

## CG-FFT for Nonuniform Inverse Fast Fourier Transforms (NU-IFFT's)

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### I. Introduction

Fast Fourier Transform (FFT) [1] has become a popular computational tool in the study of physics and engineering since its development in 1960's. The condition of using FFT algorithm is that the data acquisition must be equispaced. However, in some cases, the data acquisition is not uniformly spaced. Dutt and Rokhlin [2] and Beylkin [3] presented some algorithms for the problems of FFT for nonuniform data. Their idea in solving these problems by a regular FFT is to approximate a function  $F(x) = e^{-bx^2} e^{icx}$ , by a small number of uniform points on the unit circle. Recently, a more accurate algorithm was developed by using the regular Fourier matrices for the nonuniform forward FFT (NUFFT) [5, 6]. Instead of  $F(x)$ , it uses  $F(j) = s_j e^{i2\pi cj/N}$ , where  $j = -N/2, \dots, N/2 - 1$ , and  $s_j$  are chosen to minimize the approximation error. In this work, we use the conjugate-gradient method and regular FFT (CG-FFT) together with the NUFFT algorithm to develop an accurate algorithm for nonuniform inverse fast Fourier transform (NU-IFFT).

### II. Formulation

Our goal is to develop fast algorithms to find the forward and inverse solutions for the following two problems:

$$f_j = \sum_{k=0}^{N-1} \alpha_k e^{i\omega_k 2\pi j/N}, \quad \text{for } \omega_k \in [-N/2, N/2], \quad j = -N/2, \dots, N/2 - 1 \quad (1)$$

$$g_j = \sum_{k=-N/2}^{N/2-1} \beta_k e^{ikx_j}, \quad \text{for } x_j \in [-\pi, \pi], \quad j = 0, \dots, N - 1 \quad (2)$$

Note that unlike the regular FFT,  $\omega_k$  and  $x_j$  in (1) and (2) are nonuniform.

#### A. The NUFFT Algorithm

The principle of using the regular Fourier matrices [5, 6] to solve (1) and (2) is as follows. For an integer  $m \geq 2$ , let  $w = e^{i2\pi/mN}$ ,  $q$  be an even positive integer,  $s_j$  ( $j = -N/2, \dots, N/2 - 1$ ) be positive numbers, and  $c$  be a real number. Here we approximate  $w^{jmc}$  by interpolation,

$$s_j w^{jmc} = \sum_{k=[mc]-q/2}^{[mc]+q/2} X_{k-[mc]}(c) w^{jk}, \quad \text{for } j = -N/2, \dots, N/2 - 1 \quad (3)$$

where  $X_{k-[mc]}$  are unknowns to be solved. From [5, 6], the least-squares solution is  $X(c) = F^{-1}a(c)$ , where  $F = A^H A$ , and  $A_{jk} = w^{jk}$  ( $j = -N/2, \dots, N/2 - 1, k = [mc] - q/2, \dots, [mc] + q/2$ ).

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$q/2$ ), and  $A^\dagger$  denotes the complex conjugate transpose of  $A$  matrix. If we choose the cosine accuracy factors  $s_j = \cos \frac{\pi j}{Nm}$ , a closed-form solution can be found for  $a_k(c)$

$$a_k(c) = i \sum_{\gamma=-1,1} \frac{\sin[\frac{\pi}{2m}(2k - \gamma - q - 2(mc - [mc]))]}{1 - e^{i\frac{\pi}{Nm}(2(mc - [mc]) + q - 2k + \gamma)}} \quad (4)$$

The NUFFT algorithm for (1) has been given in [5, 6]. By modifying it, we have the following NUFFT algorithm for (2):

- 1) Compute  $X_l(x_j N/2\pi)$  by  $X(c) = F^{-1}a(c)$ , for  $l = 0, \dots, q$  and  $k = 0, \dots, N - 1$ ;
- 2) Calculate Fourier coefficients  $u_k = \beta_k s_k^{-1}$ ;
- 3) Use uniform FFT to evaluate

$$U_l = \sum_{k=-N/2}^{N/2-1} u_k e^{i2\pi kl/mN}, \quad \text{for } l = -mN/2, \dots, mN/2 - 1;$$

- 4) Scale the values to obtain the approximate NUFFT

$$\tilde{g}_j = \sum_{l=-q/2}^{q/2} X_l(x_j N/2\pi) U_{l+[mNx_j/2\pi]}, \quad \text{for } j = 0, \dots, N - 1.$$

The asymptotic number of arithmetic operations in this algorithm is  $O(mN \log_2 N)$ , where  $m \ll N$ . Usually we choose  $m = 2$  and  $q = 8$ .

### B. The NU-IFFT Algorithm

Defining  $D_{jk} = e^{ikx_j}$ , we rewrite (2) as

$$g_j = \sum_{k=-N/2}^{N/2-1} \beta_k D_{jk}, \quad \text{for } j = 0, \dots, N - 1 \quad (5)$$

From elementary matrix identities we see that

$$(D^\dagger D)D^{-1} = D^\dagger \quad \text{and} \quad D^{-1} = (D^\dagger D)^{-1}D^\dagger \quad (6)$$

Combining (5) and (6), we have

$$\tilde{\beta} = (D^\dagger D)^{-1}D^\dagger g \quad (7)$$

Equation (7) is further divided into two consecutive steps for its solution:

- 1) Utilize the NUFFT algorithm for (1) to solve both  $D^\dagger D = \sum_{k=0}^{N-1} e^{-i(j-l)\pi k}$  ( $j, l = -N/2, \dots, N/2 - 1$ ), and  $y_k = \sum_{j=0}^{N-1} D_{kj}^* g_j$  ( $k = -N/2, \dots, N/2 - 1$ );
- 2) Use the following CG-FFT method to solve  $\tilde{\beta} = (D^\dagger D)^{-1}y = B^{-1}y$ .

In the iterative procedure of the conjugate gradient method for  $\tilde{\beta} = B^{-1}y$ ,  $z = By$  is calculated many times and can be expedited by using the regular FFT. Recognizing that

$B_{j,l} = \sum_{k=0}^{N-1} e^{-i(j-l)x_k}$  is a Toeplitz matrix ( $j, l = -N/2, \dots, 0, \dots, N/2 - 1$ ), and by denoting  $B_{j,l}$  as 1-D form  $B_{j,l} = b(j-l)$ , we obtain

$$z_j = \sum_{l=-N/2}^{N/2-1} b(j-l)y(l), \text{ for } j = -N/2, \dots, N/2 - 1 \quad (8)$$

Applying the convolution theorem, (8) can be derived as:

$$z_j = \{\text{FFT}^{-1}[\text{FFT}(b)\text{FFT}(y)]\}_j \quad (9)$$

If  $k$  iterations are required for the CG-FFT method to converge to an accurate solution, the total number of complex multiplications is proportional to  $O[kN(2\log_2 2N + \log_2 N + 1)]$ . Similarly, we can obtain the NU-IFFT algorithm for (1).

### III. Numerical Results

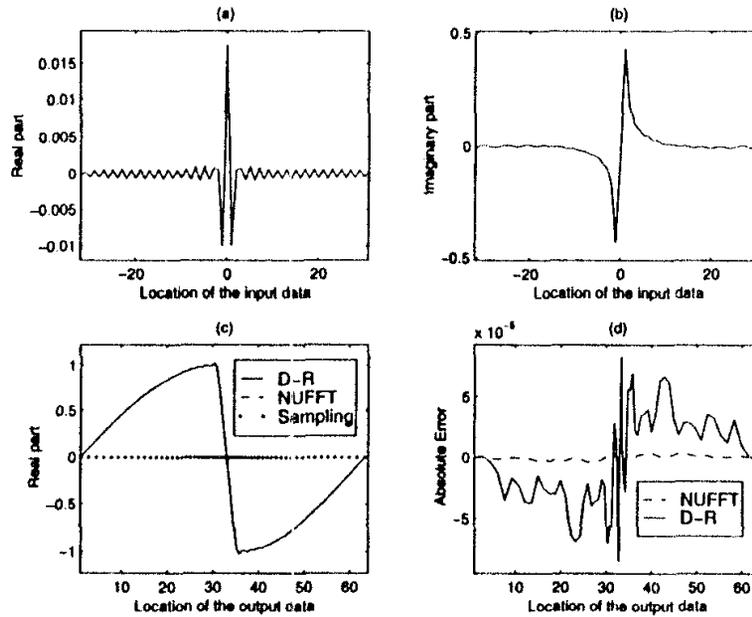
We compare Dutt-Rokhlin's algorithm and our algorithm with cosine accuracy factors, and  $m = 2$  and  $q = 8$ . Figures 1 and 2 are the examples of the NUFFT and NU-IFFT algorithms for (2). Figures 1(a) and 1(b) show the real and imaginary parts of input data  $\beta_k$ , respectively. Figure 1(c) shows the spatial distribution of output data  $g_j$ . Figure 1(d) displays the absolute error from our NUFFT and that from Dutt-Rokhlin's [2]. Quantitatively, the  $E_2$  and  $E_\infty$  defined in [2] are  $E_2 = 2.713 \times 10^{-6}$  and  $E_\infty = 1.8511 \times 10^{-6}$  for our algorithm, and  $E_2 = 5.238 \times 10^{-5}$  and  $E_\infty = 3.535 \times 10^{-5}$  for the Dutt-Rokhlin's algorithm. Figure 2(a) shows inverse input data  $g_j$ . Figures 2(b) and 2(c) represent the real and imaginary parts of output data  $\beta_k$  for the inversion, respectively. Figure 1(d) indicates the absolute error from our NU-IFFT and that from Dutt-Rokhlin's [2]. Quantitatively,  $E_2 = 3.156 \times 10^{-6}$  and  $E_\infty = 2.390 \times 10^{-7}$  for our algorithm, and  $E_2 = 2.3069 \times 10^{-6}$  and  $E_\infty = 1.549 \times 10^{-6}$  for the Dutt-Rokhlin's algorithm. From these results, the accuracy of our algorithms is about one order of magnitude higher than Dutt-Rokhlin's.

### IV. Conclusions

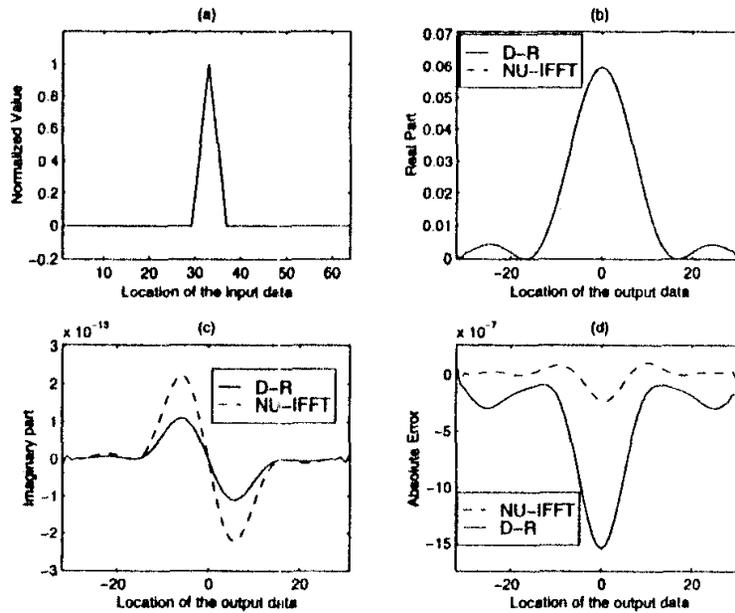
Based on the CG-FFT method and the NUFFT algorithm, a new nonuniform inverse fast Fourier transform (NU-IFFT) algorithm is developed for nonuniform data. With a comparable complexity of  $O(N\log_2 N)$ , this algorithm is much more accurate than the previously reported results since it is optimal in the least squares sense.

### References

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**Figure 1.** The NUFFT algorithm for (2). (a) Real part of input data (b) Imaginary part of input part. (b) Output data. (c) Absolute error of output data.



**Figure 2.** The NU-IFFT algorithm for (2) by CG-FFT method. (a) Input data. (b) Real part of output data. (c) Imaginary part of output part. (d) Absolute error of output data.