x / I$. Then, the DLCT of $y(n)$ is:

$$Y(m) = \begin{cases} X(m), & m \leq N, \\ e^{j(d/2b)(\frac{2\pi b}{N\Delta t_x})^2 (k^2 N^2 + 2knN)} X(r), & m = kN + r. \end{cases} \quad (6)$$

Proof. According to Equation (4), the DLCT $Y(m)$ of $y(n)$ is:

$$Y(m) = \sqrt{\frac{1}{j2\pi b}} e^{j(d/2b)(bm2\pi/Ni\Delta t_y)^2} \sum_{n=0}^{N-1} y(n) e^{j(d/2b)n^2\Delta t_y^2} \times e^{-j(2\pi mn/Ni)}$$

when $m \leq N$,

$$Y(m) = \sqrt{\frac{1}{j2\pi b}} e^{j(d/2b)(\frac{2\pi bm}{N\Delta t_x})^2} \sum_{n=0}^{N-1} x(n) e^{j(a/2b)n^2\Delta t_x^2} \times e^{-j(2\pi mn/N)}$$

$$= X(m)$$

else $m = kN + r$,

$$Y(m) = \sqrt{\frac{1}{j2\pi b}} e^{j(d/2b)(\frac{2\pi b(kN+r)}{N\Delta t_y})^2} \sum_{n=0}^{N-1} x(n) e^{j(a/2b)n^2\Delta t_x^2} \times e^{-j(kN+r)n\frac{2\pi}{N}}$$

$$= e^{j(d/2b)(\frac{2\pi b}{N\Delta t_x})^2(k^2N^2+2krN)} X(r)$$

Theorem 2. Based on $X(k), \hat{Y}(k)$ the DLCT of output signal can be derived as:

$$\hat{Y}(k) = \begin{cases} IX(k), & 0 \leq k \leq \frac{N}{2}, \\ 0, & \frac{N}{2} \leq k \leq NI - \frac{N}{2}, \\ IX(k - IN + N), & NI - \frac{N}{2} \leq k \leq NI. \end{cases} \quad (9)$$

Proof. Let us define $\hat{Y}(k)$ to be the N_1 -point DLCT of there constructed sequence $y(n)$ where $N_1 = IN$. Theorem1 shows that up-sampling creates an imaging effect. To remove the imaging effect, low-pass filtering is needed. The spectrum after filter is equal to insert $I(N-1)$ -point zero $\sin X(k)$. To get the same result of the filtering, the spectrum $X(k)$ can be used to form $\hat{Y}(k)$ by inserting several zeros into $\hat{Y}(k)$'s frequency region between π/I and $2\pi - \pi/I$ as show in Equation (9). Meanwhile Equation (9) is not the only method to form $\hat{Y}(k)$. We can choose other methods by inserting $(I-1)N$ zeros into $\hat{Y}(k)$'s frequency region.

3. 2. Integer Sample Rate Decrease Theorem 3. To reduce the sampling rate by an integer factor D , it can be achieved by passing the original signal $x(n)$ through an D -fold decimator and the sequence $y(n)$ is $y(n) = x(Dn)$. The sampling period of $y(n)$ is $\Delta t_y = D\Delta t_x$, then the DLCT of $y(n)$ is:

$$Y(m) = \frac{1}{D} \sum_{i=0}^{D-1} X(m - \frac{N}{D}i) e^{-j(a/2b)(\frac{b2\pi}{N\Delta t_x})^2 2((\frac{Ni}{D})^2 - \frac{2mNi}{D})} \quad (10)$$

Proof. According to Equation (4), the DLCT $Y(m)$ of $y(n)$ is:

$$Y(m) = \sqrt{\frac{1}{j2\pi b}} e^{-j(a/2b)(\frac{bm2\pi}{N\Delta t_y})^2} \sum_{i=0}^{D-1} X(Di) e^{-j(a/2b)i^2\Delta t_y^2} \times e^{-jmnD\frac{2\pi}{N}}$$

The right side of Equation (10) can be rewritten as:

$$\frac{1}{D} \sum_{i=0}^{D-1} X(m - \frac{N}{D}i) e^{-j(a/2b)(\frac{b2\pi}{N\Delta t_x})^2 2((\frac{Ni}{D})^2 - \frac{2mNi}{D})}$$

$$= \frac{1}{D} \sqrt{\frac{1}{j2\pi b}} e^{-j(d/2b)(\frac{b2\pi}{N\Delta t_x})^2} \sum_{i=0}^{D-1} \sum_{k=0}^{N-1} x(k) e^{-j(d/2b)k^2\Delta t^2} \times e^{-jk(m - \frac{N}{D}i)\frac{2\pi}{k}}$$

$$= \sqrt{\frac{1}{j2\pi b}} e^{-j(d/2b)(\frac{b2\pi}{N\Delta t_x})^2} \sum_{k=0}^{N-1} y(k) e^{-j(d/2b)k^2\Delta t^2} \times e^{-jk(m - \frac{N}{D}i)\frac{2\pi}{k}}$$

$$= Y(m)$$

Theorem 4. For our SRC, $\hat{Y}(k)$, is the N_1 -point DLCT of the reconstructed sequence $y(n)$ where $N_1 = \frac{N}{D}$ can be formulated by manipulating $X(k)$. It can be presented as following.

$$\hat{Y}(k) = \begin{cases} \frac{1}{D} X(k), & 0 \leq k \leq \frac{N}{2D}, \\ X(\frac{N}{2}), & k = \frac{N}{2D}, \\ \frac{1}{D} X(k + N - \frac{N}{D}), & \frac{N}{2D} \leq k \leq \frac{N}{D}. \end{cases} \quad (13)$$

Equation (13) is also not the only method to form $\hat{Y}(k)$. We can choose other points of $X(k)$ to form $\hat{Y}(k)$.

3. 3. Overlap Approach for Long Sequences
The method described above is only suitable for short inputs sequences due to the limitation of the maximum DLCT length for practical applications. For long input sequences, a widely used approach is to divide the long input sequence into many shorter segments, which are processed individually [3].

If each $x(n)$ input time-domain segment has N points, the adjacent input segments are overlapped by

2L points, and from each segment an N-point $X(k)$ DLCT is computed. The previously described methods are then used to obtain an N_1 point $\hat{Y}(k)$ DLCT segment for each $X(k)$ segment, where $\frac{N_1}{N} = \frac{I}{D}$.

After performing an N_1 -point inverse DLCT on each $\hat{Y}(k)$ segment, the corresponding N_1 -point time-domain segments are obtained. As shown in Figure 2, the cascading operation with L_1 -point overlapping, where $L_1/L = I/D$ for each pair of adjacent time segments, is performed to recover the final $y(n)$ time-domain sequence with the desired F_y sampling frequency, where the shaded time samples are discarded.

4. SIMULATION RESULTS

A real function is used in the experiment; it can be seen as a bandlimited signal in the LCT domain. However, it is not a bandlimited signal in the FT domain. Under such circumstances, we can use the method proposed in this article to solve the problem.

It is shown in Figures (3-4) when the sampling rate increased by two times. According to Equation (4), Figure 3 shows the computed LCT $X(m)$ of $x(n)$ and the generation of $\hat{Y}(m)$ from $X(m)$. From the inverse LCT of $\hat{Y}(m)$ and $y(n)$ is the ideal up-sampling or down-sampling sequence. We can see in Figure 4 that in the time domain, $y(n)$, has small differences from the desired signal $y(n)$. According to Equation (9), the values of $X(m)$, where $\frac{N}{2} \leq k \leq NI - \frac{N}{2}$, are ignored.

The other values are rearranged as shown in Figure 3 to obtained the $\hat{Y}(k)$. The decreasing case is shown in Figures 5 and 6.

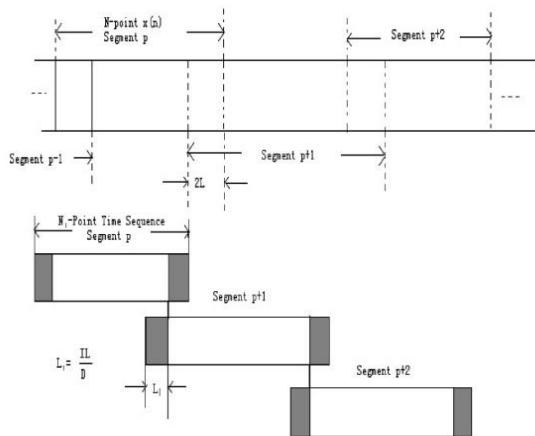


Figure 2. Processing long sequences

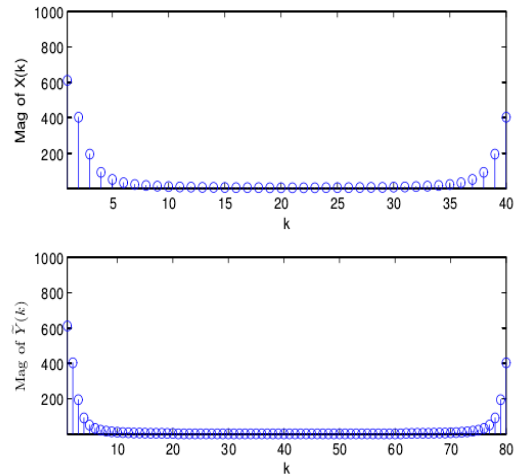


Figure 3. The spectrum $X(m), \hat{Y}(m)$ when $D=1, I=2$ and $N=40$.

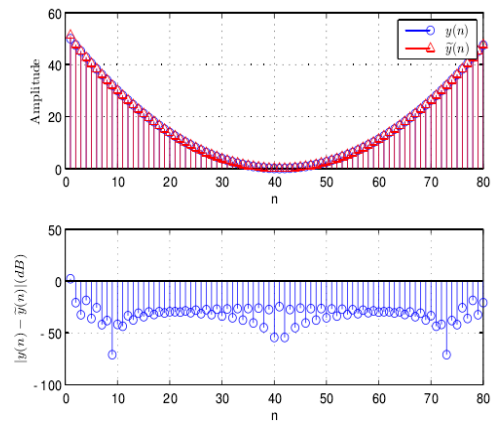


Figure 4. SRC errors: $\hat{y}(n), y(n)$ and $|\hat{y}(n) - y(n)|$, when $D=1, I=2$ and $N=40$

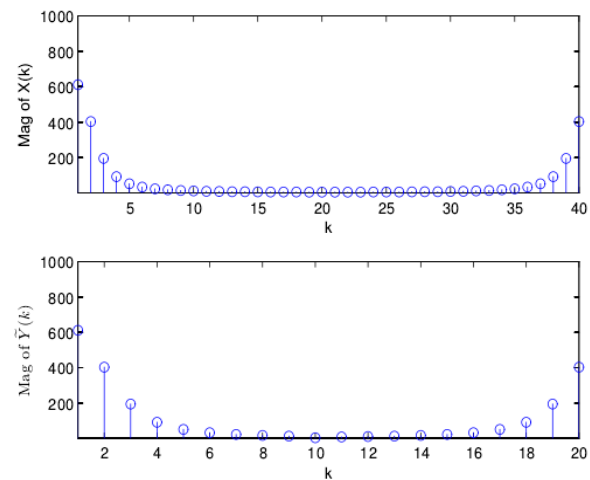


Figure 5. The spectrum $X(m), \hat{Y}(m)$ when $D=2, I=1$ and $N=40$.

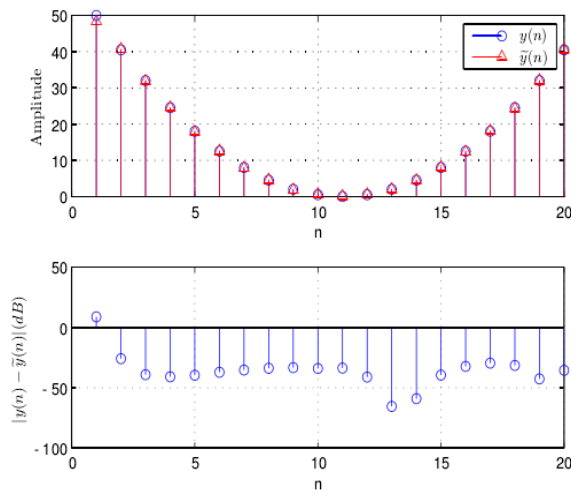


Figure 6. SRC errors: $\hat{y}(n)$, $y(n)$ and $|y(n) - \hat{y}(n)|$, when $D=1$, $I=2$ and $N=40$

5. CONCLUSIONS

Unlike the traditional methods in [11] associate with the discrete time LCT (DTLCT) domain, this paper propose a new method for sampling rate conversion in the discrete LCT (DLCT) domain. We show that the proposed SRC can be realized by combining an efficient DLCT. The simulations are also performed to verify the correctness of the derived results. What's more, with the overlapping technique, the long sequences can also be discussed. The derived results in this paper can be used in signal and image processing in future works [5, 33, 34].

6. ACKNOWLEDGEMENTS

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Sampling Rate Conversion in the Discrete Linear Canonical Transform Domain

Z. Zhuo^a, N. Zhong^b, X. Zhan^a

^a School of Information Communication Engineering, Beijing Information Science and Technology University, Beijing, China

^b China Youth University of Political Sciences Beijing, China

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تبدیل نرخ نمونه برداری (SRC) یکی از مسائل مهم در تئوری نمونه برداری مدرن است. این مسئله می تواند توسط نمونه برداری بالا، فیلتر کردن و عملیات نمونه برداری پایین، که نیاز به پیچیدگی بزرگ دارد، درک شود. اگر چه برخی از الگوریتم های کارآمد برای انجام تبدیل نرخ نمونه برداری ارائه شده است، همه آنها نیاز به محاسبه سیگنال اصلی نقطه N برای به دست آوردن نمونه برداری بالا و یا سیگنال نمونه برداری پایین در حوزه زمان دارد. بسیاری از مقالات منتشر شده در مورد تبدیل نرخ نمونه برداری نیاز به سیگنال به باند محدود در دامنه تبدیل فوریه دارد و تعداد کمی مقاله منتشر شده مربوط به SRC در دامنه متعارف تبدیل خطی (LCT) وجود دارد. این مقاله به بررسی نحوه انجام SRC در دامنه متعارف خطی گسسته (DLCT) برای عدد صحیح و تبدیل نرخ کسری می پردازد. این شبیه سازی ها به منظور بررسی صحت نتایج ارائه شده انجام می شود.

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