PERIODICALLY NONUNIFORM SAMPLING OF A NEW CLASS OF BANDPASS SIGNALS

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Abstract. It is known that a continuous time signal x(t) with Fourier transform $X(\nu)$ band-limited to $|\nu| < \Theta/2$ can be reconstructed from its samples $x(T_0n)$ with $T_0 = 2\pi/\Theta$. In the case that $X(\nu)$ consists of two bands and is band-limited to $\nu_0 < |\nu| < \nu_0 + \Theta/2$, successful reconstruction of x(t) from $x(T_0n)$ requires that these two bands be located properly. When the two bands are not located properly, Kohlenberg showed that we can use a Periodically Nonuniform Sampling (PNS) scheme to recover x(t). In this paper, we show that PNS scheme can be generalized and applied to a wider class. Further generalizations will be made to two-dimensional case and discrete-time case.

1. INTRODUCTION

It is well-known that successful reconstruction of a continuous-time bandpass signal x(t) (Fig. 1) from samples $x(nT_0)$, where $T_0 = 2\pi/\Theta$, depends on the relative positions of these two bands [1]. A necessary and sufficient condition is that, the band edge ν_0 must be an integer multiple of $\Theta/2$. It can be shown that a much wider class of signals with total bandwidth Θ can be recovered from samples at nT_0 . To be more specific, define the support of $X(\nu)$ (denoted by $Supp\{X(\nu)\}$) to be the set of frequencies for which $X(\nu) \neq 0$. Then x(t) can be obtained from $x(nT_0)$ if and only if no two frequencies in $Supp\{X(\nu)\}$ overlap under modulo Θ operation [2]. Such signals are called aliasfree (T_0) zones.

When the two bands of $X(\nu)$ (Fig. 1) are not properly located, Kohlenberg [3] proposed a periodically nonuniform sampling approach to recover x(t) (Fig. 2 with L = 2). In this scheme two sets of samples, x(nT) and $x(nT + d_1)$, where $T = 2T_0$, as shown in Fig. 3, are used. The average sampling rate is still Θ . Then x(t) can be reconstructed by properly choosing d_1 and the synthesis filters $f_0(t)$ and $f_1(t)$ [3]. This is called periodically nonuniform sampling of second order (PNS(2)), [4], for there are two sets of uniform samples involved. Recently, general *L*th order periodically nonuniform sampling (PNS(*L*)) for such two bands signals has been considered in [5].

In discrete time case, sampling is replaced by dec-

imation. PNS(L) sampling retains L sets of samples, $x(Mn + d_0)$, $x(Mn + d_1)$, \cdots , $x(Mn + d_{L-1})$ or the d_0 th, d_1 th, \ldots , d_{L-1} th polyphase components [6]. In [7], PNS(L) sampling and reconstruction has been considered for a subclass of L-band signals. The subclass addressed therein are those whose frequency supports are the union of L bands, each band with bandwidth $2\pi/M$ and band edges at integer multiples of $2\pi/M$. Such a L-band sequence x(n) can be reconstructed from its first L polyphase components, i.e., x(Mn), x(Mn+1), \cdots , x(Mn+L-1) [7].

In this paper, we will generalize the results in [3] and [7] to a wider class of signals in terms of frequency supports. We will show that from PNS(L) samples we can reconstruct signals in the class U(T, L), which is the collection of signals whose supports are the union of L non overlapping aliasfree(T) sets [8].[‡] The discretetime version of these will also be addressed. We will see that 1D discrete-time U(M, L) sequences can always be reconstructed from their first L polyphase components. However, in 2D discrete-time case only a subclass of U(M, L) signals allows reconstruction from Lpolyphase components.

Notations

1. The support of $X(\nu)$ (denoted by $Supp\{X(\nu)\}$) is defined as the set of frequencies for which $X(\nu) \neq 0$.

2. A set S is called an aliasfree(T) zone if no two frequencies in S overlap under modulo $2\pi/T$ operation. When $Supp\{X(\nu)\}$ is an aliasfree(T) zone, x(t) is called aliasfree(T).

3. The notation U(T, L) represents the collection of signals whose frequency supports are the union of L non overlapping aliasfree(T) sets.

2. PERIODICALLY NONUNIFORM SAMPLING OF LTH ORDER—CONTINUOUS TIME

In this section, we consider periodically nonuniform sampling of *L*th order (PNS(*L*)) for the class U(T, L). In PNS(*L*) sampling of x(t), there are *L* sets of samples, x(nT), $x(nT + d_1)$, \cdots , $x(nT + d_{L-1})$. Referring to Fig. 2, the sampling rate is $\sigma = 2\pi/T$ in each channel

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[‡] Throughout this paper, we will assume that aliasfree(T) sets contain only finitely many intervals.

and the average sampling rate is $L\sigma$, which is the total bandwidth of x(t). We will show that these L sets of uniform samples can be used to reconstruct x(t).

In the ℓ th channel, $y_{\ell}(t)$ contains the samples $x(nT + d_{\ell})$; $Y_{\ell}(\nu)$ consists of shifted versions of $X(\nu)$,

$$Y_{\ell}(\nu) = \frac{1}{T} \sum_{k} X(\nu - k\sigma) e^{-jkd_{\ell}\sigma}$$

Because the total bandwidth of x(t) is $L\sigma$ and the sampling rate is σ in each channel, serious aliasing occurs in $y_{\ell}(t)$. To be more quantitative, we first partition the support of $X(\nu)$ into L non overlapping aliasfree(T) sets, $\{S_i\}_{i=0}^{L-1}$. Define $X_i(\nu)$ to be the part of $X(\nu)$ on S_i , i.e., $X_i(\nu) = \begin{cases} X(\nu), & \nu \in S_i \\ 0, & \text{otherwise.} \end{cases}$ the set $S_i, Y_{\ell}(\nu)$ contains $X_i(\nu)$ and L-1 shifted copies, one from each $X_m(\nu), m \neq i$. Suppose the shifted amounts are respectively $\beta_1(\nu)\sigma, \beta_2(\nu)\sigma, \ldots, \beta_{L-1}(\nu)\sigma$. Because $X_i(\nu)$ are non overlapping, it can be verified that for $\nu \in Supp\{X(\nu)\}$,

$$Y_{\ell}(\nu) = \frac{1}{T} \left(X(\nu) + \sum_{i=1}^{L-1} X(\nu - \beta_i(\nu)\sigma) e^{-j\beta_i(\nu)d_{\ell}\sigma} \right).$$

Notice that $\beta_i(\nu)$ thus defined are piecewise constant because $Supp\{X(\nu)\}$ is the union of finitely many intervals.

Lemma 1 [3]. A U(T,L) signal x(t) can be recovered from its PNS(L) samples if and only if the equation below has a solution $f(\nu)$ for every $\nu \in Supp\{X(\nu)\}$.

$$\mathbf{A}(\boldsymbol{\nu})\mathbf{f}(\boldsymbol{\nu}) = T\mathbf{e}_0 \tag{1}$$

where $A(\nu)$ is

$$[\mathbf{A}(\nu)]_{0\ell} = [\mathbf{A}(\nu)]_{\ell 0} = 1, \quad 0 \le \ell \le L - 1 [\mathbf{A}(\nu)]_{i\ell} = e^{-j\beta_i(\nu)d_\ell\sigma}, \quad 1 \le i, \ell \le L - 1$$

 $\mathbf{f}(\boldsymbol{\nu}) = \begin{bmatrix} F_0(\boldsymbol{\nu}) & F_1(\boldsymbol{\nu}) & \dots & F_{L-1}(\boldsymbol{\nu}) \end{bmatrix}^T \text{ and the vector} \\ \mathbf{e}_0 \text{ is } \begin{bmatrix} 1 & 0 & \dots & 0 \end{bmatrix}^T.$

A nonsingular $A(\nu)$ will yield unique solutions for the synthesis filters $F_{\ell}(\nu)$. If we choose $d_{\ell} = \ell d_1$, $A(\nu)$ becomes a Vandermonde matrix; the nonsingularity condition becomes much more tractable.

Theorem 1 [8]. Consider a U(T, L) signal x(t). There always exist constant d_{ℓ} , $0 < \ell < L$ and synthesis filters $F_{\ell}(\nu)$, $0 \leq \ell < L$ such that $x(t) = \sum_{\ell=0}^{L-1} \sum_{n} x(nT + d_{\ell}) f_{\ell}(t - nT)$ (with $d_0 = 0$ in this expression). In particular, the choice

$$d_{\ell} = \ell d_1, \quad \ell = 1, 2, \cdots, L-1,$$

leads to a Vandermonde $A(\nu)$, which is nonsingular if

$$d_1 \neq rac{nT}{eta_i(
u)} \quad ext{and} \quad d_1 \neq rac{nT}{(eta_i(
u) - eta_m(
u))}, i \neq m,$$

for all integer *n*. The existence of such d_1 is guaranteed. In this case, $\mathbf{f}(\nu) = [F_0(\nu) \dots F_{L-1}(\nu)]^T$ is given by

$$\mathbf{f}(\nu) = \begin{cases} T\mathbf{A}(\nu)^{-1}\mathbf{e}_0, & \nu \in Supp\{X(\nu)\}, \\ 0, & \text{otherwise,} \end{cases}$$

where $e_0 = \begin{bmatrix} 1 & 0 & \dots & 0 \end{bmatrix}^T$.

Remark. The synthesis filters thus obtained are functions of $\beta_i(\nu)$ and can be verified to be piecewise constant [8].

Two-dimensional (2D) case. The aliasfree(T) property and aliasfree(T) zones can be defined as in 1D case. But now the sampling period T is a 2×2 nonsingular matrix and the samples are located on the lattice defined by T, i.e., located at Tn for all integer vectors n. It can be shown that 2D U(T, L) signals can be recovered from PNS(L) samples, Tn, Tn + d₁, ..., Tn + d_{L-1}. A result similar to that presented in Theorem 1 can be derived.

3. ONE-DIMENSIONAL DISCRETE-TIME PERIODICALLY NONUNIFORM SAMPLING

In discrete-time PNS(L) sampling (Fig. 4) the total amount of data after decimation is L/M times the original input; the nonuniform sampling scheme makes sense only for L < M, which will be assumed throughout this paper. In 1D continuous-time case, we saw that the class U(T, L) allows reconstruction from PNS(L) samples. In this section, a parallel theorem for 1D discrete-time U(M, L) signals will be developed.

The signal $y_{\ell}(n)$ is the polyphase component $x(nM + d_{\ell})$ and $Y_{\ell}(\omega)$ can be expressed in terms of shifts of $X(\omega)$,

$$Y_{\ell}(\omega) = \frac{1}{M} \sum_{k=0}^{M-1} X(\omega - k \frac{2\pi}{M}) e^{-j \frac{2\pi}{M} k d_{\ell}}.$$

The U(M, L) nature of $X(\omega)$ implies that only L terms in the above summation are nonzero. In particular, on the support of $X(\omega)$, $X(\omega)$ and L-1 shifted copies of $X(\omega)$ are nonzero. Let us denote these shifted copies by $X(\omega - \frac{2\pi}{M}\beta_i(\omega))$, $i = 1, 2, \dots, L-1$.

Lemma 2. A 1D discrete-time U(M, L) signal x(n) can be recovered from L of its polyphase components if and only if the equation to follow has a solution for every $\nu \in Supp\{X(\omega)\}$.

$$\mathbf{A}(\omega) \begin{bmatrix} F_0(\omega) & F_1(\omega) & \dots & F_{L-1}(\omega) \end{bmatrix}^T = T\mathbf{e}_0, \quad (2)$$

where $\mathbf{e}_0^T = \begin{bmatrix} 1 & 0 & \dots & 0 \end{bmatrix}$ and the matrix $\mathbf{A}(\omega)$ is given by

$$\begin{split} [\mathbf{A}(\omega)]_{0\ell} &= [\mathbf{A}(\omega)]_{\ell 0} = 1, \quad 0 \le \ell \le L - 1 \\ [\mathbf{A}(\omega)]_{i\ell} &= e^{-j\frac{2\pi}{M}\beta_i(\omega)d_\ell}, \quad 1 \le i, \ell \le L - 1 \end{split}$$

Observe that the matrix $A(\omega)$ is a $L \times L$ submatrix of the $M \times M$ DFT matrix, W_M given by $[W_M]_{mn} = e^{-j\frac{2\pi}{M}mn}, 0 \leq m, n < M$. Notice that any $L \times L$ sub-matrix of W_M obtained by retaining the first L columns of W_M and some L rows of W_M is a nonsingular Vandermonde matrix. So the choice $d_{\ell} = \ell, \ell = 1, 2, \dots, L-1$, leads to a nonsingular Vandermonde $A(\omega)$. Unique solutions of $\{F_{\ell}(\omega)\}$ can be obtained from (2). The theorem below follows.

Theorem 2 [8]. A 1D discrete-time U(M, L) signal x(n) can be recovered from its first L polyphase components, x(Mn), x(Mn+1), ..., x(Mn+L-1).

4. TWO-DIMENSIONAL SAMPLING AND RECONSTRUCTION

For 2D discrete-time signals, aliasfree(M) property, aliasfree(M) zone and U(M, L) can be defined in the same manner, where M is a 2×2 nonsingular integer matrix. Similar to 1D case, a necessary and sufficient condition for reconstructing 2D U(M, L) signals can be derived.

Lemma 3. A 2D discrete-time $U(\mathbf{M}, L)$ signal $x(\mathbf{n})$ can be recovered from L of its polyphase components if and only if the equation to follow has a solution for every $\omega \in Supp\{X(\omega)\}$.

$$\mathbf{A}(\omega) \begin{bmatrix} F_0(\omega) & F_1(\omega) & \dots & F_{L-1}(\omega) \end{bmatrix}^T = T\mathbf{e}_0, \quad (3)$$

where the matrix $A(\omega)$ is given by

$$[\mathbf{A}(\boldsymbol{\omega})]_{0\ell} = [\mathbf{A}(\boldsymbol{\omega})]_{\ell 0} = 1, \quad 0 \le \ell \le L - 1$$
$$[\mathbf{A}(\boldsymbol{\omega})]_{i\ell} = e^{-j2\pi\beta_i^T(\boldsymbol{\omega})\mathbf{M}^{-1}\mathbf{d}_\ell}, \quad 1 < i, \ell < L - 1 \quad \blacksquare$$

In 1D case, we can always choose d_{ℓ} such that $A(\omega)$ is a nonsingular Vandermonde matrix for every $\omega \in Supp\{X(\omega)\}$. However, it is not always possible to do so in 2D case. In fact, the above equation may not have a solution for some $\omega \in Supp\{X(\omega)\}$ and hence $x(\mathbf{n})$ can not be reconstructed from L of its polyphase components. To explain this, we take a closer look at $A(\omega)$.

The matrices $\mathbf{A}(\omega)$ and $\mathbf{W}^{(g)}$. It can be verified that $\mathbf{A}(\omega)$ is a $L \times L$ sub-matrix of a $J_{\mathbf{M}} \times J_{\mathbf{M}}$ matrix $\mathbf{W}^{(g)}$, called the generalized DFT matrix (possibly with some row and column exchanges), where $J_{\mathbf{M}} = |\det \mathbf{M}|$. The elements of $\mathbf{W}^{(g)}$ are given by

$$[\mathbf{W}^{(g)}]_{in} = e^{-j 2\pi \mathbf{k}_i^T \mathbf{M}^{-1} \mathbf{m}_n}, \mathbf{m}_n \in \mathcal{N}(\mathbf{M}), \mathbf{k}_i \in \mathcal{N}(\mathbf{M}^T),$$

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where $\mathcal{N}(\mathbf{M})$ denotes the set of integer vectors of the form $\mathbf{Mx}, \mathbf{x} \in [0 \ 1)^2$. Let Λ be the Smith form of \mathbf{M} [6], $\Lambda = \begin{bmatrix} \lambda_0 & 0 \\ 0 & \lambda_1 \end{bmatrix}$. It can be shown that when \mathbf{m}_n and \mathbf{k}_i are properly ordered, $\mathbf{W}^{(g)} = \mathbf{W}_{\lambda_0} \otimes \mathbf{W}_{\lambda_1}$, where \mathbf{W}_{λ} denotes a $\lambda \times \lambda$ DFT matrix and \otimes denotes the Kronecker product. The Kronecker product of two matrices \mathbf{A} and \mathbf{B} is defined as

$$\underbrace{\mathbf{A}}_{I\times K} \otimes \underbrace{\mathbf{B}}_{J\times L} = \underbrace{\begin{bmatrix} a_{0,0}\mathbf{B} & \dots & a_{0,K-1}\mathbf{B} \\ \vdots & \dots & \vdots \\ a_{I-1,0}\mathbf{B} & \dots & a_{I-1,K-1}\mathbf{B} \end{bmatrix}}_{IJ\times KL}.$$

Unlike 1D DFT matrices, $W^{(g)}$ is not Vandermonde in general; nor are its $L \times L$ sub-matrices obtained by retaining the first L columns and some Lrows. The natural question to ask next is whether a particular set of $\{d_{\ell}\}$ will make $A(\omega)$ nonsingular for all $\omega \in Supp\{X(\omega)\}$. In terms of the generalized DFT matrix $\mathbf{W}^{(g)}$, the question can be recast as follows: can we find L columns of $W^{(g)}$ such that for arbitrarily chosen L rows of $W^{(g)}$, the resulting sub-matrix is always nonsingular? The answer is unfortunately, no. Although for every $\omega_0 \in Supp\{X(\omega)\}$, there always exist $\{d_{\ell}\}$ such that $A(\omega_0)$ is nonsingular, the same d_{ℓ} may yield a singular $A(\omega_1)$ for a different frequency vector ω_1 . To follow is an example which demonstrates that there are cases when (3) is not solvable with frequency independent $\{\mathbf{d}_{\ell}\}$.

Example 1. Consider a discrete-time 2D $U(\mathbf{M}, 2)$ signal $x(\mathbf{n})$, where $\mathbf{M} = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}$ and $|\det \mathbf{M}| = 4$. The four vectors in $\mathcal{N}(\mathbf{M})$ are

$$\mathbf{n}_0 = \begin{bmatrix} 0\\0 \end{bmatrix}$$
, $\mathbf{n}_1 = \begin{bmatrix} 1\\0 \end{bmatrix}$, $\mathbf{n}_2 = \begin{bmatrix} 0\\1 \end{bmatrix}$, $\mathbf{n}_3 = \begin{bmatrix} 1\\1 \end{bmatrix}$,

Order $\mathbf{k}_i \in \mathcal{N}(\mathbf{M}^T)$ by letting $\mathbf{k}_i = \mathbf{n}_i$, then $\mathbf{W}^{(g)}$ is

The support of $X(\omega)$, as shown in Fig. 5, consists of two aliasfree(**M**) zones, S_0 and S_1 with S_1 being the union of three regions R_0 , R_1 , and R_2 . Because L = 2, we only have one beta function, $\beta(\omega)$. Observe that

$$eta(\omega) = egin{cases} \mathbf{k}_1, & \omega \in R_0 \ \mathbf{k}_2, & \omega \in R_1 \ \mathbf{k}_3, & \omega \in R_2. \end{cases}$$

So for $\omega \in R_0$, $A(\omega)$ is a sub-matrix of $W^{(g)}$ obtained by first keeping the 0th and 1st rows of $W^{(g)}$ and two columns $W^{(g)}$. That is, $A(\omega)$ is a 2 × 2 sub-matrix of $\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \end{bmatrix}$ obtained by keeping two columns. As to which two columns depends on the choice of d_1 . For $\omega \in R_0$, (3) has a solution only if d_1 is n_1 or n_3 . If (3) has a solution in each R_i , then

$$\mathbf{d}_1 = \begin{cases} \mathbf{n}_1 \text{ or } \mathbf{n}_3, & \omega \in R_0 \\ \mathbf{n}_2 \text{ or } \mathbf{n}_3, & \omega \in R_1 \\ \mathbf{n}_1 \text{ or } \mathbf{n}_2, & \omega \in R_2. \end{cases}$$

There is no common solution of d_1 for the three regions; (3) does not have a solution for all ω in the support of $X(\omega)$. Therefore x(n) cannot be reconstructed from two of its polyphase components.

A subclass of $U(\mathbf{M}, L)$. Although it is not always possible to reconstruct a $U(\mathbf{M}, L)$ signal from L of its polyphase components, it is always possible to do so when the Smith form Λ of $\mathbf{M} = \mathbf{U}\Lambda\mathbf{V}$ is

$$\mathbf{\Lambda} = \begin{bmatrix} \mathbf{1} & \mathbf{0} \\ \mathbf{0} & J_{\mathbf{M}} \end{bmatrix},$$

In this case the generalized DFT matrix $\mathbf{W}^{(g)}$ is the $J_{\mathbf{M}} \times J_{\mathbf{M}}$ DFT matrix $\mathbf{W}_{J_{\mathbf{M}}}$. Similar to the reconstruction of 1D U(M, L) signals, choose $\mathbf{d}_{\ell} = \ell \mathbf{U} \begin{bmatrix} 0 & 1 \end{bmatrix}^T$. Then the matrix $\mathbf{A}(\omega)$ will be nonsingular for all $\omega \in Supp\{X(\omega)\}$ and by (3) we can invert $\mathbf{A}(\omega)$ to obtain the synthesis filters.

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Fig. 2 Illustration of Lth order periodically nonuniform sampling.



Fig. 3 Reconstruction by using second order periodically nonuniform sampling.



Fig. 4 Periodically nonuniform sampling and reconstruction in discrete-time case.



