

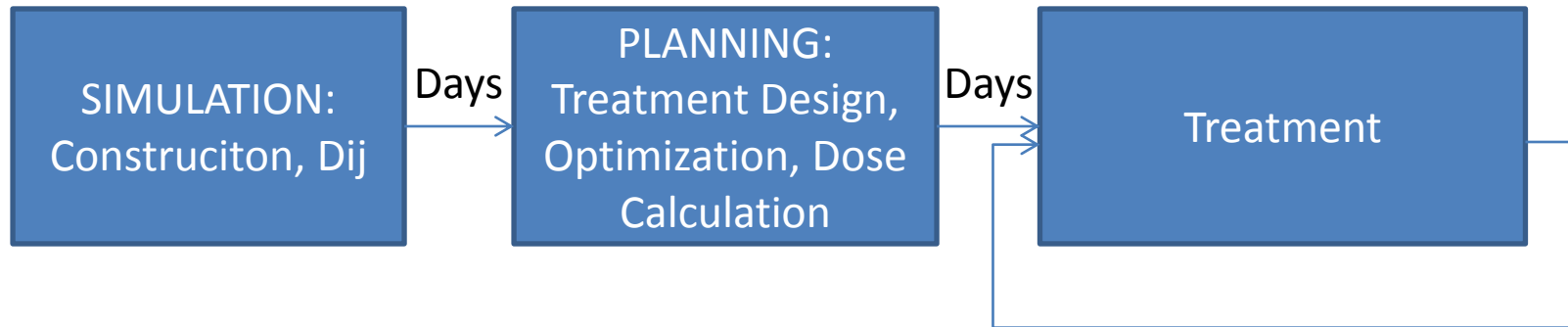
# GPU applications in Cancer Radiation Therapy at UCSD

Steve Jiang, UCSD Radiation Oncology

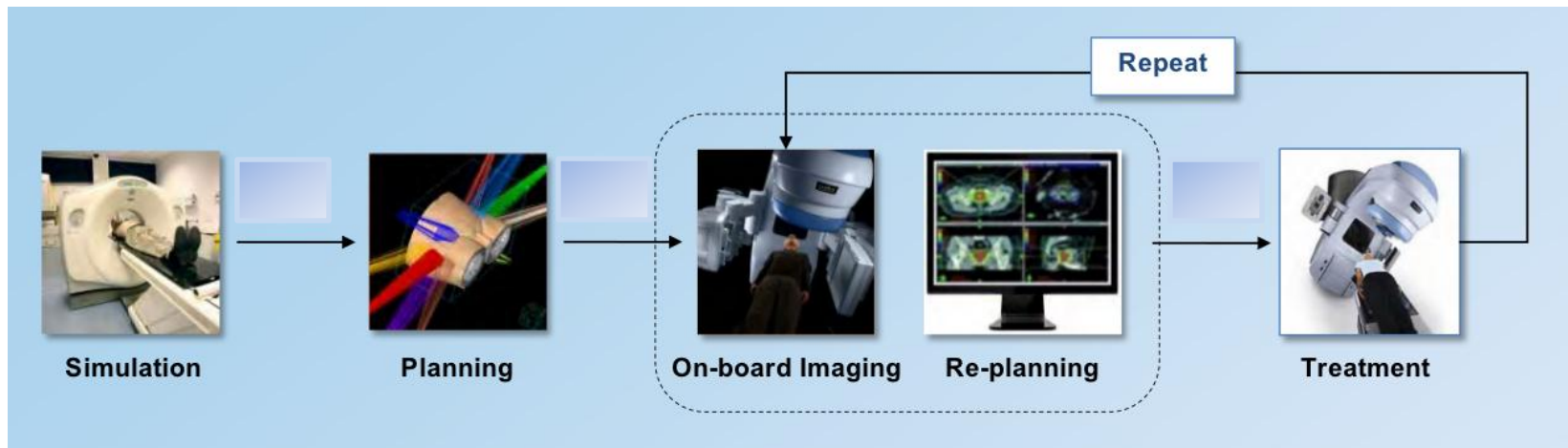
Amit Majumdar, SDSC

Dongju (DJ) Choi, SDSC

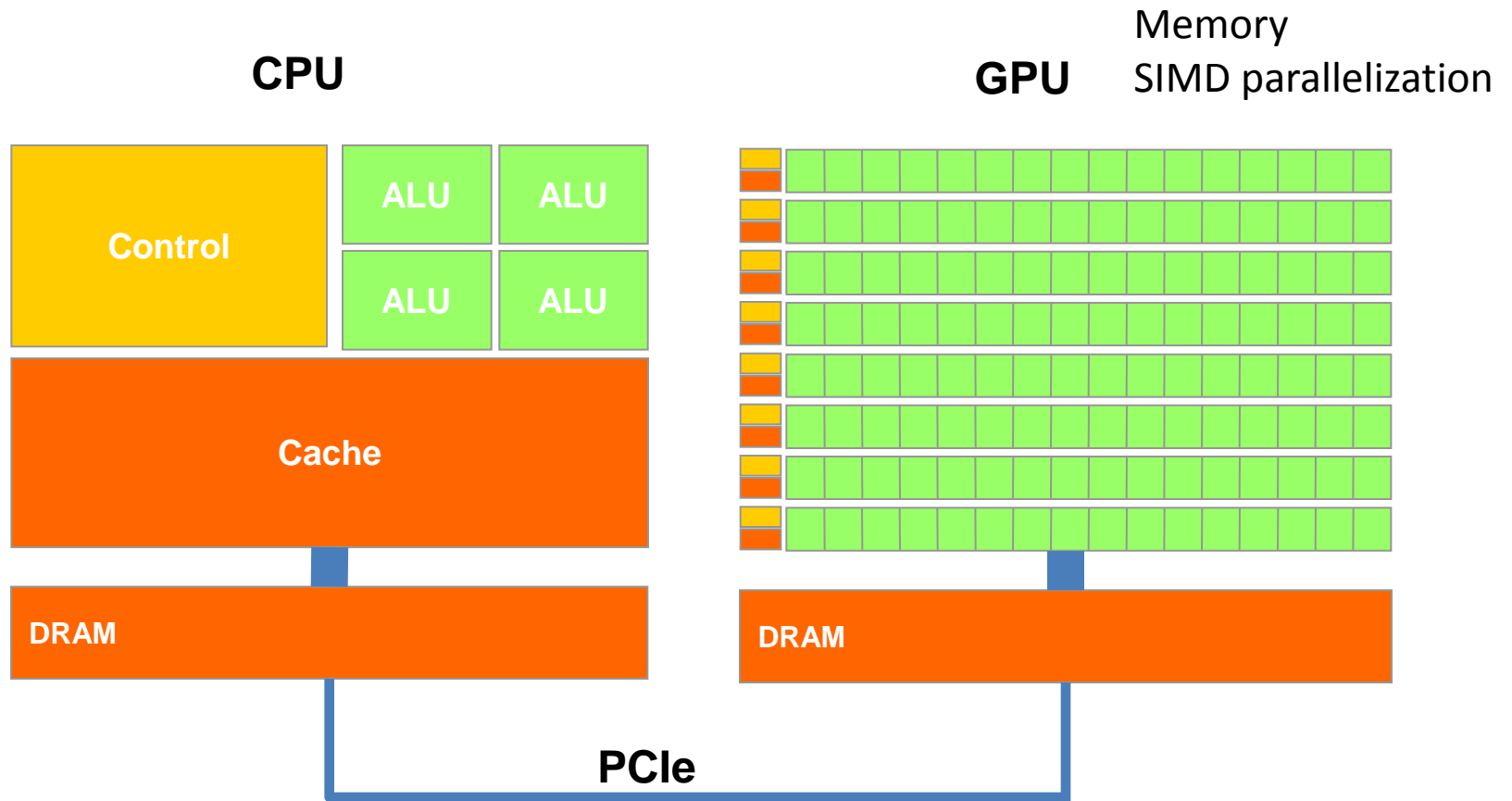
# Conventional Radiotherapy



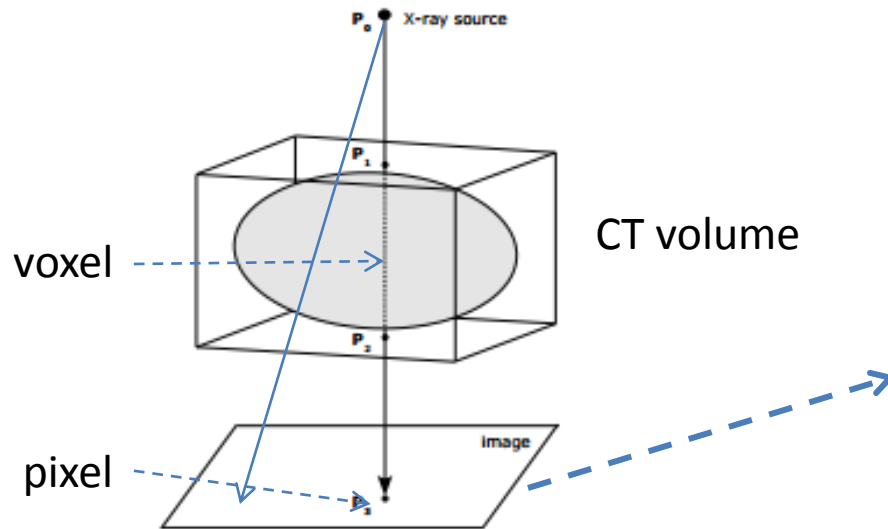
## Online ART (Adaptive Radiation) We Want



# CPU vs GPU



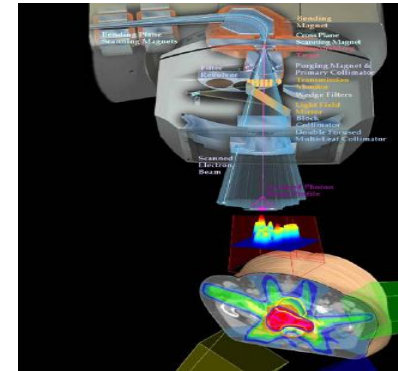
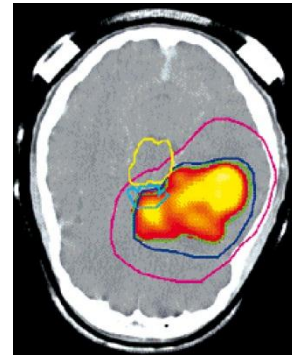
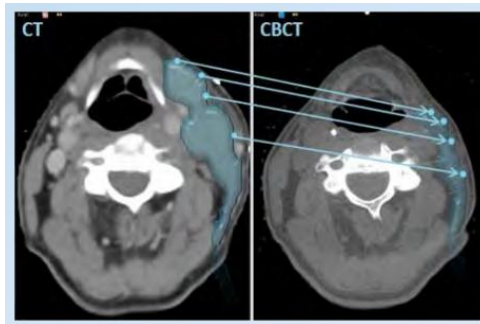
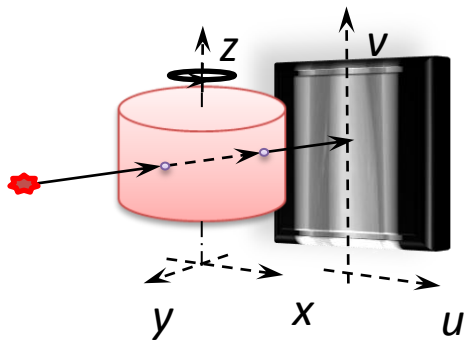
# Simple Radiation Application in GPU: DRR (Digitally Reconstructed Radiographs)



- Image pixel calculated in each thread
- CT volume in Texture memory
- Few seconds in CPU becomes few 1/10s seconds in GPU

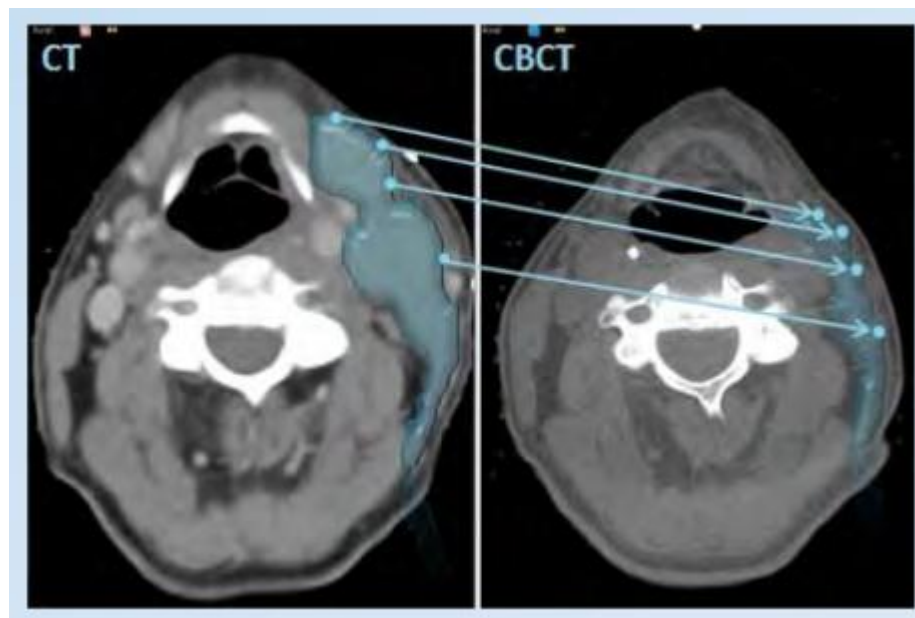
# Online Re-planning Process Needs..

- CBCT (Cone-Beam CT) Reconstruction
- Deformable Image Registration
- Re-contours
- Dose Calculation
- Plan Re-Optimization



# Deformable Image Registration with 'Demons'

- Morphing one image (volume) into another
- Displacement vector on each pixel in each thread
- Iterate until displacement vector becomes less than criteria
- For CT size 256X256X100, the speed up was ~50X (C1060 vs. Intel Xeon 2.27GHz)



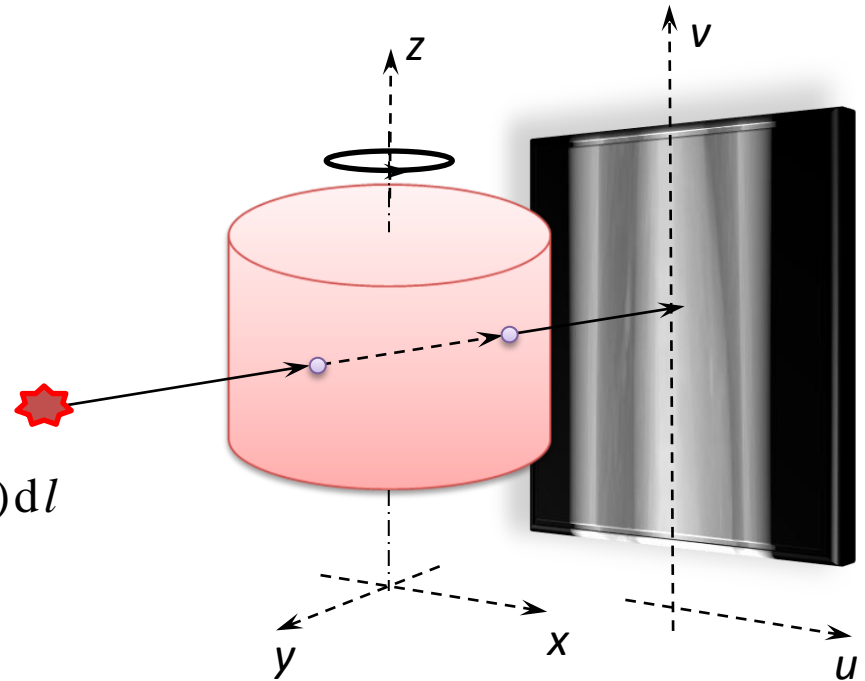
$$\vec{d}(x) = \frac{2[I_0(x) - I_1(x)][\vec{\nabla}I_0(x) + \vec{\nabla}I_1(x)]}{\|\vec{\nabla}I_0(x) + \vec{\nabla}I_1(x)\|^2 + [I_0(x) - I_1(x)]^2/K^2},$$

# Cone Beam CT

- CBCT reconstruction problem

$$I(u, v) = I_0 e^{-\int_S^{(u,v)} f(x, y, z) dl}$$

$$g(u, v) = -\log\left(\frac{I}{I_0}\right) = \int_S^{(u,v)} f(x, y, z) dl$$



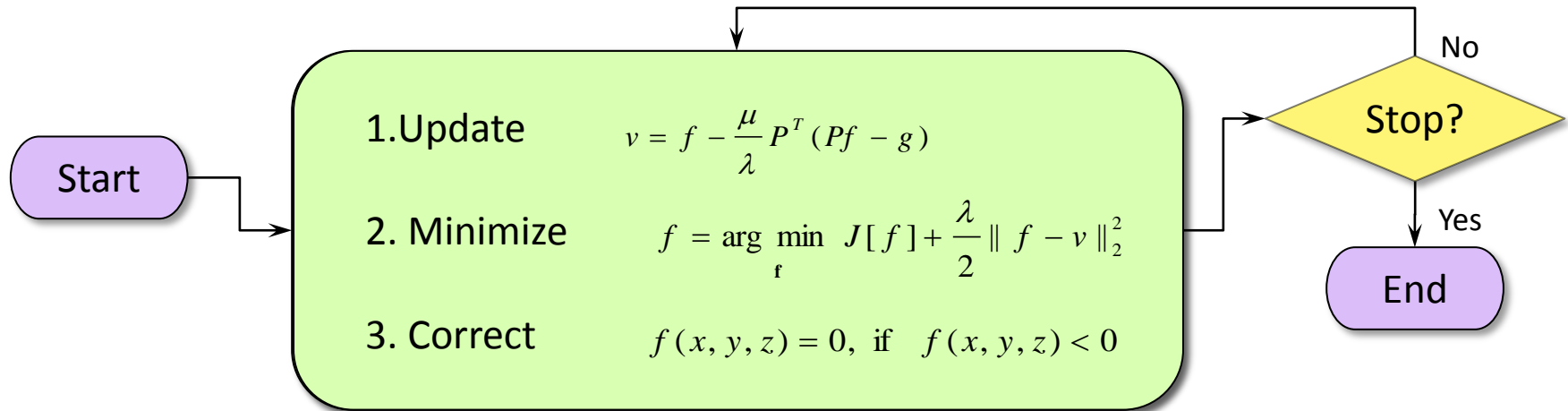
- Mathematical problem

Find a volumetric image  $f(x, y, z)$  given  $g(u, v)$  s.t.  $Pf \sim g$

- $P$  --- projection operator in cone beam geometry
- $f(x, y, z)$  --- volumetric image to be reconstructed
- $g(u, v)$  --- measured projections on imager

# Algorithm

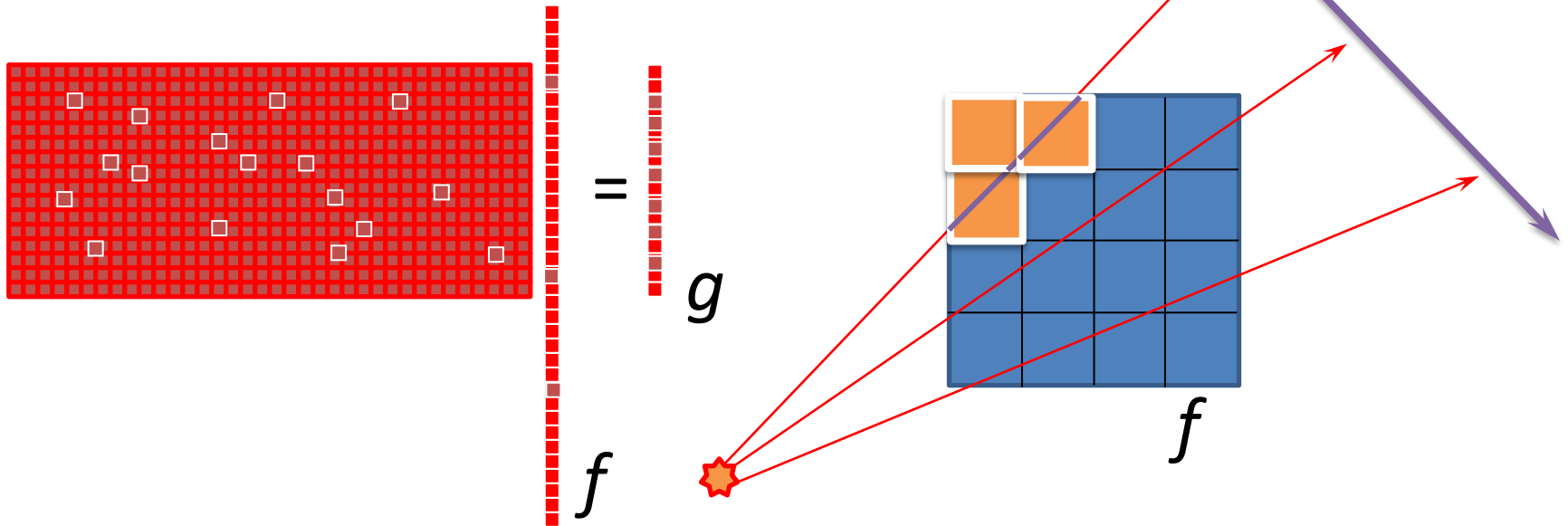
- Optimization procedure





# Computation of $g = Pf$

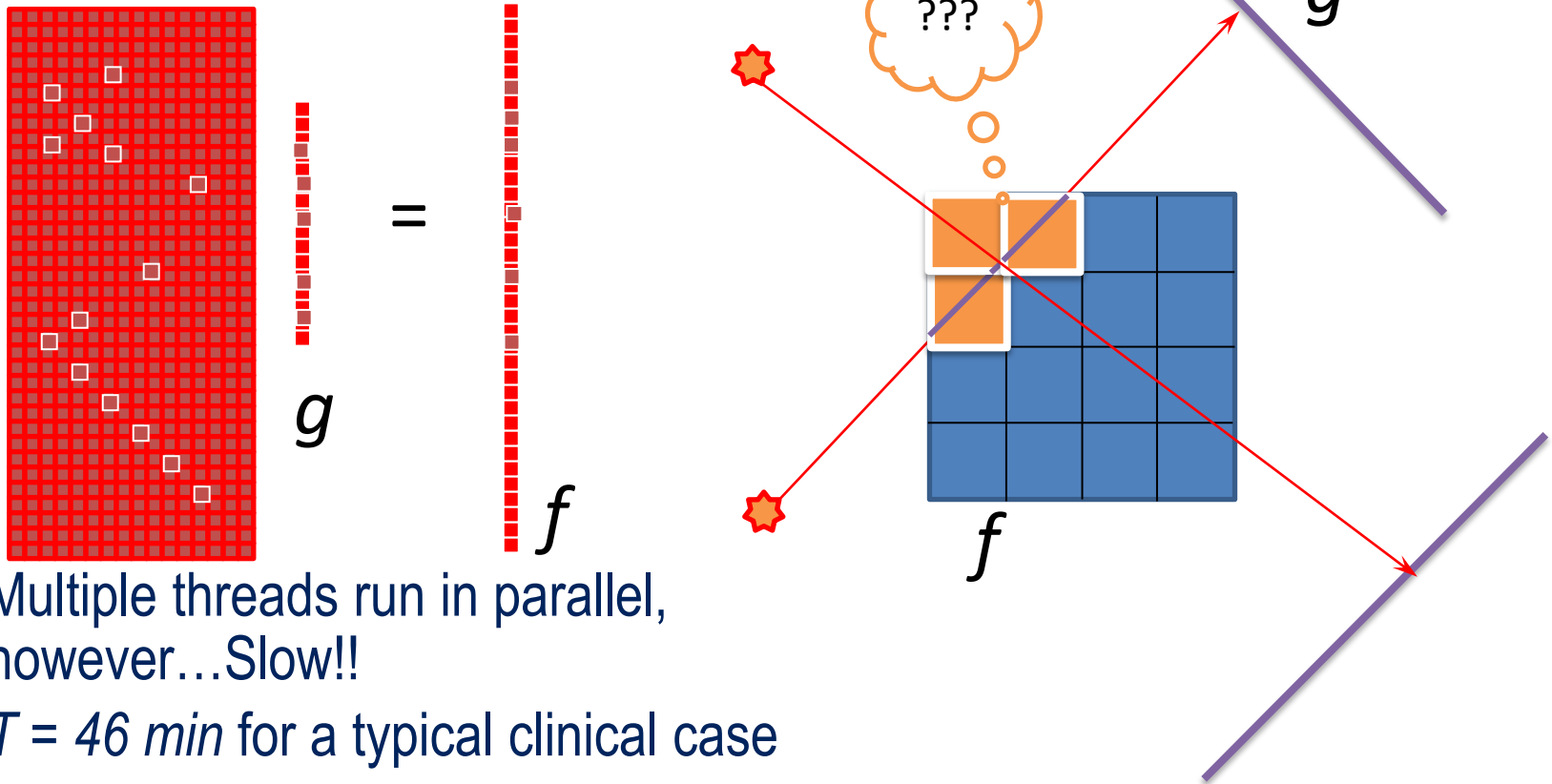
- Forward calculation
- Ray tracing algorithm on each GPU thread for one element in  $g$



- Multiple threads run in parallel ~100 times faster than CPU implementation

# Computation of $f = P^T g$

- Backward calculation
- Ray tracing algorithm on each GPU thread



- Multiple threads run in parallel, however...Slow!!
- $T = 46 \text{ min}$  for a typical clinical case

# GPU-friendly Algorithm for $P^T$

- Consider  $g = Pf$  in a continuum case

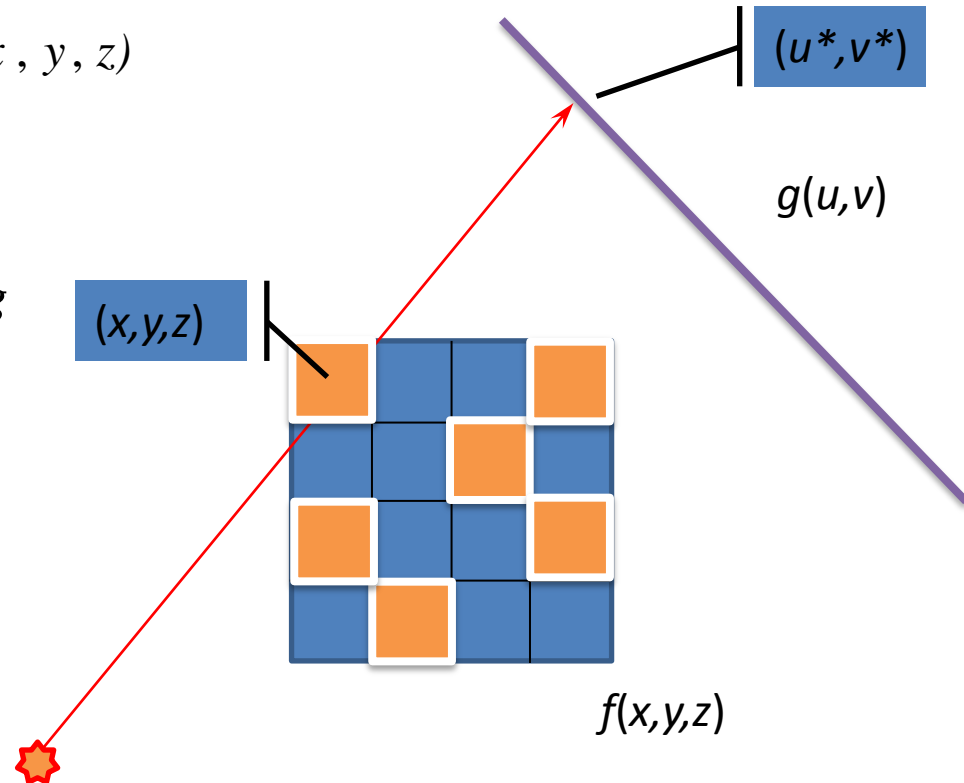
$$P[f(x, y, z)](u, v) = \int_L dl f(x, y, z)$$

- Mathematically,  $P^T$  defined as

$$\langle f, P^T g \rangle = \langle Pf, g \rangle \text{ for } \forall f, g$$

- Solve for  $P^T$

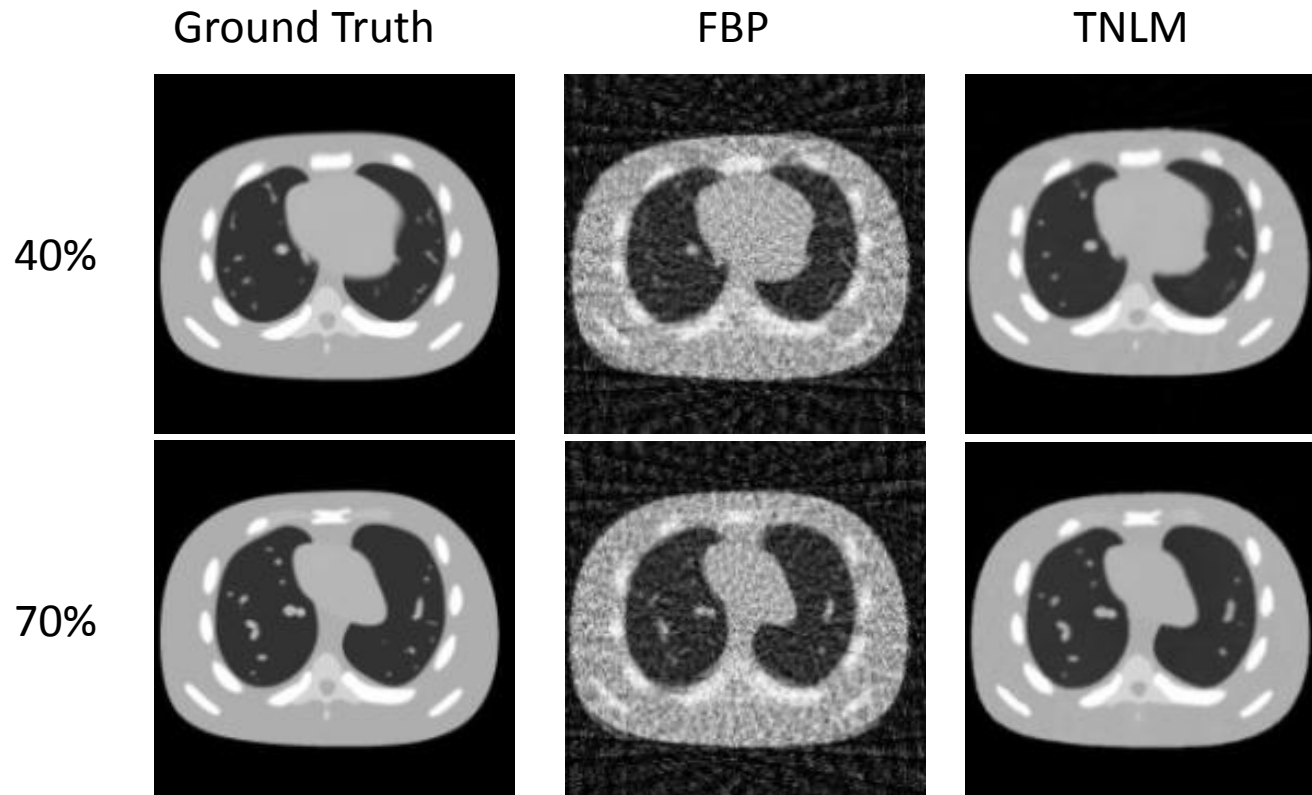
$$\begin{aligned} P^T[g(u, v)](x, y, z) \\ = \frac{L^3(u^*, v^*)}{L_0 l^2(x, y, z)} g(u^*, v^*) \end{aligned}$$



- Multiple threads run in parallel. Speed up by a factor of  $\sim 20$

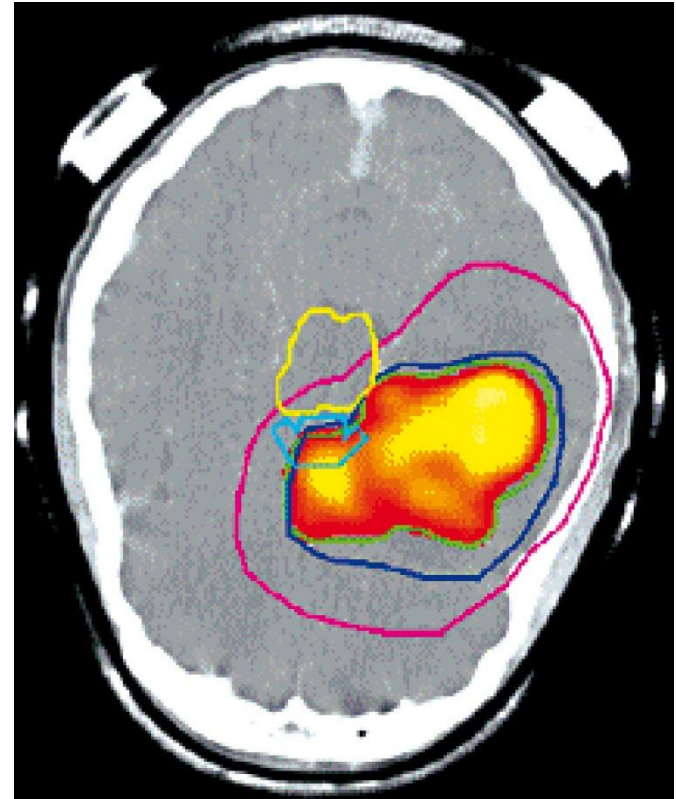
# Results

- Digital NCAT phantom in a realistic GE CT geometry
- Case #3: Low (500) projections acquired with low (20) mAs protocol



# Online Re-planning Process Needs..

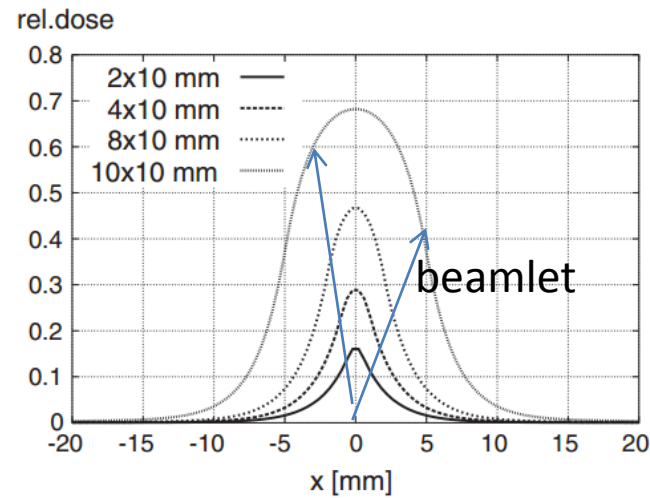
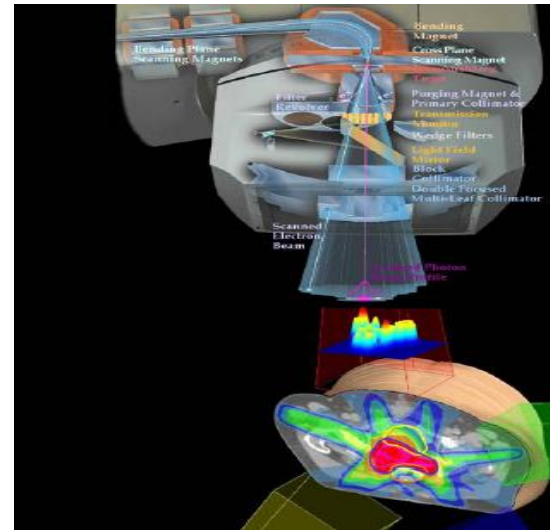
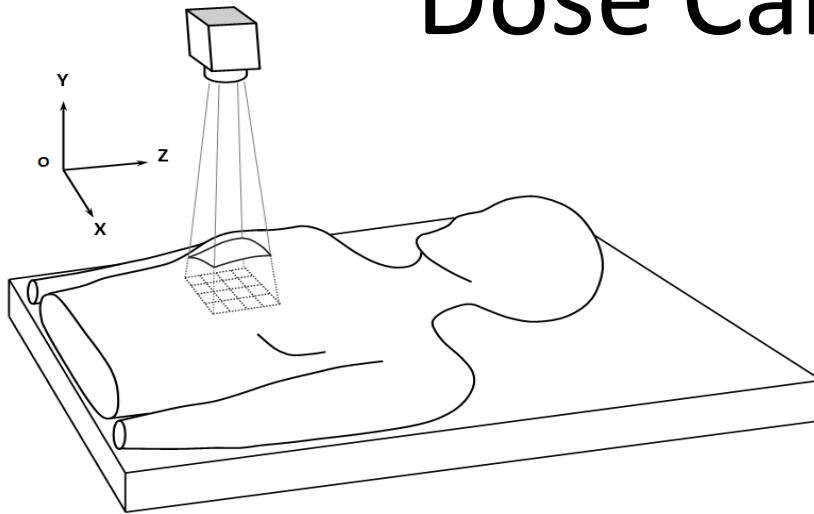
- ~~CBCT Reconstruction~~
- ~~Deformable Image Registration~~
- Re-contours
- Dose Calculation
- Plan Re-Optimization



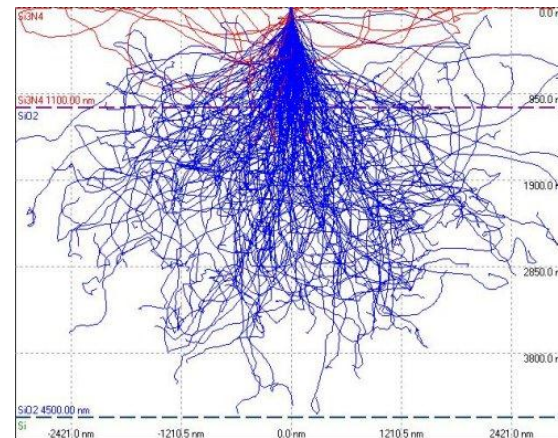
# GPU-based Treatment Planning @ UCSD

- Dose calculation
  - gFSPB: finite size pencil beam model
  - gDPM: DPM Monte Carlo code to GPU
- ~~Plan optimization~~
  - ~~gFMO: fluence map optimization~~
  - ~~gDAO: direct aperture optimization~~
  - ~~gVMAT: optimization for volumetric modulated arc therapy~~

# Dose Calculation

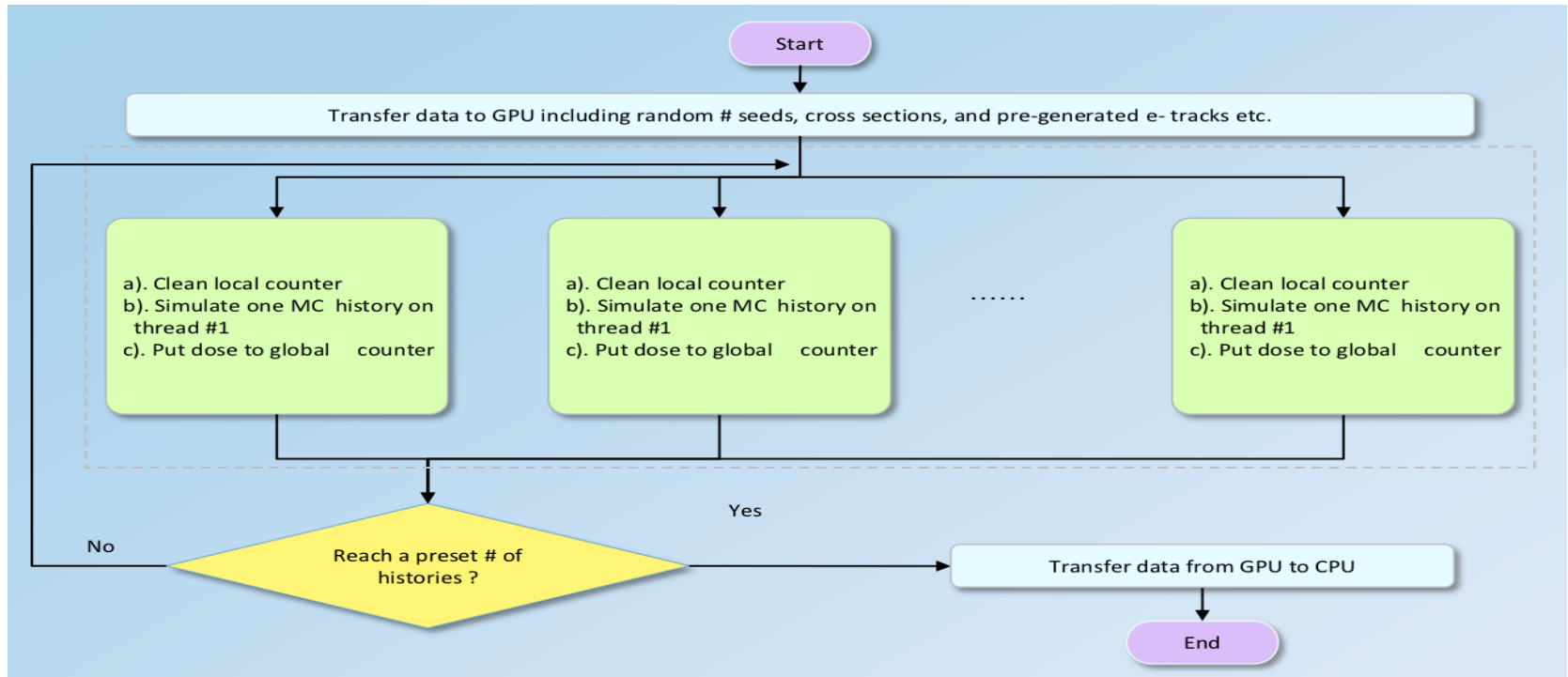


FSPB Beam Cross-profiles



Monte Carlo histories

# Monte Carlo Dose Calculation on GPU





# Results (photon point source, 6MV on head)

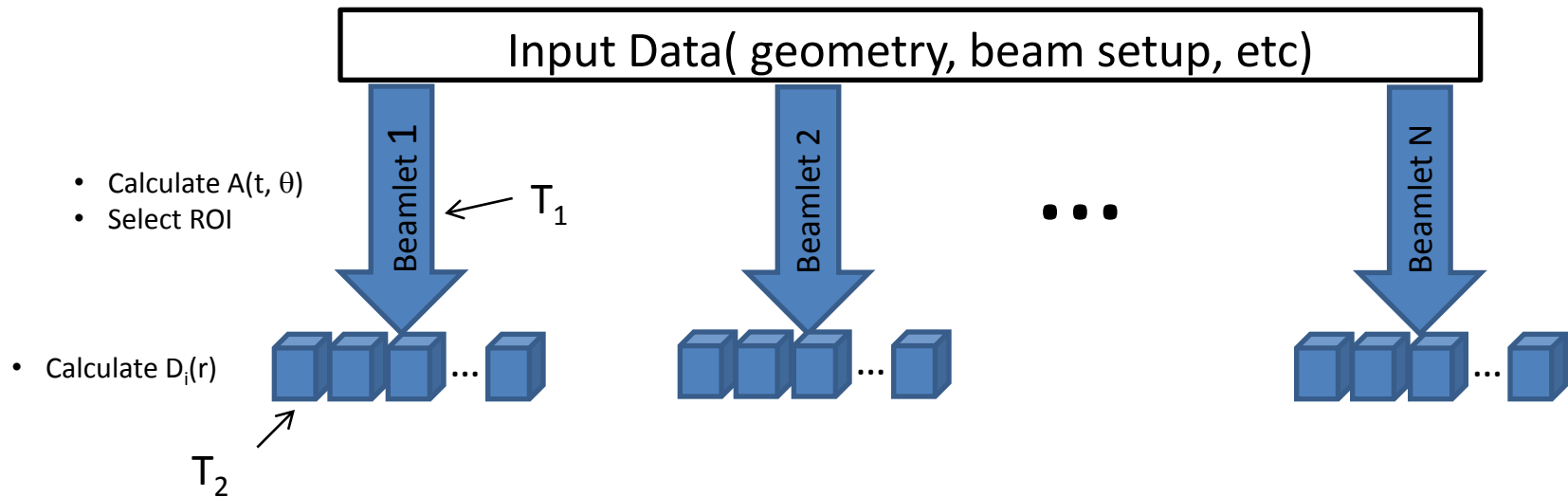
Execution time  $T$ , and speed-up factor  $T_{CPU}/T_{GPU}$  for four different testing cases.

| Source type    | # of Histories    | Case                  | $T_{CPU}$<br>(sec) | $T_{GPU}$<br>(sec) | $T_{CPU}/T_{GPU}$ |
|----------------|-------------------|-----------------------|--------------------|--------------------|-------------------|
| 20MeV Electron | $2.5 \times 10^6$ | water-lung-water      | 117.5              | 2.05               | 57.3              |
| 20MeV Electron | $2.5 \times 10^6$ | water-bone-water      | 127.0              | 1.97               | 64.5              |
| 6MV Photon     | $2.5 \times 10^8$ | water-lung-water      | 1403.7             | 18.6               | 75.5              |
| 6MV Photon     | $2.5 \times 10^8$ | water-bone-water      | 1741.0             | 24.2               | 71.9              |
| 6MV Photon     | $2.5 \times 10^8$ | VMAT HN patient       | N/A                | 36.7               | N/A               |
| 6MV Photon     | $2.5 \times 10^8$ | VMAT Prostate patient | N/A                | 39.6               | N/A               |
| 6MV Photon     | $2.5 \times 10^8$ | IMRT HN patient       | N/A                | 36.1               | N/A               |

CPU: Intel Xeon processor with 2.27GHz

GPU: NVIDIA Tesla C2050

# Finite-size Pencil Beam Model Data-Parallel Task



- Uses Thrust library for sorting
- Beams in loop (cpu), beamlets in loop (gpu), voxels in threads

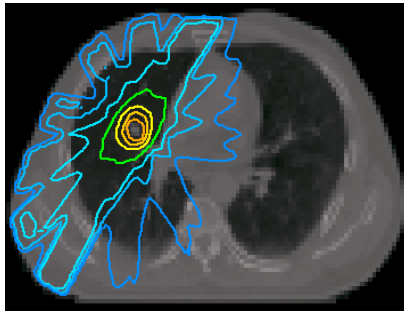
# Patient Cases

**Table 1. Tumor site, number of beams, and case dimension for 5 head-and-neck (H1-H5) cases and 5 lung (L1-L5) cases.**

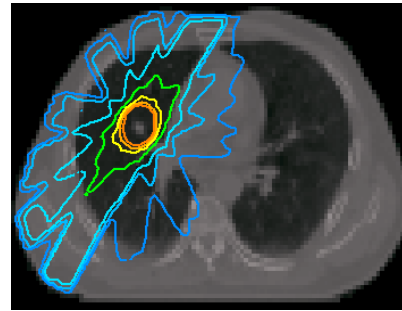
| Case | Tumor Site                              | # of Beams       | # of Beamlets | # of Voxels |
|------|---|------------------|---------------|-------------|
| H1   | Parotid                                 | 8 (non-coplanar) | 7,264         | 128×128x72  |
| H2   | Hypopharynx                             | 7 (non-coplanar) | 4,429         | 128x128x72  |
| H3   | Nasal Cavity                            | 8 (non-coplanar) | 3,381         | 128x128x72  |
| H4   | Parotid                                 | 5 (coplanar)     | 4,179         | 128x128x72  |
| H5   | Larynx                                  | 7 (non-coplanar) | 10,369        | 128x128x72  |
| L1   | Left lung, low lobe(close to pleura)    | 6 (coplanar)     | 637           | 128x128x80  |
| L2   | Right lung, low lobe (paravertebral)    | 6 (coplanar)     | 1,720         | 128x128x103 |
| L3   | Left lung, upper lobe (close to pleura) | 5 (coplanar)     | 921           | 128x128x80  |
| L4   | Right lung, upper lobe (close to heart) | 7 (coplanar)     | 841           | 128x128x80  |
| L5   | Left lung (middle)                      | 5 (coplanar)     | 686           | 128x128x80  |

# Lung Cancer Patient Case (L4)

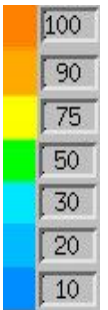
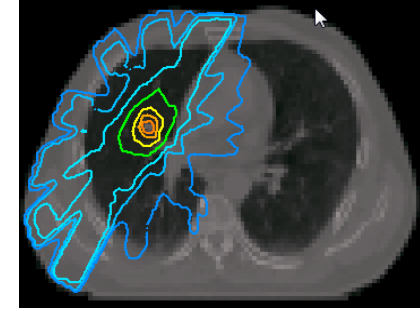
Monte Carlo



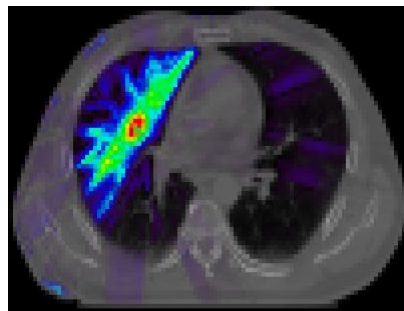
g-FSPB



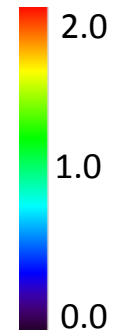
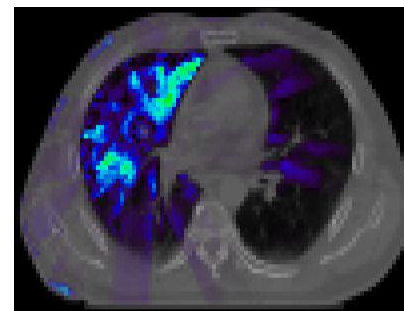
DC(Density Correction)-FSPB



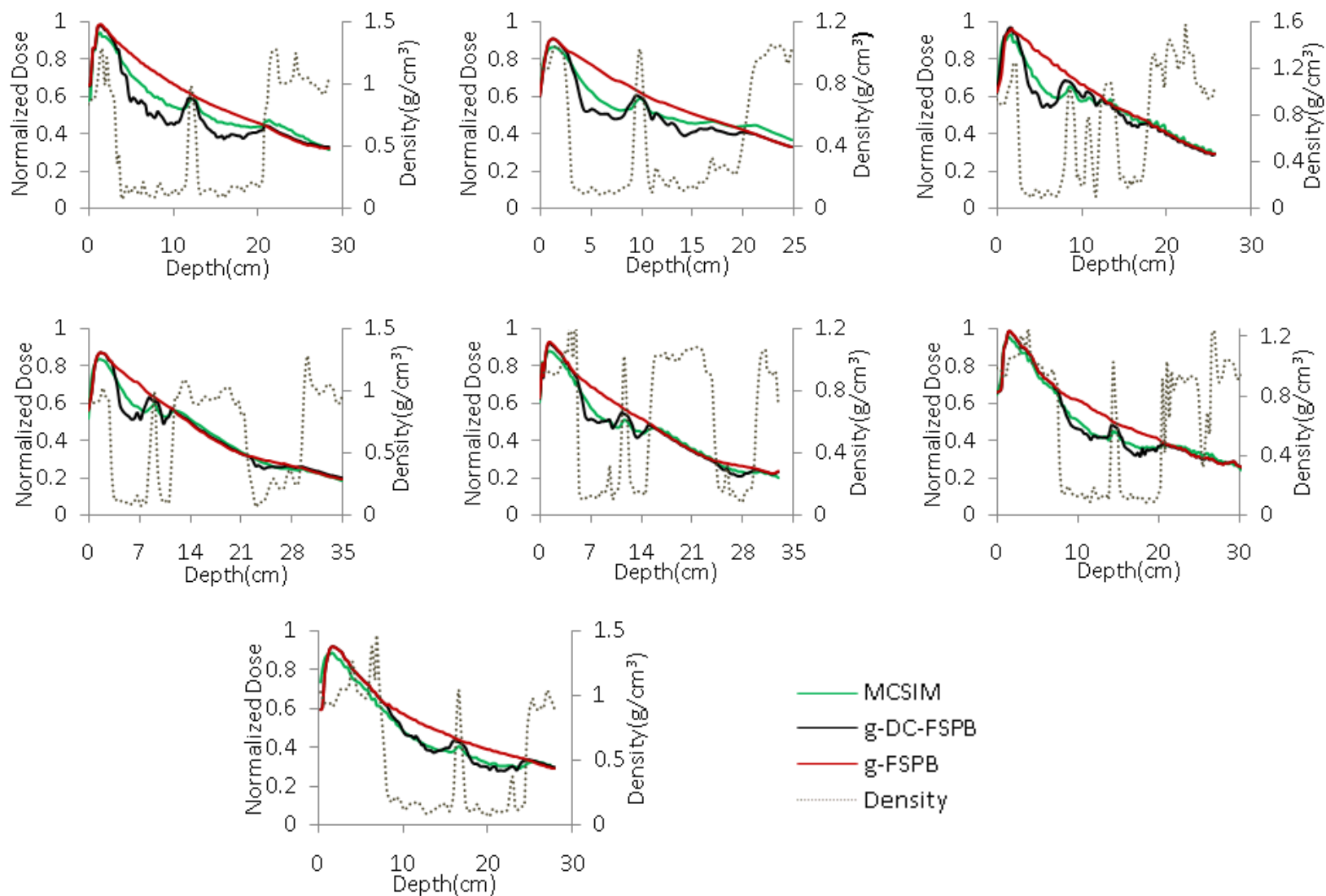
$\gamma$ : g-FSPB



$\gamma$ : g-DC-FSPB



# Lung Cancer Patient Case (L4)



# Accuracy and Efficiency

Table 2. Gamma index evaluation results and dose calculation computation time for 10 testing cases using the g-DC-FSPB algorithm. The corresponding g-FSPB results are given in parenthesis for comparison purpose.

| Case # | $\gamma^{max}$ | $\gamma_{50}^{avg}$ | $P_{50}$        | $T_{tr}(\text{sec})$ | $T_{gpu}(\text{sec})$ | $T_{tot}(\text{sec})$ |
|--------|----------------|---------------------|-----------------|----------------------|-----------------------|-----------------------|
| H1     | 2.12 (2.16)    | 0.30 (0.31)         | 97.53% (97.32%) | 0.20                 | 0.64 (0.55)           | 0.84 (0.75)           |
| H2     | 3.44 (4.11)    | 0.28 (0.28)         | 97.80% (97.01%) | 0.20                 | 0.40 (0.35)           | 0.60 (0.55)           |
| H3     | 2.27 (2.36)    | 0.46 (0.52)         | 92.29% (86.39%) | 0.20                 | 0.38 (0.34)           | 0.58 (0.54)           |
| H4     | 3.08 (3.11)    | 0.61 (0.63)         | 82.96% (81.56%) | 0.19                 | 0.35 (0.32)           | 0.54 (0.51)           |
| H5     | 3.33 (3.37)    | 0.61 (0.61)         | 86.19% (86.09%) | 0.20                 | 1.31 (1.10)           | 1.51 (1.30)           |
| L1     | 1.53 (1.92)    | 0.24 (0.45)         | 99.35% (94.81%) | 0.21                 | 0.22 (0.20)           | 0.43 (0.41)           |
| L2     | 2.35 (3.30)    | 0.36 (0.71)         | 96.64% (76.38%) | 0.22                 | 0.40 (0.36)           | 0.62 (0.58)           |
| L3     | 1.68 (3.07)    | 0.32 (0.75)         | 99.16% (76.60%) | 0.21                 | 0.30 (0.25)           | 0.51 (0.46)           |
| L4     | 2.70 (4.59)    | 0.63 (1.53)         | 81.33% (28.55%) | 0.18                 | 0.25 (0.23)           | 0.43 (0.41)           |
| L5     | 2.19 (4.34)    | 0.49 (1.13)         | 90.24% (57.03%) | 0.21                 | 0.33 (0.29)           | 0.54 (0.50)           |

# GPU-based Treatment Planning @ UCSD

- ~~Dose calculation~~
  - ~~gFSPB: finite size pencil beam model~~
  - ~~gDPM: DPM MC code to GPU~~
- Plan optimization
  - gFMO: fluence map optimization
  - gDAO: direct aperture optimization
  - gVMAT: optimization for volumetric modulated arc therapy

# DAO Model

- Optimize w.r.t both aperture and intensity

# VMAT Model

- Optimize w.r.t both aperture and intensity
- One aperture is at one beam angle
- Aperture shapes at neighboring angles satisfy MLC mechanical constraints
- Smoothness of intensity changes between neighboring angles included in the objective function.



# FMO model and flow chart

$$\min \sum_{j \in V} F_j(z_j^l)$$

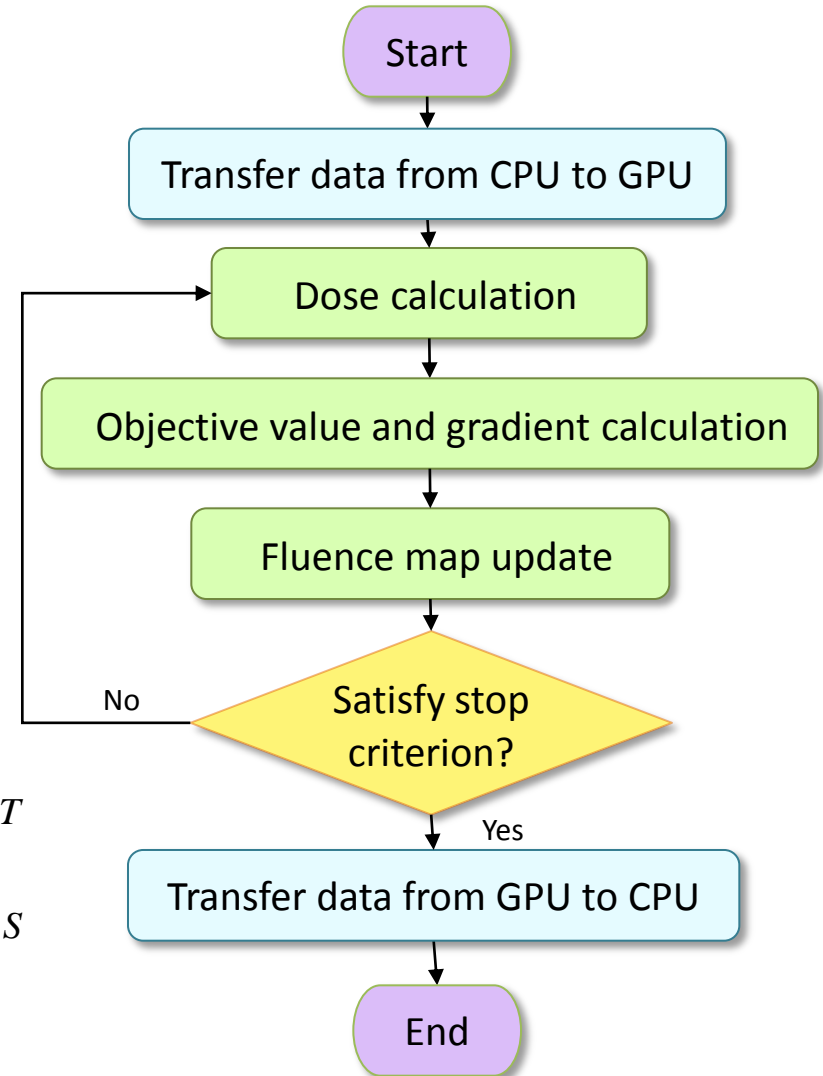
Subject to

$$\begin{aligned} \text{Voxel dose } z_j^l &= \sum_{i \in N} D_{ij}^l x_i & j \in V \\ x_i &\geq 0 & i \in N \end{aligned}$$

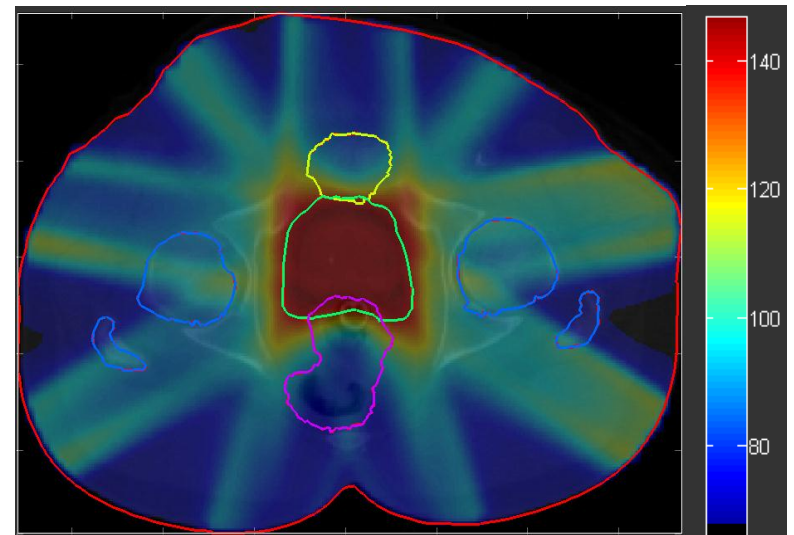
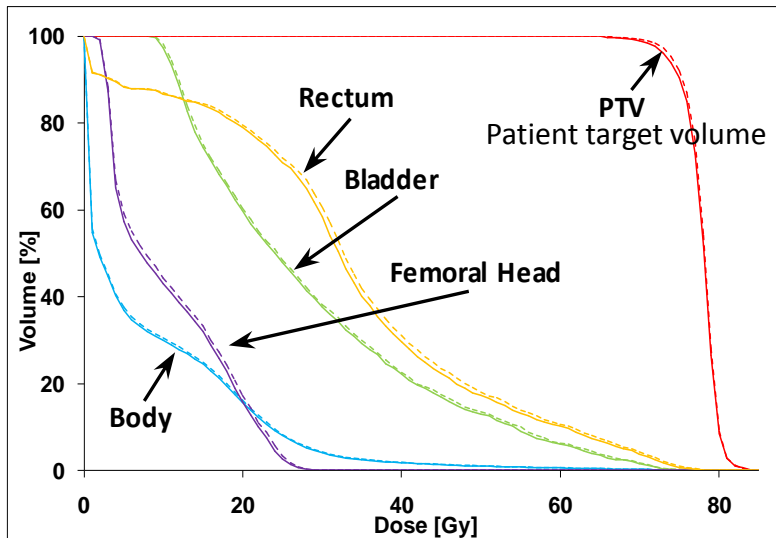
Where

$$F_{s-}(\mathbf{z}) = \sum_{j \in V_s} (\max(0, z_j - z_j^l))^2 \quad s \in T$$

$$F_{s+}(\mathbf{z}) = \sum_{j \in V_s} (\max(0, z_j^l - z_j))^2 \quad s \in S$$



# Results for GPU-based FMO Algorithm



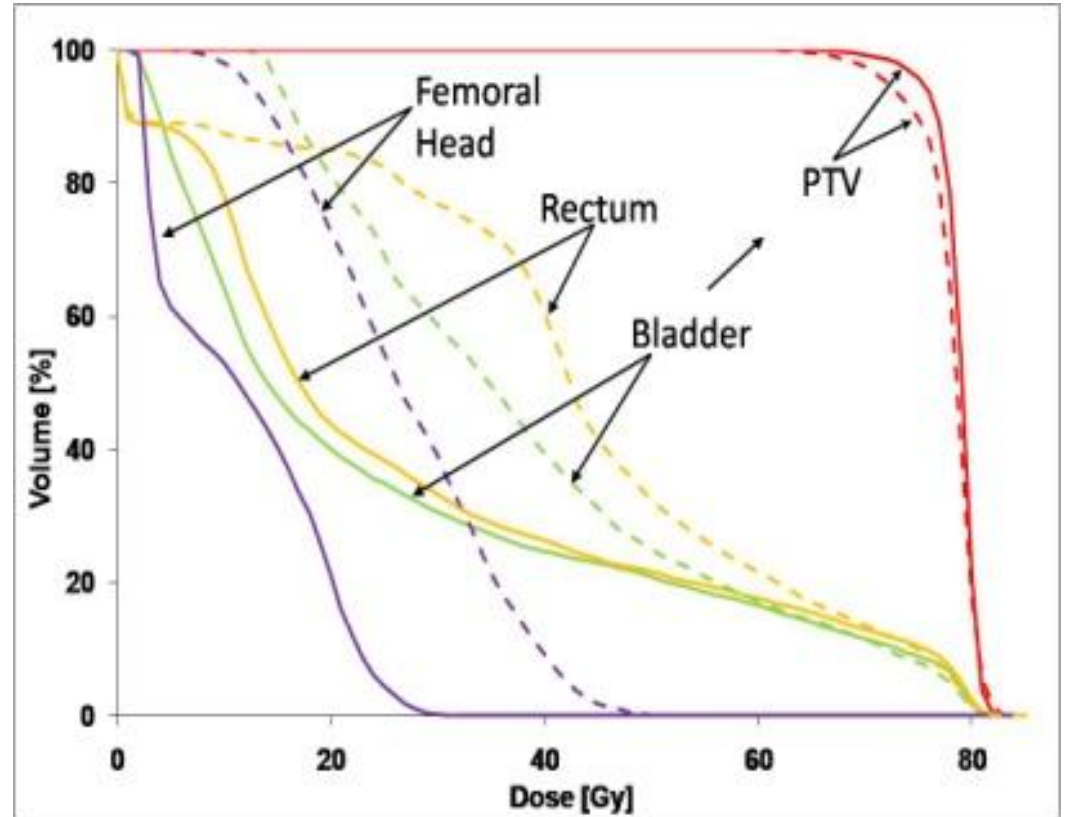
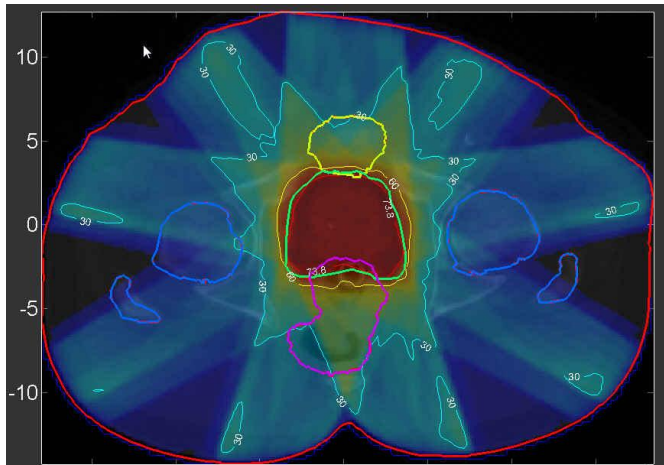
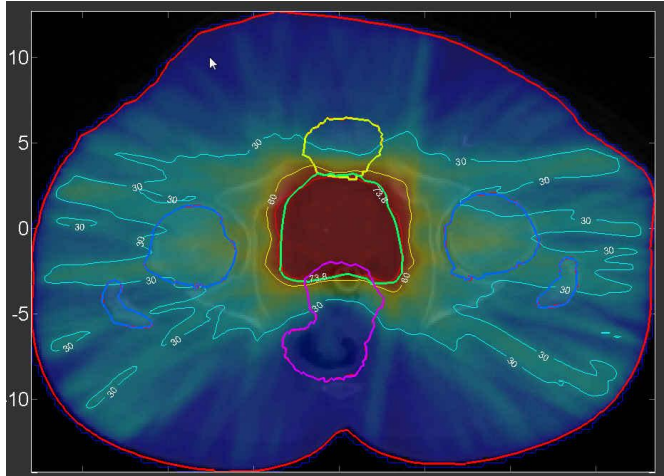
| # Case | beamlet size (mm <sup>2</sup> ) | # beamlets | voxel size (mm <sup>3</sup> ) | # voxels (× 10 <sup>4</sup> ) | # non-zero $D_{ij}$ 's (× 10 <sup>6</sup> ) | GPU time (s) |
|--------|---------------------------------|------------|-------------------------------|-------------------------------|---|--------------|
| 1      | 10 × 10                         | 2,055      | 4 × 4 × 4                     | 3.6                           | 3.1   | 0.2          |
| 2      | 5 × 5                           | 6,433      | 4 × 4 × 4                     | 3.6                           | 10.6  | 0.5          |
| 3      | 5 × 5                           | 6,433      | 2.5 × 2.5 × 2.5               | 14.0                          | 43.3  | 2.8          |

~40x speedup compared to an Intel Xeon 2.27 GHz CPU

~0.5 sec for re-optimizing a 9-field prostate IMRT plan

Men et al *Phys Med Biol* 54(21):6565-6573, 2009

# Prostate Cancer Case



— VMAT  
- - - IMRT

# Summary – GPU-based Treatment Planning

- We have developed GPU-based computational tools for real-time treatment planning
  - Conventional problem size have been well fit to one GPU hardware
- For a typical prostate case
  - The dose calculation takes less than 1 second with FSPB with 3D density correction, less than 40 seconds with Monte Carlo
  - The plan optimization takes less than 1 second with FMO, 2 seconds with DAO (intensity, aperture) , and 30 seconds with VMAT

## ➤ Next step

- ☐ Faster and Finer → algorithm improvement, larger problems, multiple GPUs
- ☐ Software integration → Analysis/Data integration
- ☐ Clinical implementation and evaluation

