

# FFT-based Algorithms for 3-D Space Charge Calculation

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## Outline

- MOEVE PIC Tracking package
- Space charge effect calculation
- FFT-based algorithms for 3-D space charge calculation
- Advantages and disadvantages
- Space charge including image charge
- Tracking comparison with ASTRA

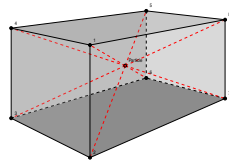
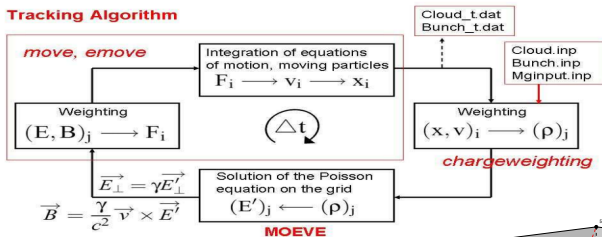
## MOEVE PIC Tracking package

Gains: [Enhanced by A. Marković, G. Pöplau]

- Simulate the interaction between a positron beam and electron clouds
- Simulate the interaction between an electron bunch and ion clouds
- 3D particle tracking including space charge effect with rectangular (elliptical transverse) cross-section



## Tracking algorithm of MOEVE PIC



- cubic computational domain
- Particle-mesh method
- (Non)Eqidistant grid in each axis

<sup>0</sup>PhD.Thesis of A. Markovik

## Poisson Equation

The potential  $\varphi$  of the bunch (distribution of macro particles) is calculated from Poisson equation:

$$-\Delta\varphi = \frac{\rho}{\epsilon_0} \quad \text{in } \Omega \subset \mathbb{R}^3$$

$\varphi$	potential
$\rho$	space charge density
$\Omega$	computational domain

$$E = -\text{grad}\varphi$$

Boundary conditions:

Dirichlet:

$$\varphi = 0 \quad \text{on } \partial\Omega$$

Neumann:

$$\frac{\partial\varphi}{\partial n} = \varphi_{\partial\Omega}$$

periodic:

$$\varphi_{-\partial\Omega} = \varphi_{\partial\Omega}$$

open:

$$\frac{\partial\varphi}{\partial n} + \frac{1}{r}\varphi = 0$$

free and mixed boundary conditions.



## Numerical methodologies sketch

- Multigrid method:

MG (multigrid), CG(conjugate gradient), MGCG , SOR (successive over-relaxation ), BiCG, BiCGSTAB, PCBiCG

- FFT-based method:

Spectral method

GF (Green's function), IGF (Integrated Green's function)

## FFT-based method: Spectral Method

**Discretization** by finite difference provides the following system of linear equations:

$$Lu = f$$

$$L = \begin{pmatrix} -2 & \boxed{N} & 0 & \dots & \boxed{P} \\ 1 & -2 & 1 & & \vdots \\ \vdots & \ddots & \ddots & \ddots & \vdots \\ 0 & & 1 & -2 & 1 \\ \boxed{P} & 0 & \dots & \boxed{N} & -2 \end{pmatrix}$$

Dirichlet:	$P=0$ ,	$N=1$
Neumann:	$P=0$ ,	$N=2$
Periodic:	$P=1$ ,	$N=1$

## Theorem (Spectral Decomposition, Charles Van Loan, 1992)

If we define the discrete Laplace matrices for the above five boundary conditions as  $L$ , there is an invertible matrix  $V$  for each boundary condition, such that

$$V^{-1}LV = \Lambda,$$

here  $\Lambda$  is the spectral matrix, which is diagonal.

Dirichlet-Dirichlet:  $V=DST, \quad \Lambda_{jj} = 4 \sin^2\left(\frac{j\pi}{2n}\right), \quad j = 1 : n - 1.$

Dirichlet-Neumann:  $V=DST-II, \quad \Lambda_{jj} = 4 \sin^2\left(\frac{2j-1}{4n}\pi\right), \quad j = 1 : n.$

Neumann-Dirichlet:  $V=DCT-II, \quad \Lambda_{jj} = 4 \sin^2\left(\frac{2j-1}{4n}\pi\right), \quad j = 1 : n.$

and

Neumann-Neumann:  $V=DCT, \quad \Lambda_{jj} = 4 \sin^2\left(\frac{j\pi}{2n}\right), \quad j = 0 : n.$

Periodic:  $V=IDFT, \quad \Lambda_{jj} = 4 \sin^2\left(\frac{j\pi}{2n}\right), \quad j = 0 : n - 1.$



with this factorization of  $L$ , we solve the linear system  $Lu = f$  as:

$$\begin{aligned}Lu &= f \\ \underline{V V^{-1} L V^{-1}} u &= f \\ \Lambda V^{-1} u &= V^{-1} f \\ V^{-1} u &= \Lambda^{-1} V^{-1} f \\ \hat{u} &= \Lambda^{-1} \hat{f} \\ u &= V \Lambda^{-1} V^{-1} f\end{aligned}$$

## 3D Situation

The resulting system of linear equations:

$$\left( \frac{1}{h_x^2} (I_{N_z} \otimes I_{N_y} \otimes L_{N_x}) + \frac{1}{h_y^2} (I_{N_z} \otimes L_{N_y} \otimes I_{N_x}) + \frac{1}{h_z^2} (L_{N_z} \otimes I_{N_y} \otimes I_{N_x}) \right) u = f$$

Eigenvalues:  $\Lambda_{x,y,z} = [\lambda_{i,j,k}]$ .  $\Rightarrow$  the linear system of equations can be written as:

$$\left( I \otimes I \otimes \frac{1}{h_x^2} \Lambda + I \otimes \frac{1}{h_y^2} \Lambda \otimes I + \frac{1}{h_z^2} \Lambda \otimes I \otimes I \right) (V^{-1} \otimes V^{-1} \otimes V^{-1}) u = (V^{-1} \otimes V^{-1} \otimes V^{-1}) f$$

$(V^{-1} \otimes V^{-1} \otimes V^{-1})$ : three dimensional Fourier-class transform. In the "frequency domain" we have:

$$\hat{u}_{i,j,k} = \frac{1}{\frac{\lambda_i}{h_x^2} + \frac{\lambda_j}{h_y^2} + \frac{\lambda_k}{h_z^2}} \hat{f}_{i,j,k}$$

## Standard Green's function (GF) method

Standard Green's function method:

$$\underbrace{\varphi(x, y, z)}_{\varphi(x_i, y_j, z_k)} = \frac{1}{4\pi\epsilon_0} \underbrace{\int \int \int}_{\Sigma \Sigma \Sigma} \underbrace{G(x, x', y, y', z, z')}_{G(x_i, x_{i'}, y_j, y_{j'}, z_k, z_{k'})} \underbrace{\rho(x', y', z')}_{\rho(x_{i'}, y_{j'}, z_{k'})} \underbrace{dx' dy' dz'}_{h_x h_y h_z}$$

With Green's function:

$$G(x, x', y, y', z, z') = \frac{1}{\sqrt{(x - x')^2 + (y - y')^2 + (z - z')^2}}$$

Discrete form:

$$\varphi(x_i, y_j, z_k)_{GF} \approx \frac{h_x h_y h_z}{4\pi\epsilon_0} \sum_{i'=1}^{N_x} \sum_{j'=1}^{N_y} \sum_{k'=1}^{N_z} G(x_i, x_{i'}, y_j, y_{j'}, z_k, z_{k'}) \rho(x_{i'}, y_{j'}, z_{k'})$$

## Integrated Green's function (IGF) method

Integrated Green's function method [Ji Qiang, 2006]:

$$\underbrace{\varphi(x, y, z)}_{\varphi(x_i, y_j, z_k)} = \frac{1}{4\pi\epsilon_0} \underbrace{\int \int \int G(x, x', y, y', z, z') \rho(x', y', z') dx' dy' dz'}_{\sum \sum \sum \tilde{G}(x_i, x_{i'}, y_j, y_{j'}, z_k, z_{k'})}$$

With integrated Green's function:

$$\tilde{G}(x_i, x_{i'}, y_j, y_{j'}, z_k, z_{k'}) = \int_{x_{i'} - h_x/2}^{x_{i'} + h_x/2} \int_{y_{j'} - h_y/2}^{y_{j'} + h_y/2} \int_{z_{k'} - h_z/2}^{z_{k'} + h_z/2} G(x_i, x', y_j, y', z_k, z') dx' dy' dz'$$

Discrete form:

$$\varphi(x_i, y_j, z_k)_{IGF} \approx \frac{1}{4\pi\epsilon_0} \sum_{i'=1}^{N_x} \sum_{j'=1}^{N_y} \sum_{k'=1}^{N_z} \rho(x_{i'}, y_{j'}, z_{k'}) \tilde{G}(x_i, x_{i'}, y_j, y_{j'}, z_k, z_{k'})$$

## Extensions:

[R. W. Hockney and J. W. Eastwood, 1994]

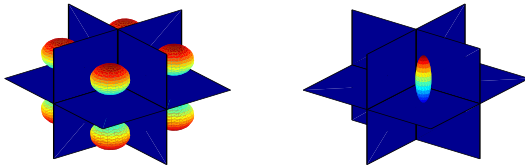


Figure: The expansion of Green's function (Left side) and the expansion of charge density (Right side).

The expansion of Green's function is called **real even** symmetry expansion.

Example:  $[x_0, x_1, x_2, x_3] \rightarrow [x_0, x_1, x_2, x_3, x_2, x_1]$ .

The expansion of charge density, **zeros** expansion:

$[\rho_0, \rho_1, \dots, \rho_{N-1}, \rho_N] \rightarrow [\rho_0, \rho_1, \dots, \rho_N, 0, \dots, 0]$ .

$$\varphi(x_i, y_j, z_k) \approx \frac{h_x h_y h_z}{4\pi\epsilon_0} \sum_{i'=1}^{N_x} \sum_{j'=1}^{N_y} \sum_{k'=1}^{N_z} G(x_i - x_{i'}, y_j - y_{j'}, z_k - z_{k'}) \rho(x_{i'}, y_{j'}, z_{k'}) \implies$$

$$\varphi(x_i, y_j, z_k) \approx \frac{h_x h_y h_z}{4\pi\epsilon_0} \sum_{i'=1}^{2N_x-2} \sum_{j'=1}^{2N_y-2} \sum_{k'=1}^{2N_z-2} G_{\text{ex}}(x_i - x_{i'}, y_j - y_{j'}, z_k - z_{k'}) \rho_{\text{ex}}(x_{i'}, y_{j'}, z_{k'})$$

The **convolution theorem**, which is:

$$\mathfrak{F}\{G * \rho\} = \mathfrak{F}G \bullet \mathfrak{F}\rho$$

reveals that

$$\varphi_{i,j,k} = \frac{h_x h_y h_z}{4\pi\epsilon_0} \mathfrak{F}^{-1} \{ [\mathfrak{F}G_{\text{ex}}]_{i,j,k} \bullet^* [\mathfrak{F}\rho_{\text{ex}}]_{i,j,k} \}$$

or

$$\varphi_{i,j,k} = \frac{1}{4\pi\epsilon_0} \mathfrak{F}^{-1} \{ [\mathfrak{F}\tilde{G}_{\text{ex}}]_{i,j,k} \bullet^* [\mathfrak{F}\rho_{\text{ex}}]_{i,j,k} \}$$

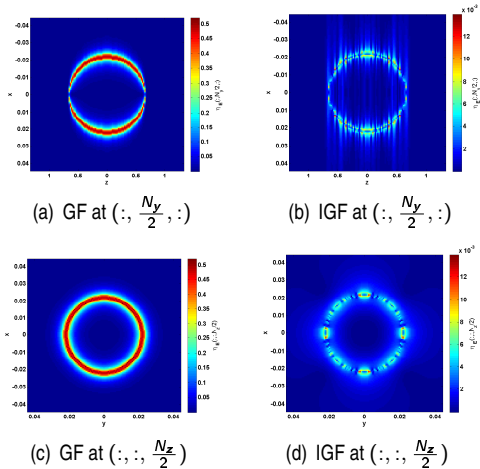
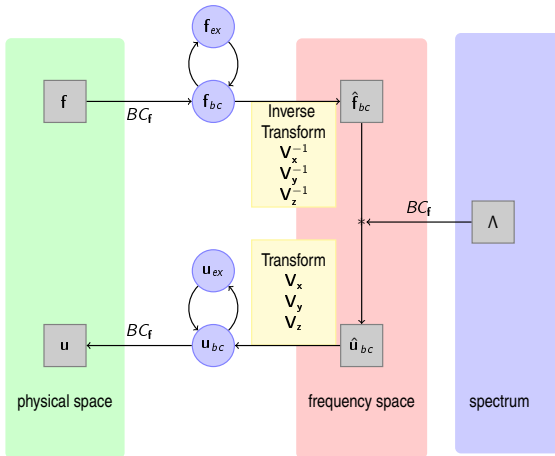


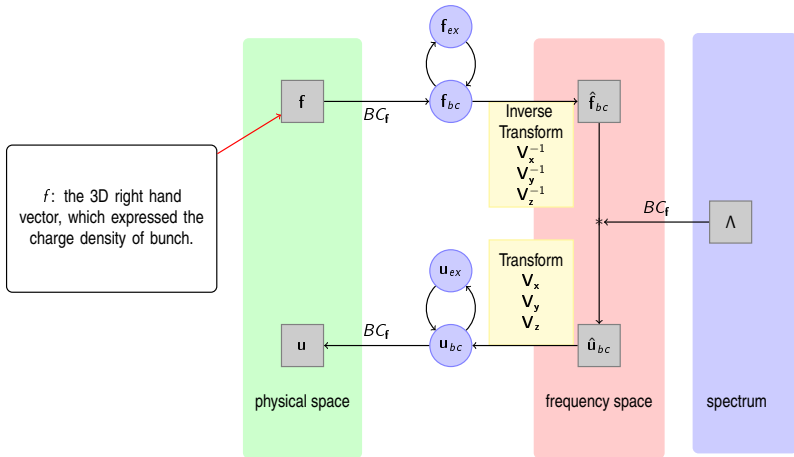
Figure : Comparison of  $\eta_E(i, j, k)$  for the two Green's function methods in Example 1. GF (a) (c), IGF (b) (d).

## FFT-based algorithms for 3-D space charge calculation

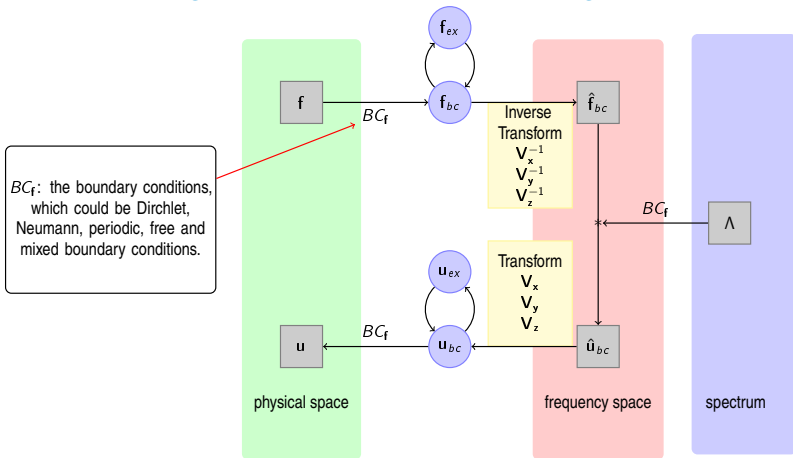




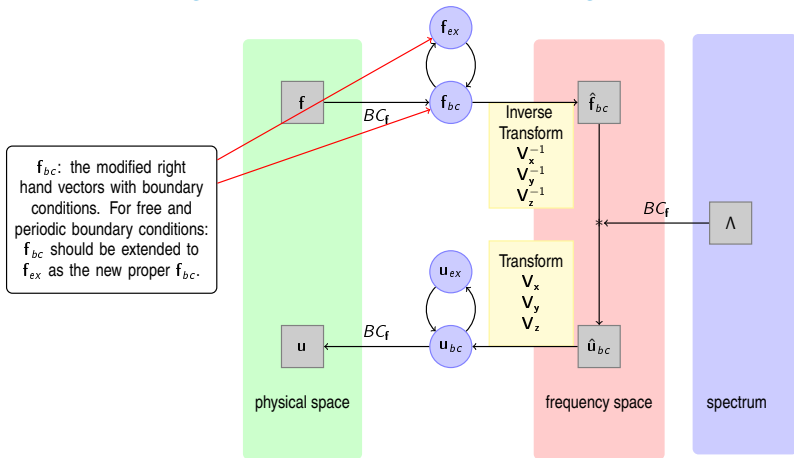
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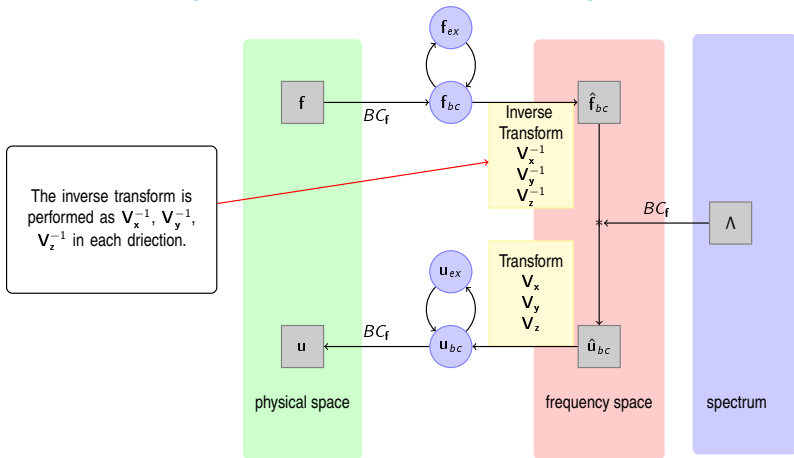
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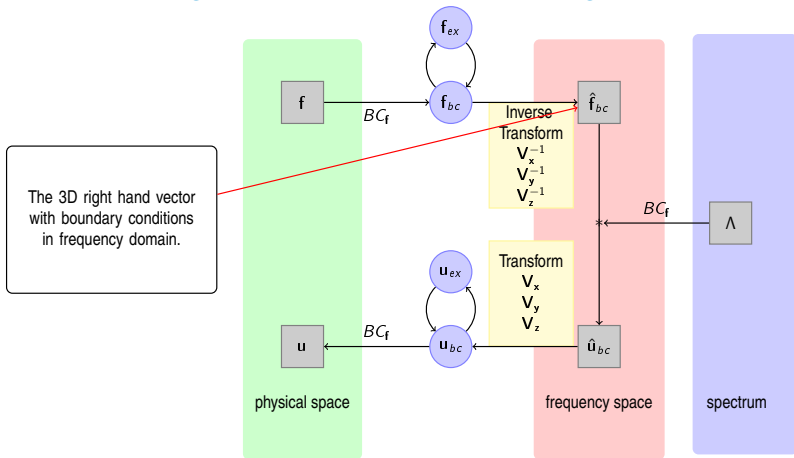
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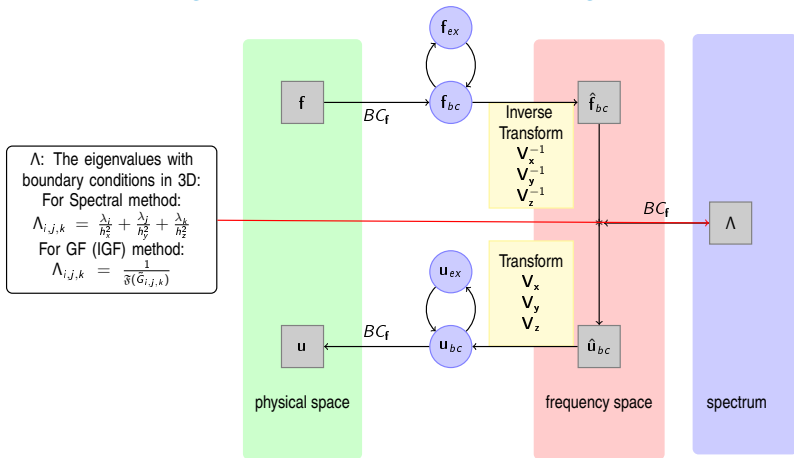
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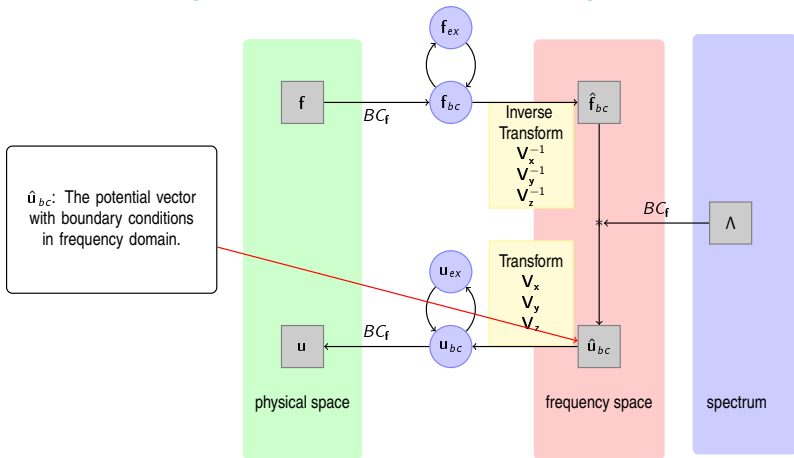
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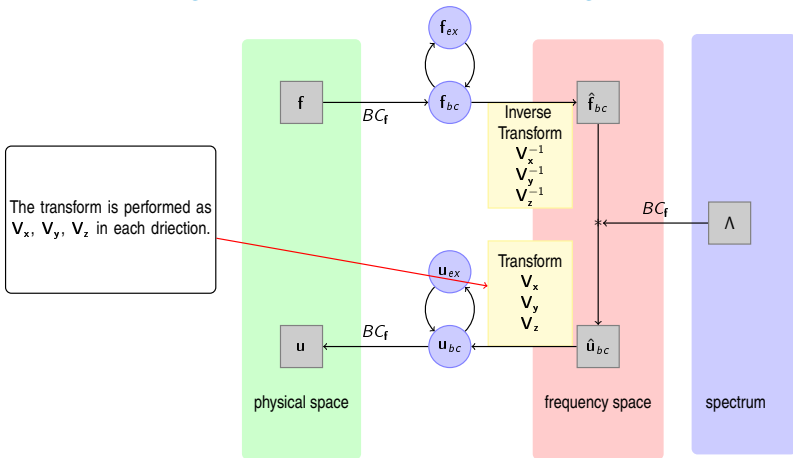
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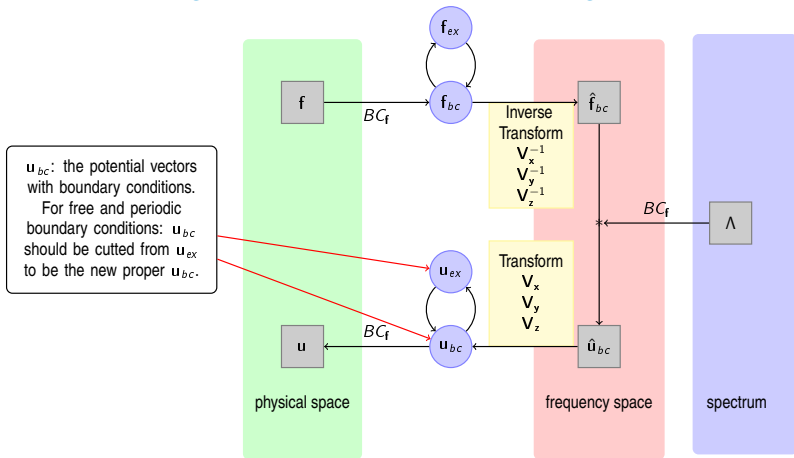


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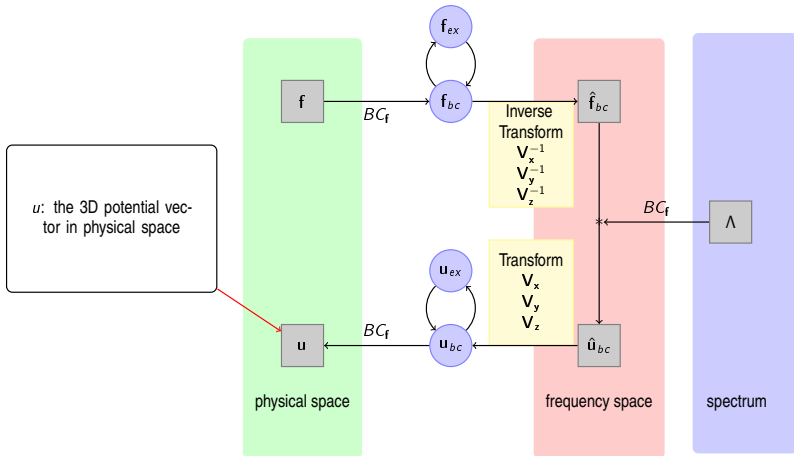




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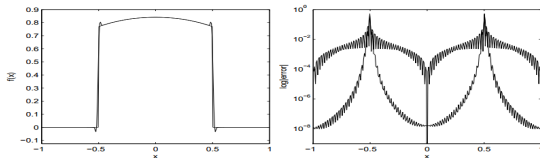
## FFT-based algorithms for 3-D space charge calculation



## Disadvantages of spectral methods and Green's function method

- Gibbs phenomenon

### Spectral Filter Example



- Still time-consuming due to many time step loops.
- High RAM computation

<sup>0</sup>[www.scottsarra.org](http://www.scottsarra.org)

## Disadvantages of spectral methods and Green's function method

- Gibbs phenomenon
- Still time-consuming due to many time step loops.

$N_x \times N_y \times N_z$	Spectral Solver	IGF Solver	MG Solver
$33 \times 33 \times 33$	0.035 s	0.115 s	0.077 s
$65 \times 65 \times 65$	0.282 s	1.191	1.087 s
$129 \times 129 \times 129$	2.317 s	11.084 s	9.568 s
$257 \times 257 \times 257$	20.030 s		51.324 s

- High RAM computation



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- Gibbs phenomenon
- Still time-consuming due to many time step loops.
- High RAM computation



## Advantages of spectral methods and Green's function method

- Together with multigrid methods in MOEVE, the spectral methods can deal with various boundary conditions
- The routine could be efficiently accelerated by parallel technology.
- There are rich theories and tools in spectral theory, that may use to optimize solution
- The spectral methods deal with not too long or short bunches
- The Integrated Green's function method is preferred for too long or short bunches



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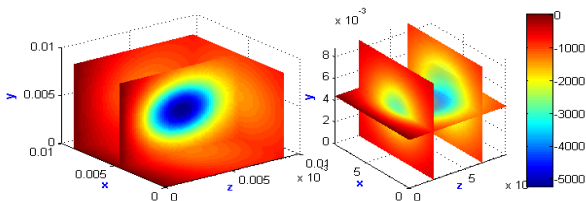


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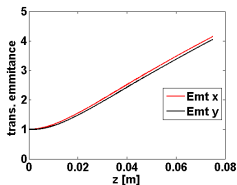
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## Space charge including image charge

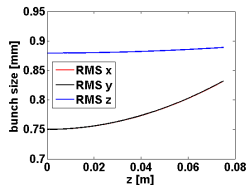
- double sized domain with periodic boundary conditions
- shifted Green's function



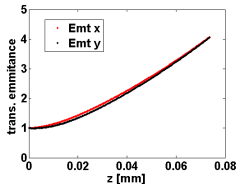
## Comparison with ASTRA: First result



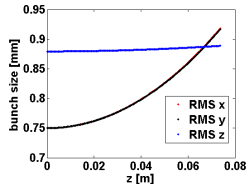
(a) Astra



(b) Astra

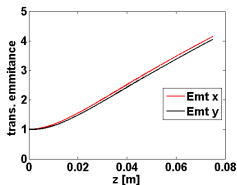


(c) Moeve far bunch

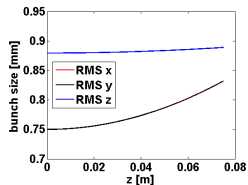


(d) Moeve far bunch

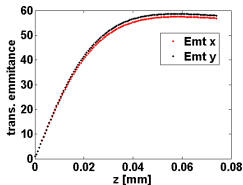
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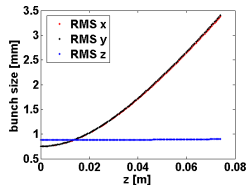
(e) Astra



(f) Astra



(g) Moeve near bunch



(h) Moeve near bunch



## Restrictions of the 3D FFT algorithm in ASTRA

- The 3D algorithm does not provide special features for the emission of particles from the cathode in its present form
- Image charge forces cannot be included
- During the emission the complete grid is set up already after the first time step.
- The field description is restricted to the grid; hence the optional use of passive particles, which may travel outside of the grid, is limited

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<sup>0</sup>[ASTRA manual V3.0 2011]



## Outlook

- More efficient Poisson solver is needed:  
RIGF, CGF and other methods
- Deep study in spectral method
- Parallelization of the FFT-based Poisson solver  
GPU, Open MP
- Check and study the Poisson solver with applications  
Enhance to simulate beam emission.



Thank you !



## Real trigonometric transform

The Discrete Sine Transform(DST):  $y(1 : m - 1) = \text{DST}(x(1 : m - 1))$

$$y_k = \sum_{j=1}^{m-1} \sin\left(\frac{kj\pi}{m}\right)x_j$$

The Discrete Cosine Transform(DCT):  $y(0 : m) = \text{DCT}(x(0 : m))$

$$y_k = \frac{x_0}{2} + \sum_{j=1}^{m-1} \cos\left(\frac{kj\pi}{m}\right)x_j + \frac{(-1)^k x_m}{2}$$

The Discrete Sine Transform-II(DST-II):  $y(1 : m) = \text{DST-II}(x(1 : m))$

$$y_k = \sum_{j=1}^m \sin\left(\frac{k(2j-1)\pi}{2m}\right)x_j$$

The Discrete Cosine Transform-II(DCT-II):  $y(0 : m - 1) = \text{DCT-II}(x(1 : m))$

$$y_k = \sum_{j=0}^{m-1} \cos\left(\frac{k(2j+1)\pi}{2m}\right)x_j$$