

# MIC-GPU: High-Performance Computing for Medical Imaging on Programmable Graphics Hardware (GPUs)

## CT Reconstruction Pipeline Components

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## Course Schedule

- 1:30 – 2:00: Introduction
- 2:00 – 2:30: GPU architecture, programming model, and programming facilities
- 2:30 – 3:00: GPU programming examples (image processing)
- Coffee Break*
- 3:30 – 4:00: CT reconstruction pipeline components
- 4:00 – 4:30: GPU-acceleration of individual components
- 4:30 – 5:00: Various CT reconstruction pipelines, load balancing and load estimation
- 5:00 – 5:30: Reconstruction visualization and final remarks

## CT Reconstruction Pipeline

A CT reconstruction pipeline is typically composed of a number of serial components

### Example 1: Filtered Backprojection

- projection filtering
- backprojection
- post-weighting

### Example 2: Iterative 3D reconstruction in blocks

- backprojection of volume into set's views
- correction factor computation
- backprojection of correction factors
- post-weighting (normalization)

## Kernel-Centric Decomposition

We can consider each of these steps to be a SIMD kernel, as follows:

### Example 1: Filtered Backprojection

- projection filtering → *filtering kernel*
- backprojection → *backprojection kernel*
- post-weighting → *post-weighting kernel*

### Example 2: Iterative 3D reconstruction in blocks

- backprojection of volume into set's views → *projection kernel*
- correction factor computation → *correction factor kernel*
- backprojection of correction factors → *backprojection kernel*
- normalization → *normalization kernel*

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— vector operations      — projector with interpolation

## Kernel Scheduling

SIMD can only execute one kernel at a time

- this prohibits kernel overlap, even if mathematically correct
- we may merge kernels if targets are identical → this favors load balancing and the reduction of passes
- but recall that scattering to multiple targets is undesirable

Therefore a decomposition of a reconstruction pipeline into components is advisable

- develop an optimized kernel for each component
- overlap (=hide) the loading of data (if needed) with execution of a prior kernel (or within kernel)
- also optimize what platform to run the computations (CPU, GPU), but then consider transfer of data

## Popular CT Reconstruction Pipelines

We will discuss:

- analytical schemes (Feldkamp)
- various algebraic schemes (SART, SIRT)
- statistical schemes (EM, OS-EM)
- in terms of anatomical and metabolic (functional) CT
- for various beam geometries: parallel, fan, cone

The projection/backprojection is typically the most expensive operation

- it is part of every algorithm and application
- with variations in
  - beam geometry
  - modeling of tissue (attenuation, scattering) and detector effect
  - each is implemented with a dedicated kernel
  - each such kernel is loaded into the GPU on demand

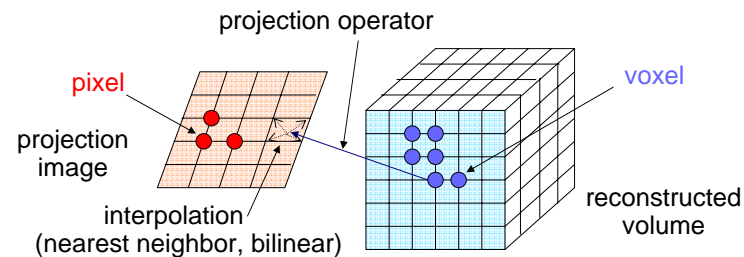
## Terminology

We shall discuss all material in terms of 3D reconstruction

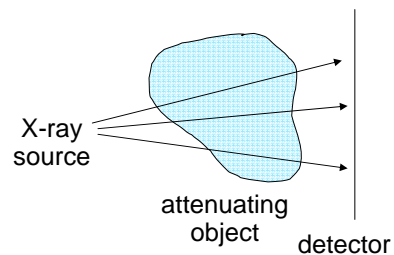
- the reduction to 2D slice reconstruction is straightforward

**Pixels:** the basis elements (point samples) of the projection image (the photon measurements)

**Voxels:** the basis elements (point samples) of the reconstruction volume (the attenuation densities or the tracer photon emissions)

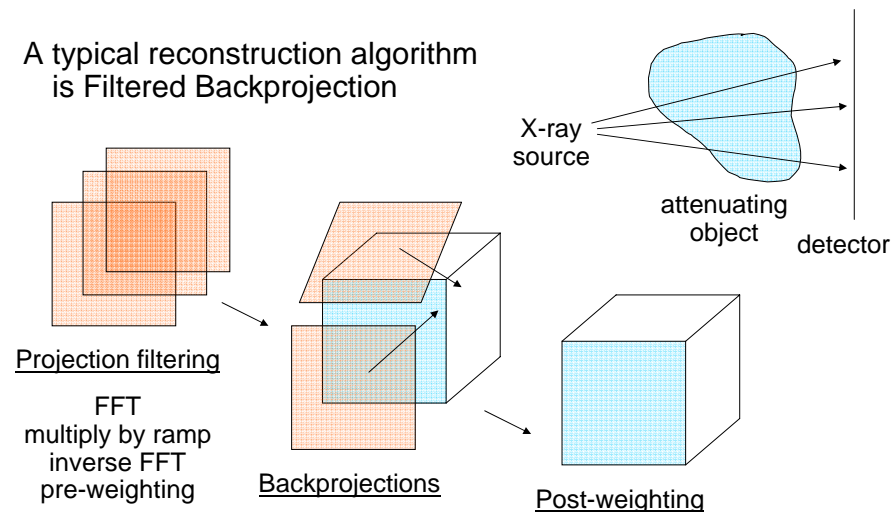


# Transmission CT



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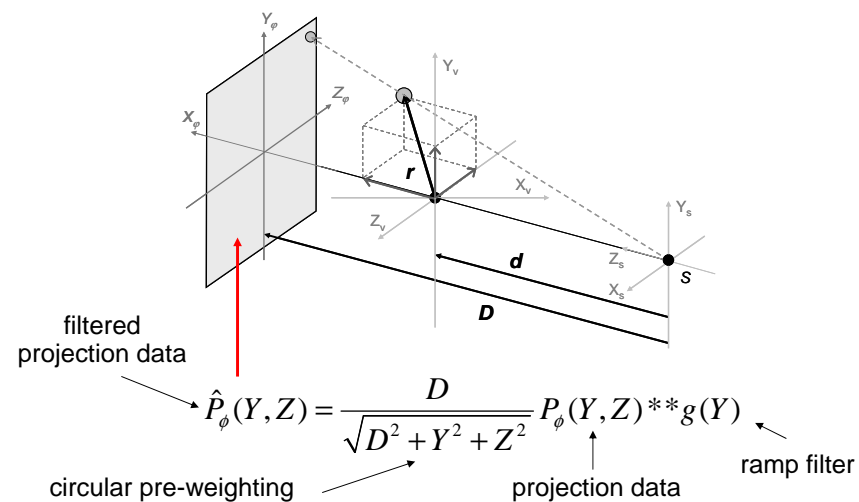
A typical reconstruction algorithm is Filtered Backprojection

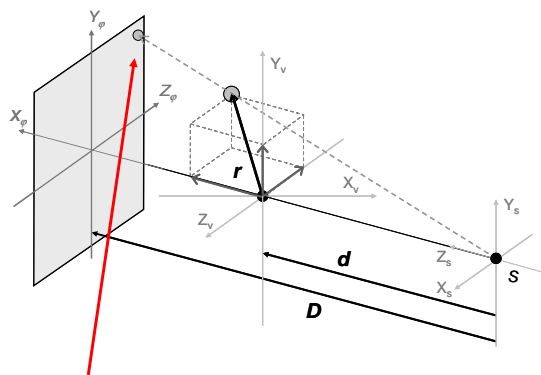


# Example

## Feldkamp-Davis-Kress (FDK) Cone-beam reconstruction

# FDK: Filtering

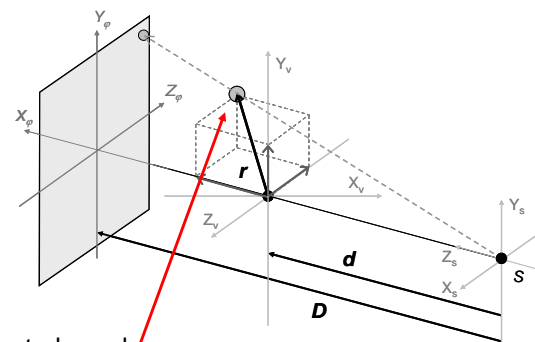




$$\hat{P}_\phi(\mathbf{r}) = \hat{P}_\phi(Y(\mathbf{r}), Z(\mathbf{r})), \quad Y(\mathbf{r}) = \frac{\mathbf{r} \cdot \mathbf{y}_\phi}{d + \mathbf{r} \cdot \mathbf{x}_\phi} D, \quad Z(\mathbf{r}) = \frac{\mathbf{r} \cdot \mathbf{z}_\phi}{d + \mathbf{r} \cdot \mathbf{x}_\phi} D$$

voxel → projection mapping

projection coordinates of mapped voxel



reconstructed voxel

$$f(\mathbf{r}) = \frac{1}{4\pi^2} \int_0^{2\pi} \frac{d^2}{(d + \mathbf{r} \cdot \mathbf{x}_\phi)^2} \hat{P}_\phi(\mathbf{r}) d\phi$$

accumulation for all projections

depth-weighting

Similar concepts apply for other analytical CT algorithms

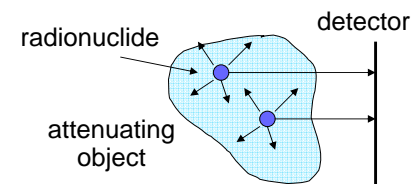
- modified FDK, multi-orbit cone-beam CT
- helical CT with exact and non-exact algorithms

Always a sequence of serial steps

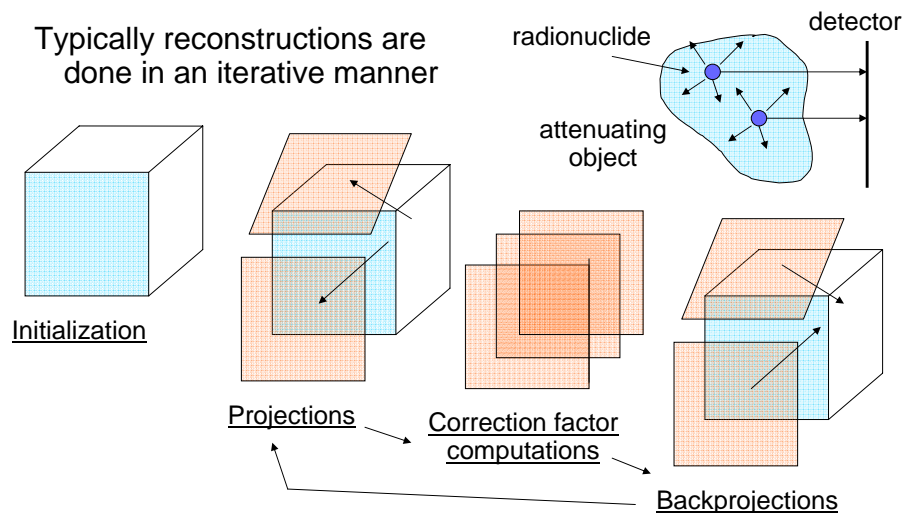
- projection filtering, possibly rebinning
- backprojection
- accumulation and weighting

Only backprojection (and rebinning) requires interpolation

- can make use of graphics texture-mapping facilities
- but can also be implemented as SIMD (fragment) programs (GPGPU)
- the remaining operations are straight vector arithmetic (GPGPU)



Typically reconstructions are done in an iterative manner



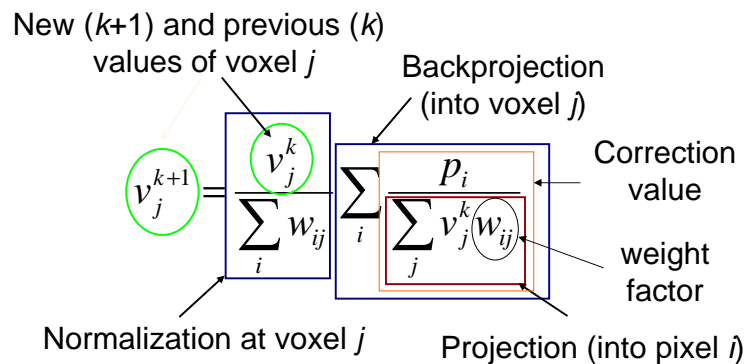
EM: volume correction after each projection

- tends to converge slowly

OS-EM: volume correction after a set of projections

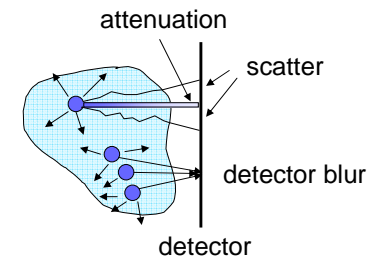
- converges faster

Maximizes the likelihood of the values of voxels  $j$ , given values at pixels  $i$



The weight factor  $w_{ij}$  can model various effects:

- interpolation filter factors (nearest neighbor, bilinear, Gaussian, Bessel, etc)
- detector geometric response (blurring due to off-angle photon contributions)
- photon attenuation (requires an attenuation map  $\mu$  obtained via transmission CT)
- photon scattering (requires a gradient map, typically the transmission CT reconstruction)



## Forward projection (front-to-back):

- the energy arriving at a detector pixel is:  $p = \int_{s=0}^l c(s) e^{-\int_{t=0}^s \mu(t) dt} ds$
- in discrete terms:

$$p \approx \sum_{i=0}^{L/\Delta s} c(i\Delta s) e^{-\sum_{j=0}^{i-1} \mu(j\Delta s)} = \sum_{i=0}^{L/\Delta s} c(i\Delta s) \prod_{j=0}^{i-1} e^{-\mu(j\Delta s)}$$

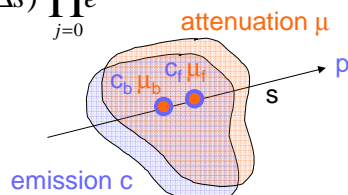
- using a Taylor series approximation:

$$p \approx \sum_{i=0}^{L/\Delta s} c(i\Delta s) \prod_{j=0}^{i-1} (1 - \mu(j\Delta s))$$

note: all values are normalized to [0,1]

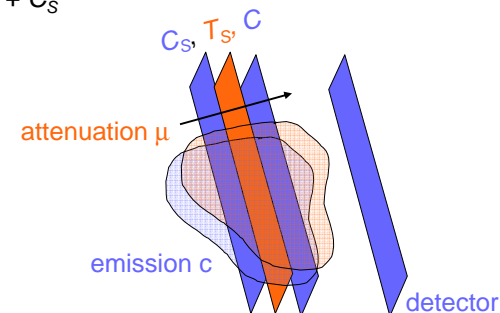
- as a recursive equation:  $c_f = c_f + c_b t_f$      $t_f = t_f (1 - \mu_b) = t_f t_b$
- equivalent back-to-front compositing:

$$c_b = c_b (1 - \mu_f) + c_f = c_b t_f + c_f$$



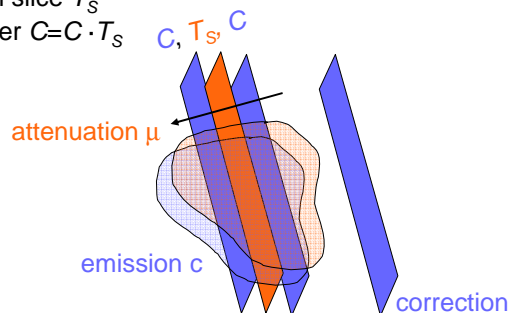
## Forward projection (back-to-front traversal):

- emission buffer  $C=0$
- step from back to front, at each step:
  - interpolate emission slice  $C_s$  and attenuation slice  $T_s$
  - composite  $C = C \cdot T_s + C_s$



## Backprojection (front-to-back traversal):

- initialize correction buffer  $C$
- step from front to back, at each step:
  - spread (and add)  $C$  into emission volume affected by slice
  - interpolate attenuation slice  $T_s$
  - update correction buffer  $C = C \cdot T_s$

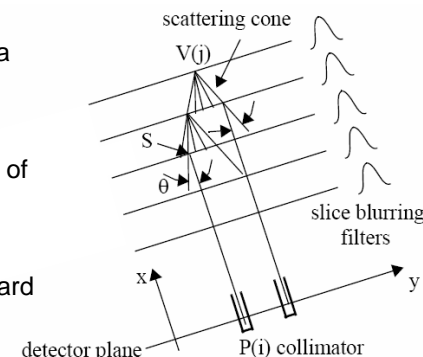


## Idea:

- scattering can be modeled by a phase function resembling a Gaussian
- the anatomical density map determines the parameters ( $\sigma$ ) of this Gaussian

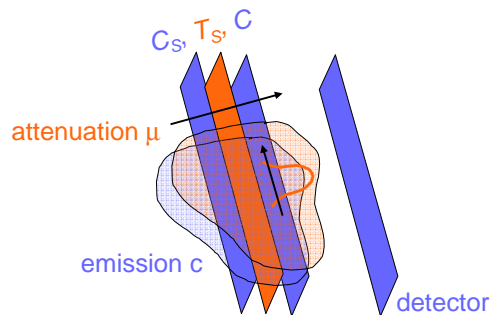
## Approach:

- scattering of emissions in forward projection is a back-to-front diffusion process (see figure)
- scattering of backprojected correction factors is a front-to-back diffusion process



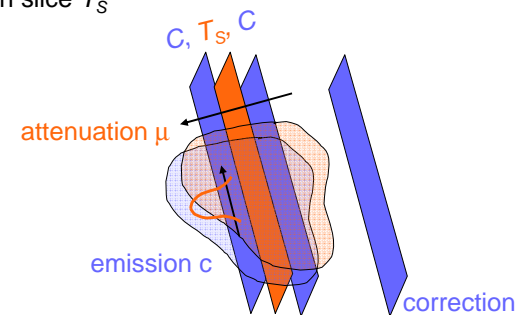
## Forward projection (back-to-front traversal):

- emission buffer  $C = 0$
- step from back to front, at each step:
  - interpolate emission slice  $C_S$  and attenuation slice  $T_S$
  - blur  $C$  using  $T_S$
  - $C = C + C_S$



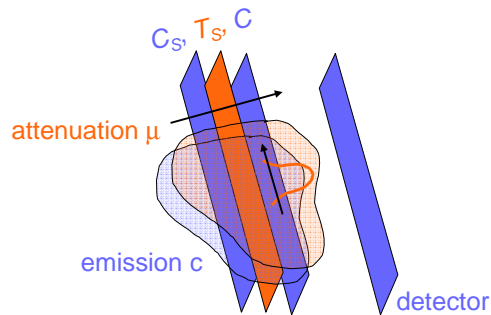
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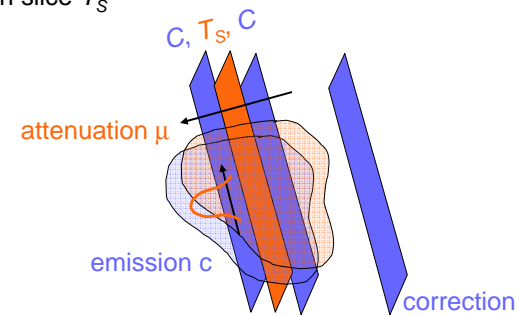
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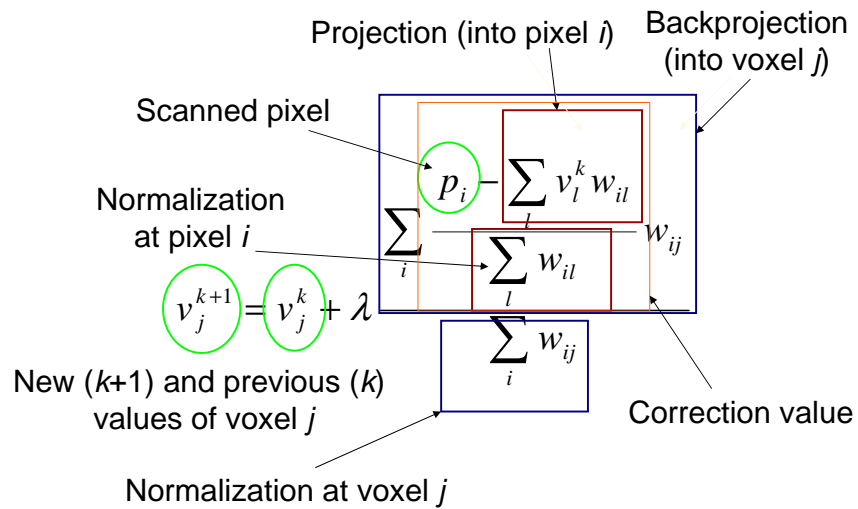
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  - $C = C \cdot T_S + C_S$



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  - update  $C = C \cdot T_S$





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