



# Practical Implementation of pCT Algorithms

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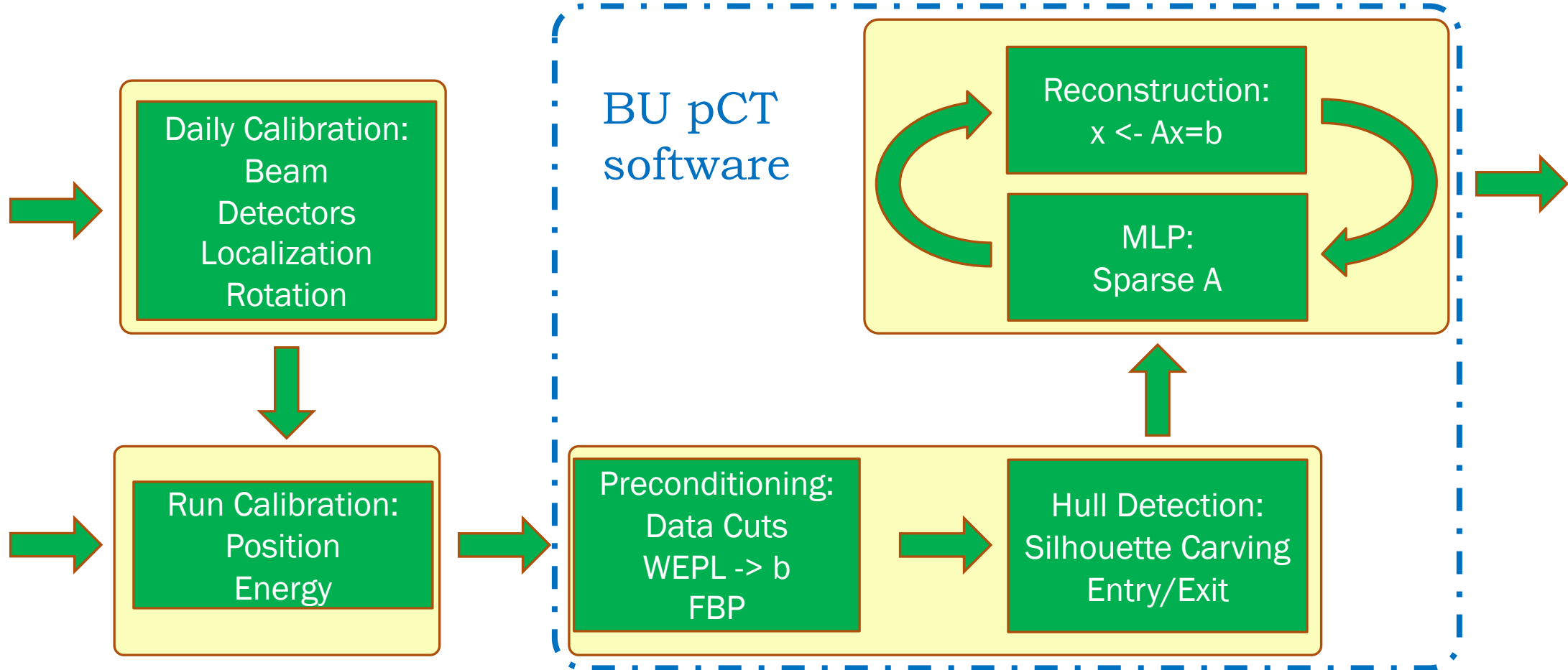
BLAKE E. SCHULTZE

BAYLOR UNIVERSITY

# Presentation Topics

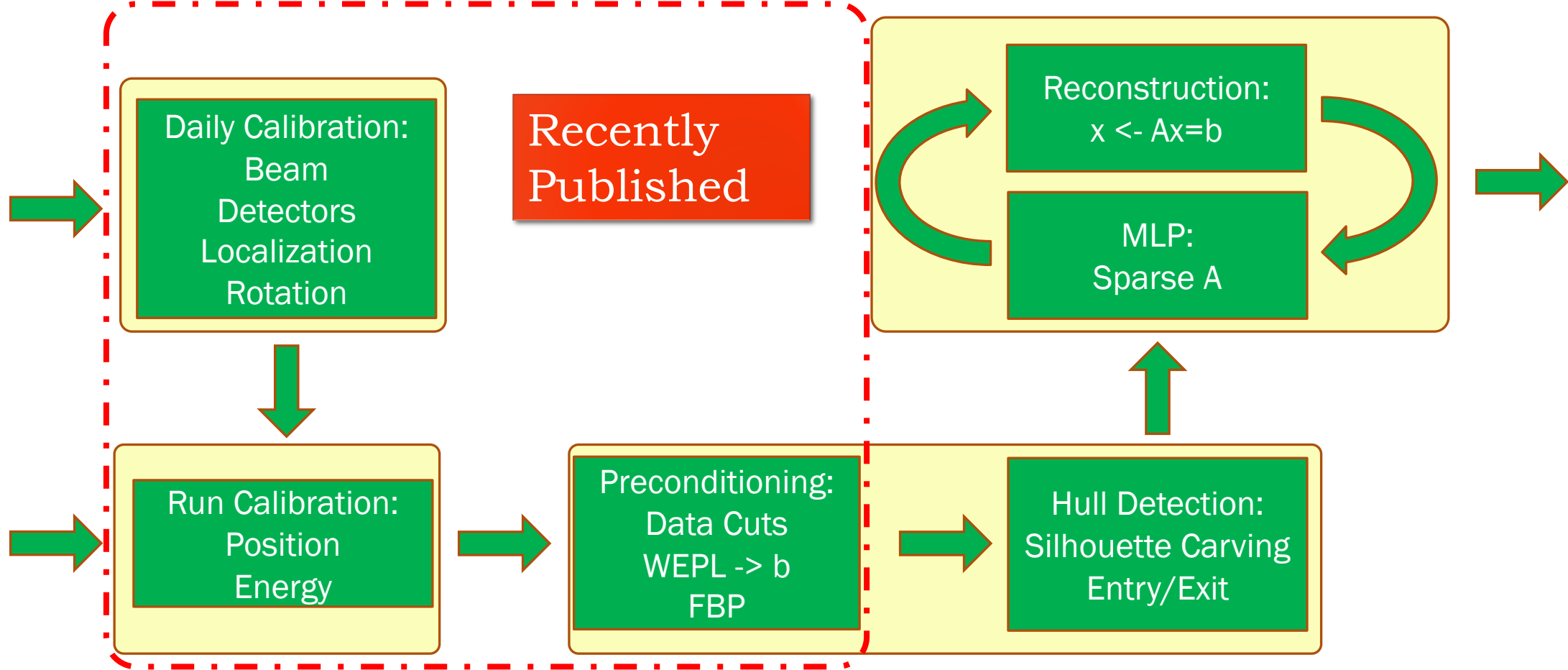
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- 1) Hull-detection: Development for object boundary detection
- 2) MLP Formalism: Development of a computationally efficient implementation
- 3) TVS: Applying recent methodological advances in theory to pCT



## pCT Data Flow

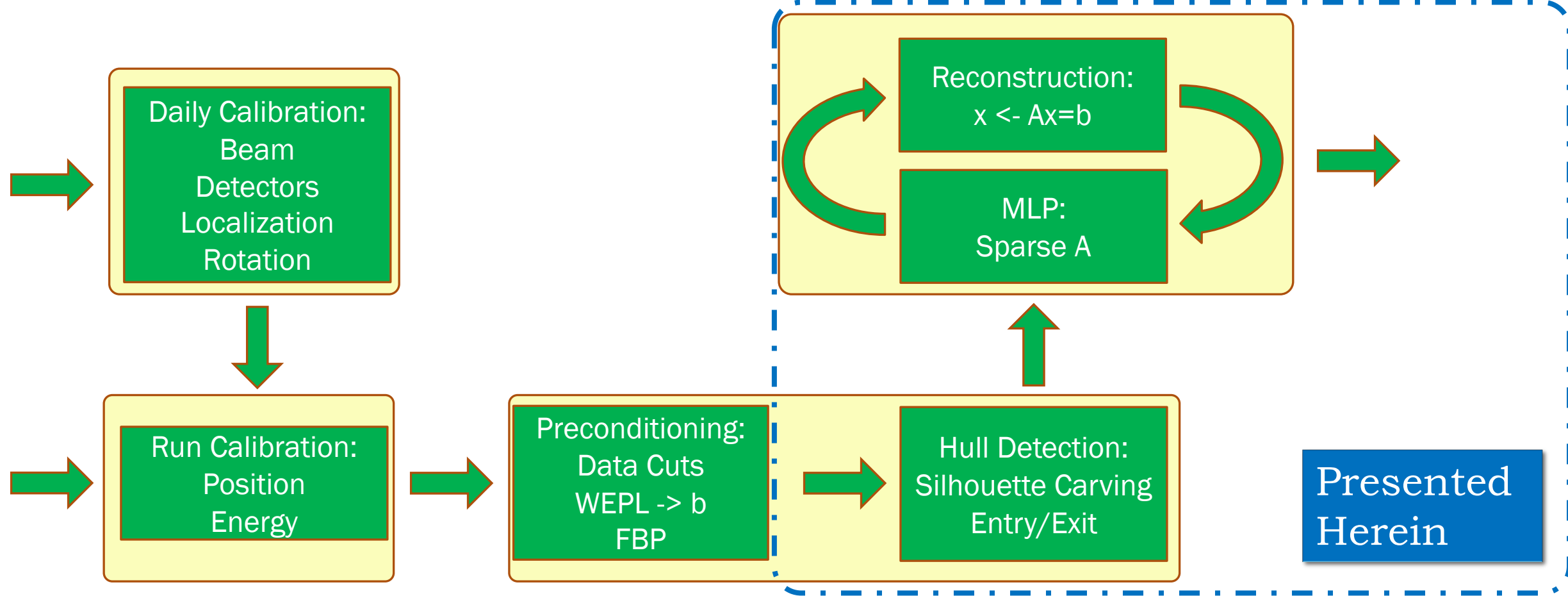
- BU pCT software performs preconditioning and image reconstruction steps



# Data Acquisition, Preprocessing, & Preconditioning

Detailed description of these steps published in *IEEE Access* (2021)

“Particle-tracking proton computed tomography—data acquisition, preprocessing, and preconditioning”



# Data Acquisition & Preprocessing

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# Hull- Detection

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## Object Boundary Detection

Object boundary detection is...

- used to determine if/where a proton entered/exited the object
  - used to restrict reconstruction to protons traversing object
  - used to restrict the reconstruction image space solely to those voxels within the object
- Published: *Contemporary Mathematics*, Vol. 36 (2015)  
“Performance of hull-detection algorithms for proton computed tomography reconstruction”

# Object Boundary Detection: Advantages of Hull-Detection?

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- can begin during pCT scans, i.e. online mode
- parallelizable & relatively fast (~5 seconds)
- hull can define mask for spatially filtering  $x_0 \Rightarrow$  restricts image space to relevant voxels
- more reliable object entry & exit coordinates
  - key to proton path estimation
- can identify & exclude protons missing object

# Silhouette Carving (SC) Method

- Fundamental  
Concepts

- Protons missing object...
  - experience negligible scattering & energy loss in air
  - travel along approximate straight line paths (SLP)
  - identifiable by WEPL and/or angular deviation
- Voxels along SLPs can be identified & excluded from hull
- Voxels not excluded after repeating for many protons defines object hull



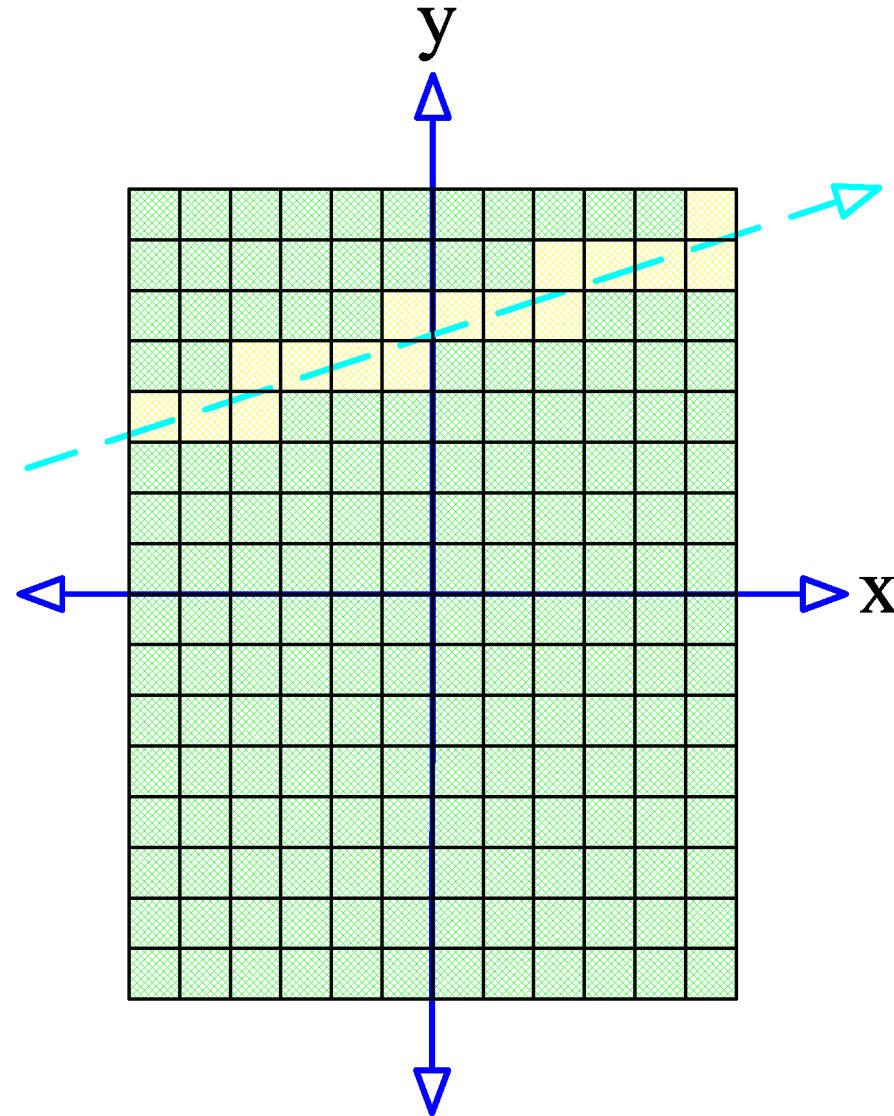
# Silhouette Carving (SC) Method

- Implementation  
Details

- Hull initialized to entire image space
- Protons missing object identified by  
 $WEPL < -1 \text{ mm}$
- Voxel walk algorithm (3D-DDA) developed to identify voxels along each proton path
- Based on 2D ray marching method: digital difference analyzer (DDA)
- 3D-DDA is efficient/accurate & numerically stable

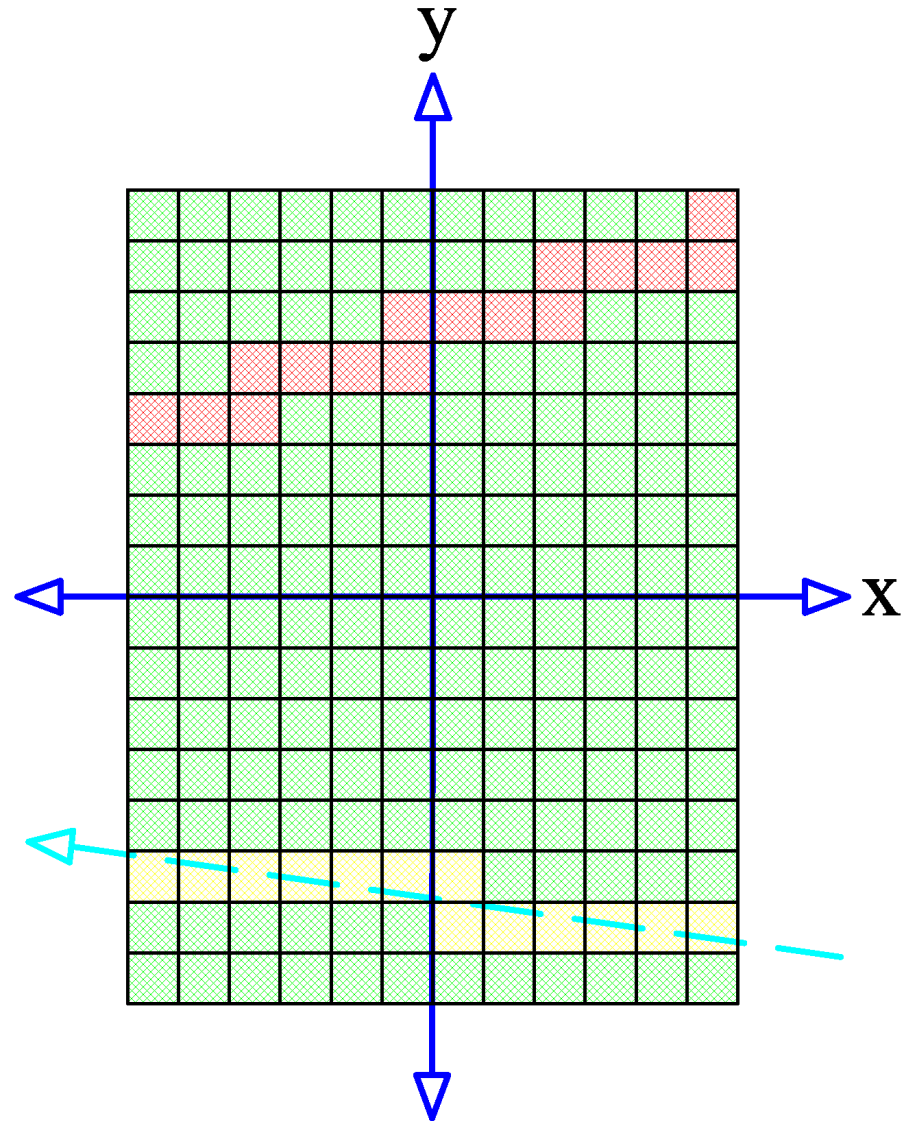
# Voxel Carving Process

- 1) Starting from hull encompassing image space
- 2) Identify voxels intersected by 1<sup>st</sup> proton missing object



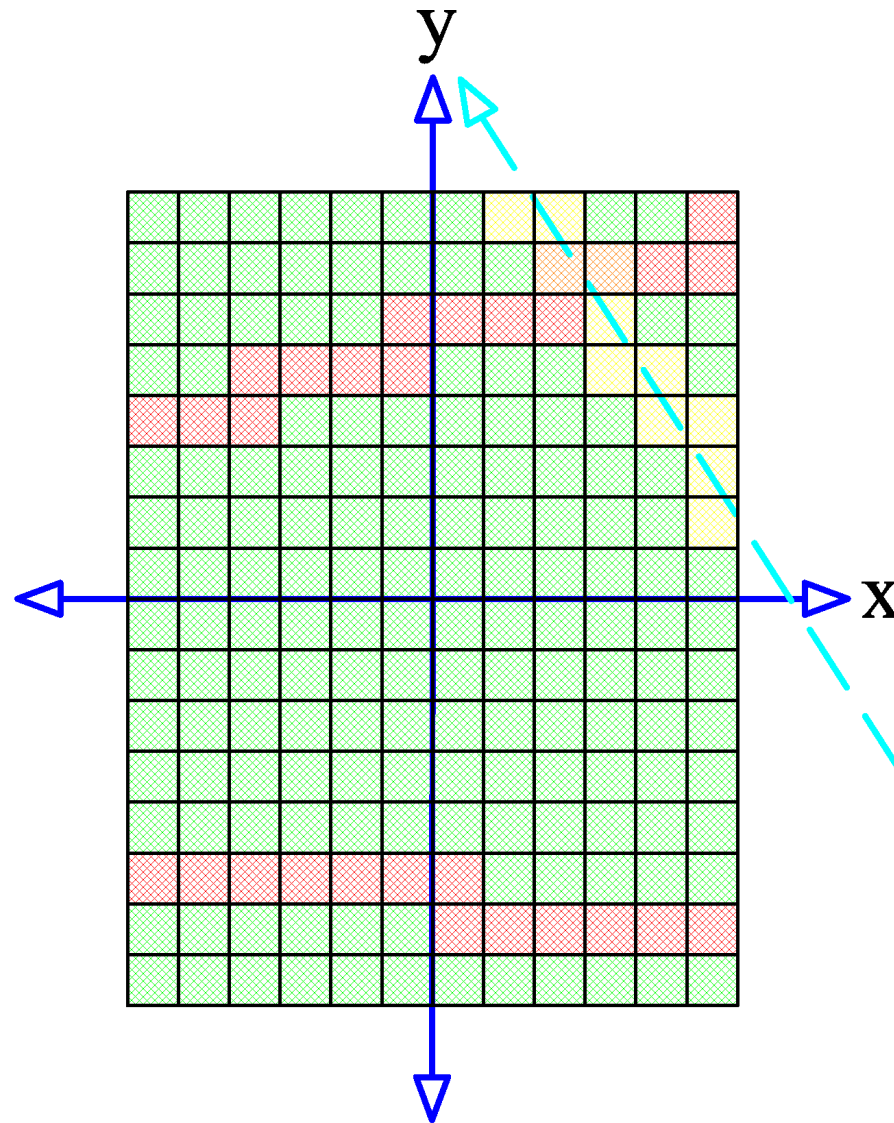
# Voxel Carving Process

- 3) Carve (remove) voxels intersected by 1<sup>st</sup> proton
- 4) Identify voxels intersected by 2<sup>nd</sup> proton missing object



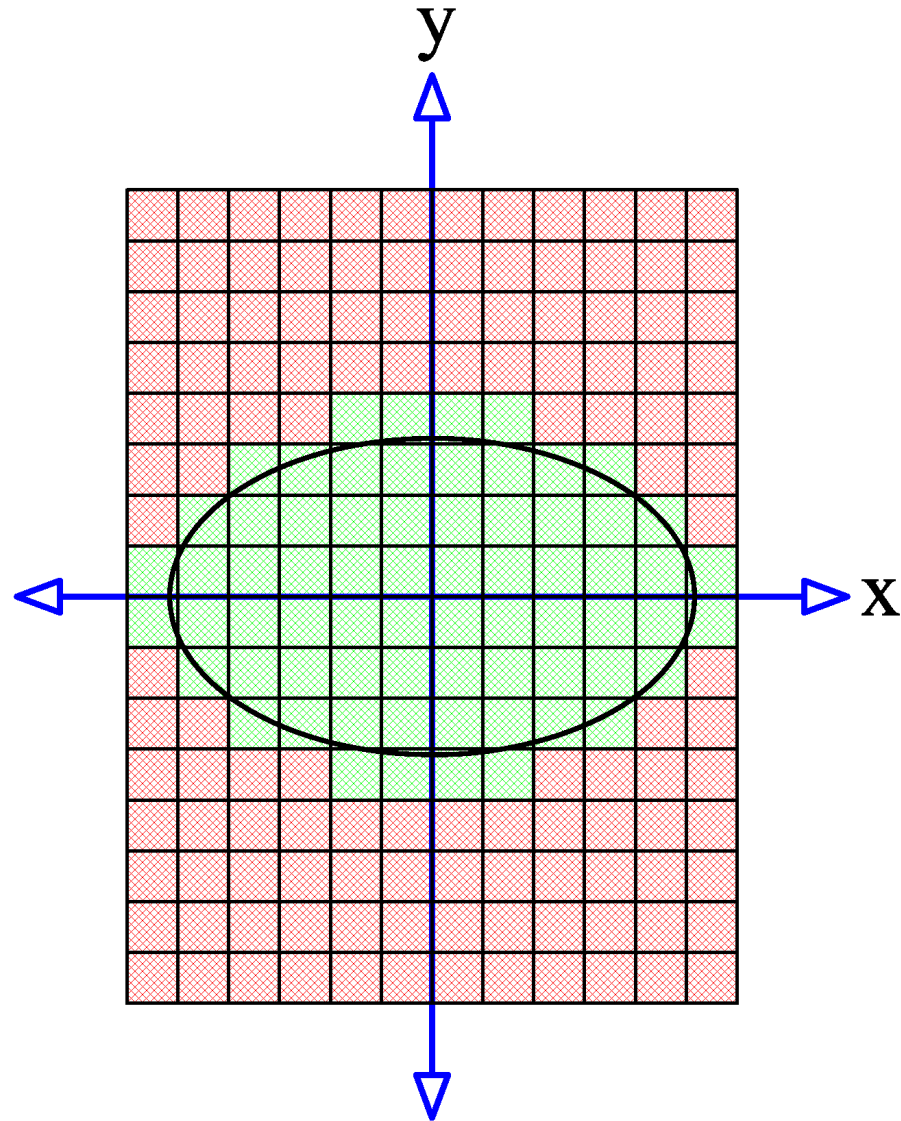
# Voxel Carving Process

- 6) Carve (remove) voxels intersected by 2<sup>nd</sup> proton
- 7) Identify voxels intersected by 3<sup>rd</sup> proton missing object



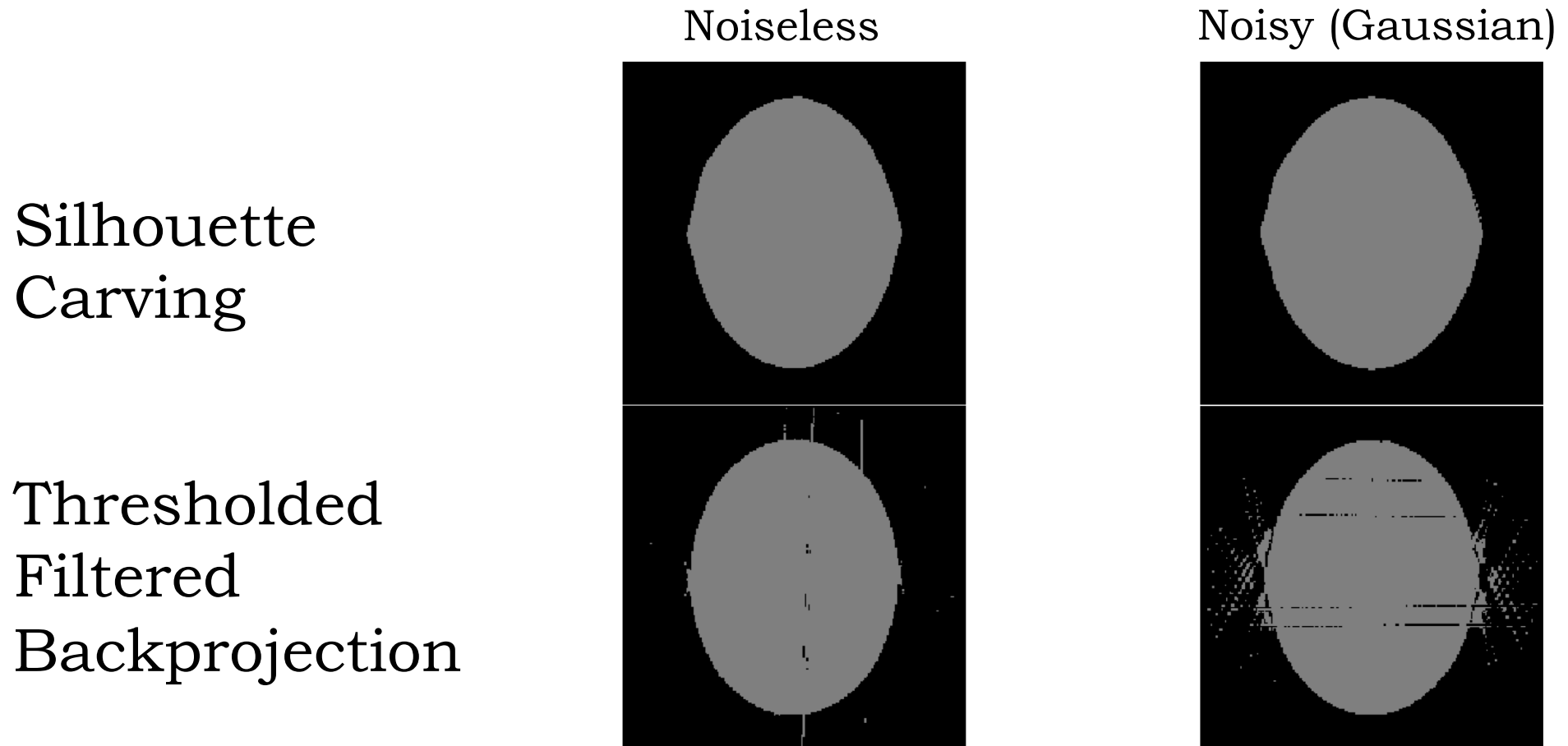
# Voxel Carving Process

- 7) Repeat preceding steps for all protons missing object
- 8) Remaining voxels define object hull



# Initial Simulated pCT Data Investigations

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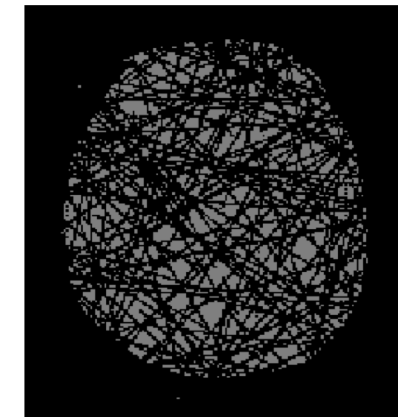


# Experimental (real) pCT Data Investigations

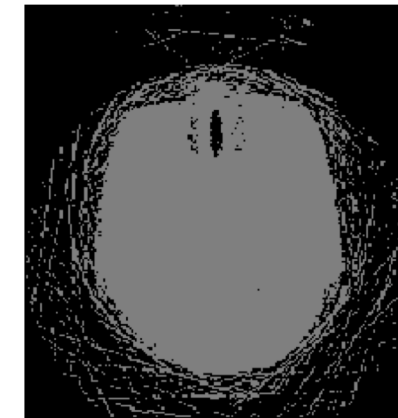
Rat Head

Pediatric Head Phantom

Silhouette  
Carving



Thresholded  
Filtered  
Backprojection



# Rectifying Observed SC Deficiency

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- 1) Silhouette Carving (SC) was adapted to use binned data after removing unsuitable proton histories
- 2) Modified Silhouette Carving (MSC) algorithm developed
  - a) Counts # of times voxel identified outside object
  - b) Uses counts to identify voxels to exclude from hull
- 3) **Space Modeling (SM) algorithm developed**
  - a) **Uses protons entering object to construct hull**
  - b) **Counts # of times voxel identified within object**
  - c) **Uses counts to identify voxels to include in hull**



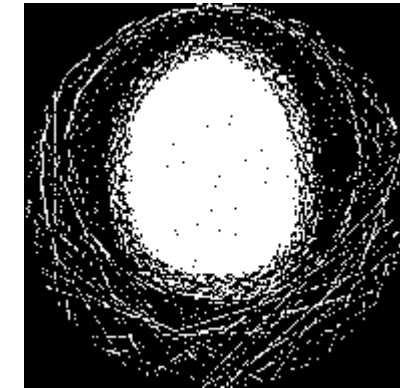
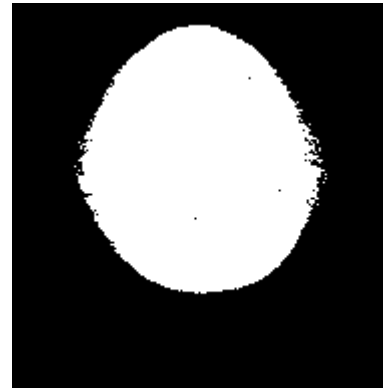
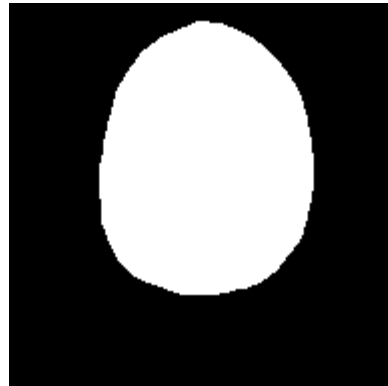
# Experimental (real) pCT Data Investigations

SC

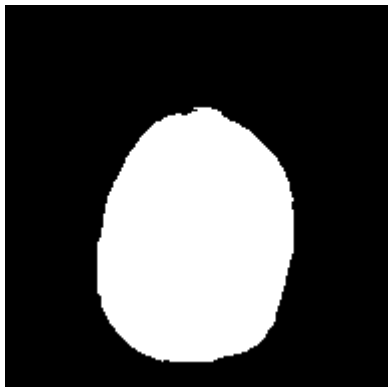
MSC

Thresholded FBP

Pediatric  
Head  
Phantom



Rat  
Head



# Hull-Detection Viability Conclusions

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- SC generated most accurate hulls, but requires unsuitable data removal & proton data binning
- MSC slightly less accurate than SC, but supports online-mode & doesn't require unsuitable data removal
- SC & MSC are viable hull-detection algorithms w/ potential for improvements

# Hull-Detection: Unpublished Updates

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- MSC is currently the preferred hull-detection algorithm as a result of the following developments:
  - much larger (~350 million protons) data sets
  - flagged protons ( $WEPL = -100$  mm) removed & no longer interfere
  - scans w/ continuous range of beamline angles
- MSC performs well under these conditions & has preferred characteristics, though it can still be improved

# MLP Formalism

Efficient

Computation

& Data

Storage

Most-Likely Path (MLP) is...

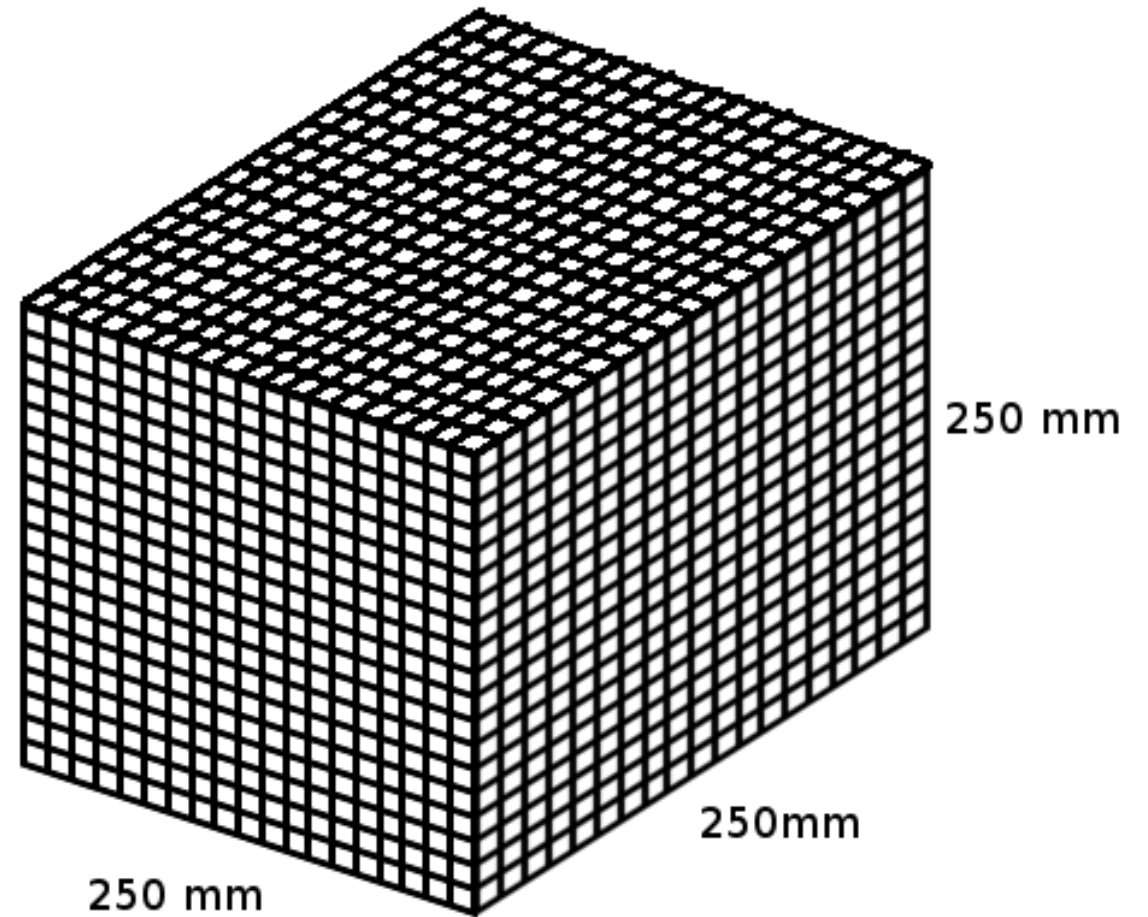
- used to approximate proton paths through object
  - used to define hyperplanes (i.e. rows of system matrix  $A \in \mathbb{R}^{m \times n}$ )
  - the most computationally expensive step of pCT
- Publication in Preparation: *IEEE Access*
- “Particle-tracking proton computed tomography—image reconstruction”

# Problem Size

$m \sim 10^9$  Proton Histories

$n \sim 10^7$  Image Space Voxels

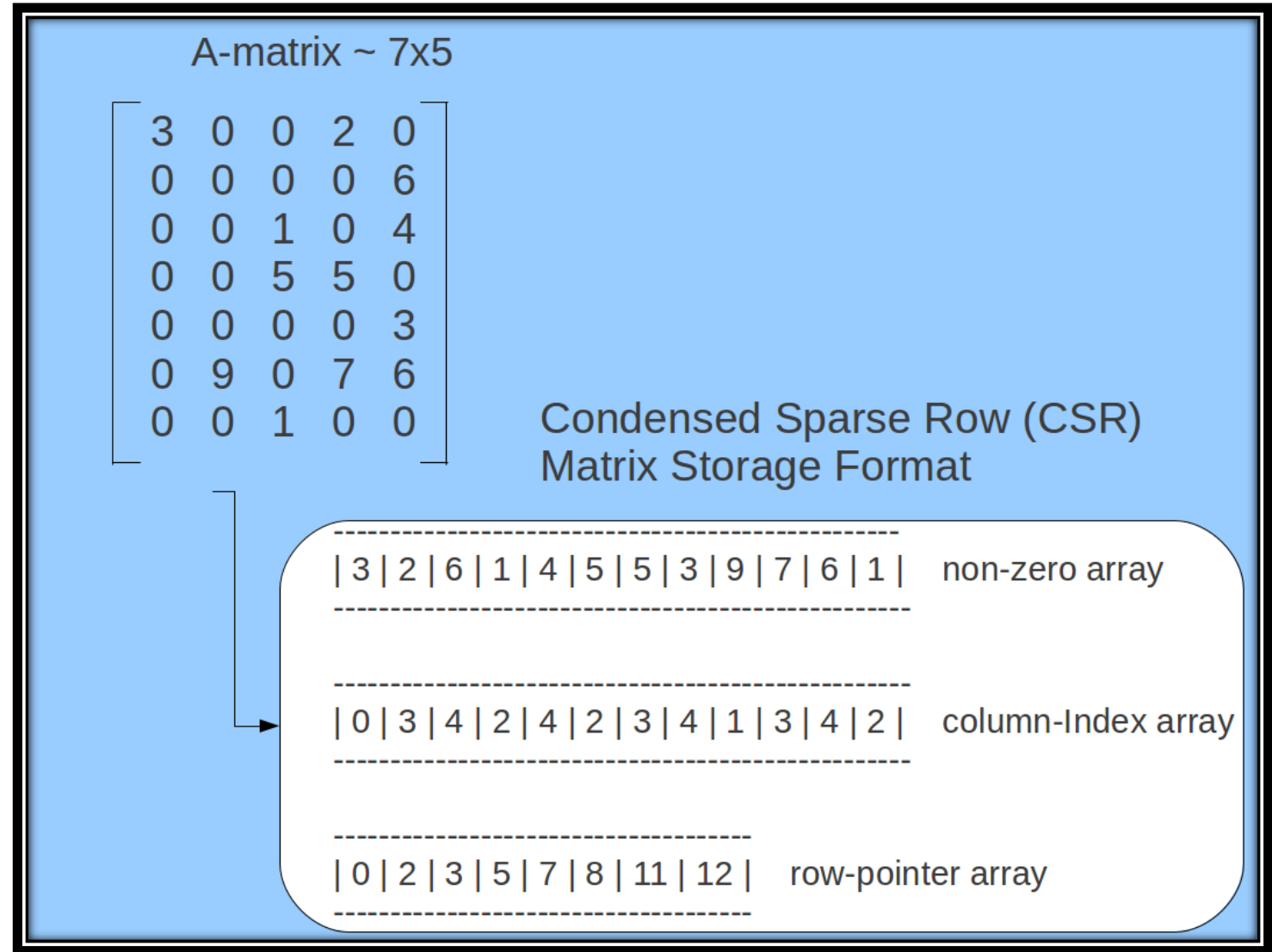
$\sim 400$  Voxels/MLP



$$A \in \mathbb{R}^{m \times n} \Rightarrow \text{size}(A) = m \cdot n = \sim 10^9 \cdot \sim 10^7 = \sim 10^{16}$$
$$\Rightarrow \text{sparsity}(A) = \sim 400 \cdot \frac{10^9}{10^{16}} = \sim 4 \cdot 10^{-5} = 0.040\%$$

# MLP

- Highly sparse  $A$  matrix
- CSR is common sparse matrix format
- Much more efficient than dense matrix storage



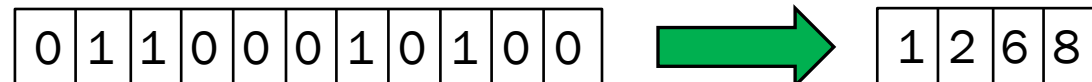
# MLP

- FS only processes individual rows of  $A$  matrix
- Each row has single value path length
- More efficient to store only the indices of each voxel intersected by MLP

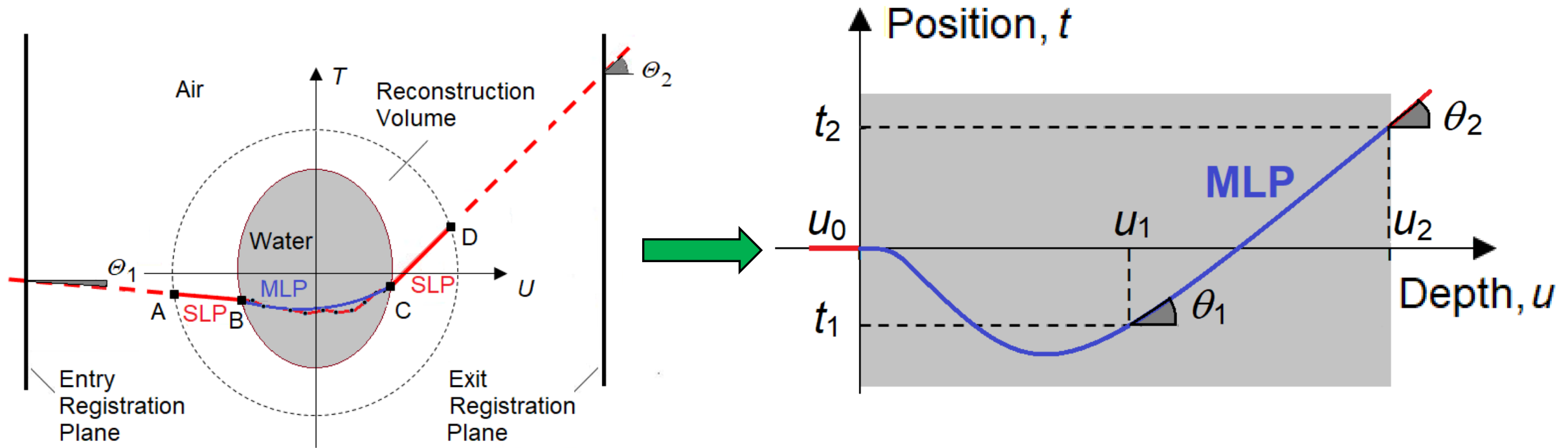
## Intersected Voxel Identification

0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
1	1	0	0	0	0	0	0	0	0	0	0	0	0	0
0	1	1	1	1	1	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	1	1	1	1	1	0	0	0	0
0	0	0	0	0	0	0	0	0	0	1	1	1	1	1
0	0	0	0	0	0	0	0	0	0	0	0	0	0	1
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
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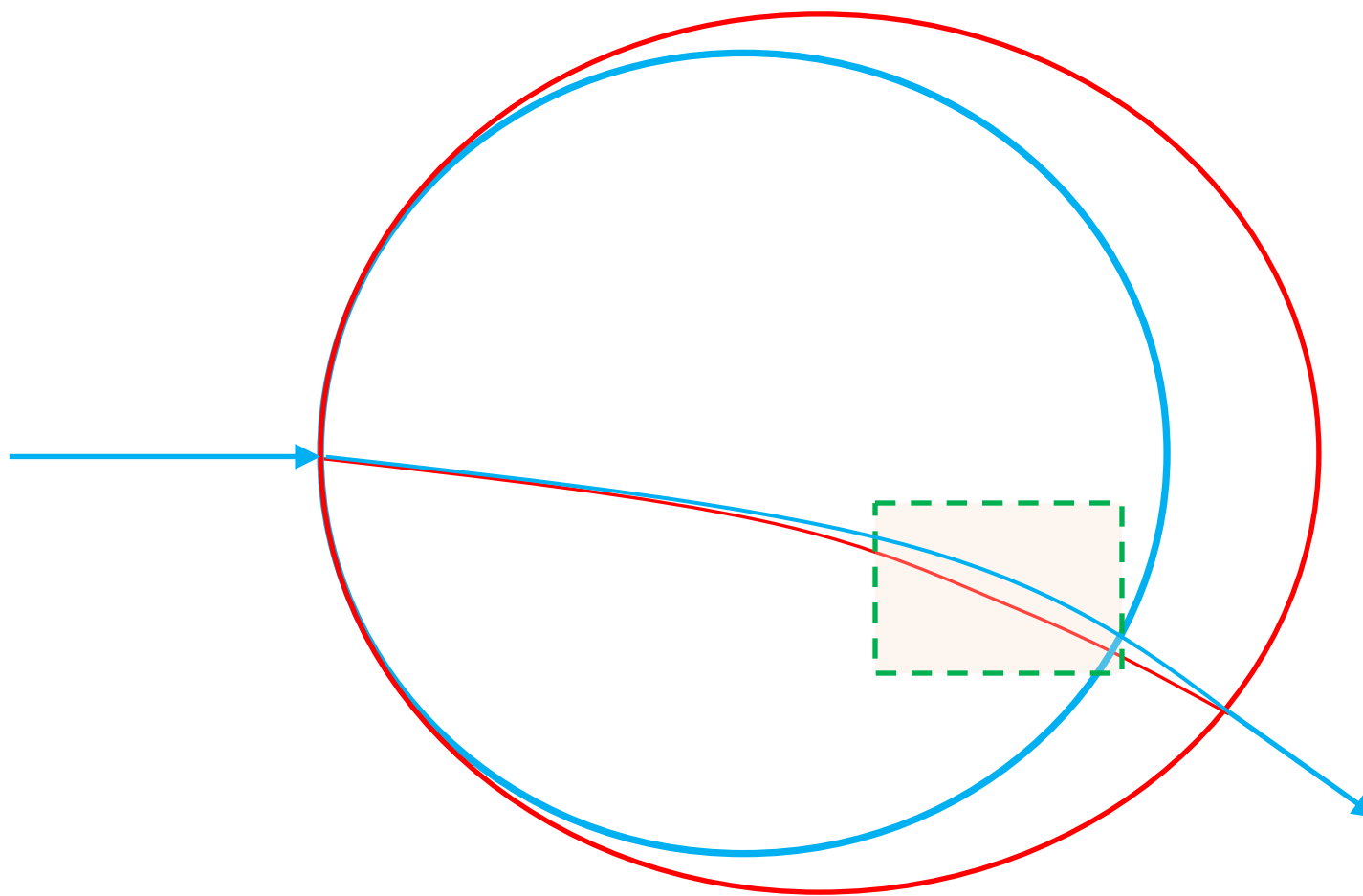
## Current MLP Storage Scheme



# MLP — Conceptual Path & Proton Coordinate System

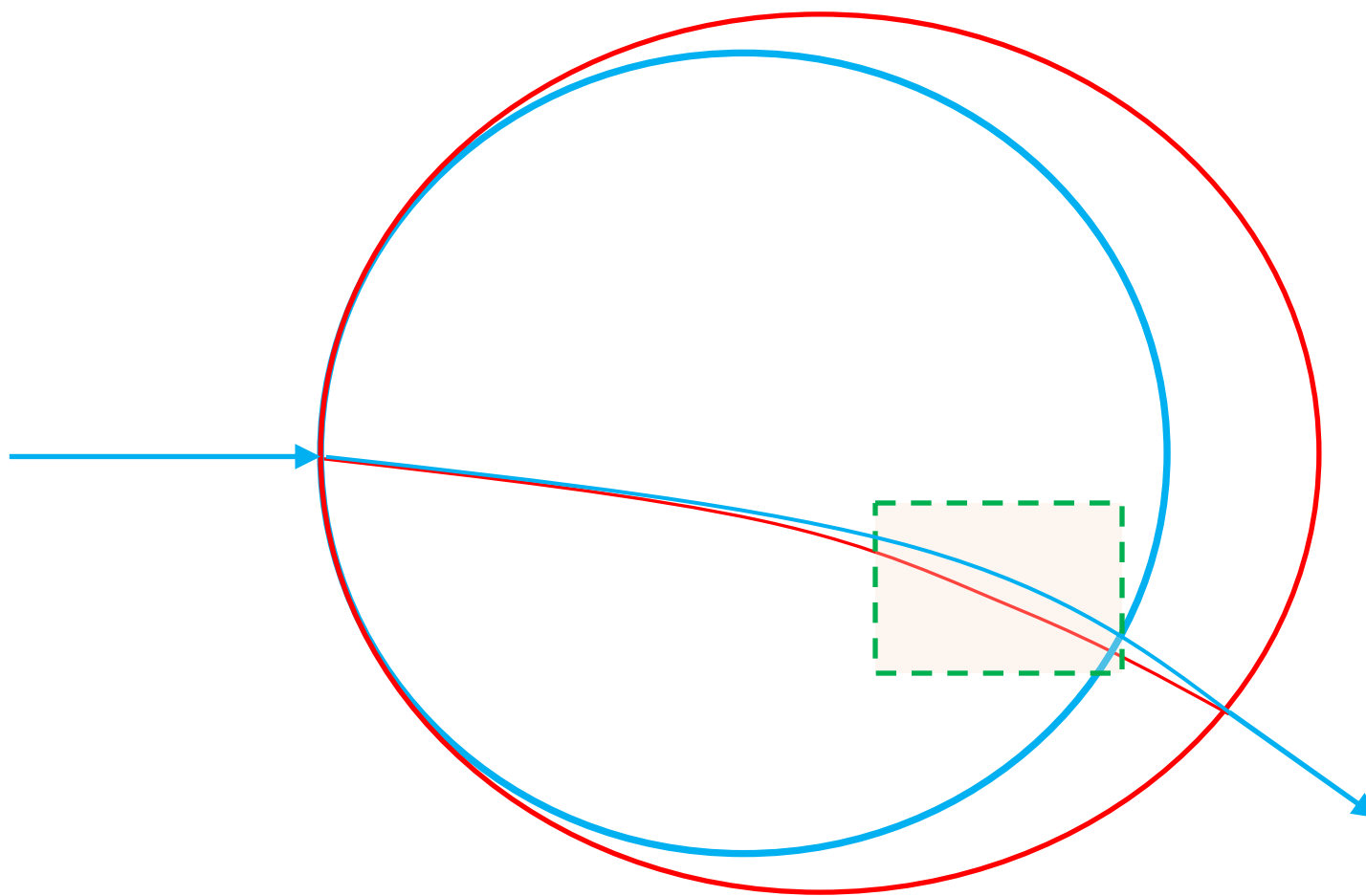






## Impact of Object Boundary Inaccuracies

- Can lead to an MLP passing through different voxels



## Impact of Entry/Exit Coordinate Inaccuracies

- Can lead to an MLP passing through different voxels (less severe)
- 3D-DDA used to determine exact hull entry/exit coordinates

# MLP

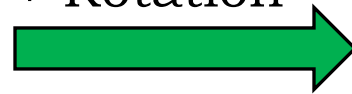
where

Equations Before  
& After Coordinate  
System  
Translation +  
Rotation

$$\vec{y}_{\text{MLP}} = \begin{bmatrix} T_1 \\ \Theta_1 \end{bmatrix} = (\Sigma_1^{-1} + R_1^T \Sigma_2^{-1} R_1)^{-1} (\Sigma_1^{-1} R_0 \vec{y}_0 + R_1^T \Sigma_2^{-1} \vec{y}_2)$$

$$\vec{y}_0 = \begin{bmatrix} T_0 \\ \Theta_0 \end{bmatrix} \quad R_0 = \begin{bmatrix} 1 & U_1 - U_0 \\ 0 & 1 \end{bmatrix} \quad \Sigma_1 = \begin{bmatrix} \sigma_{T_1} & \sigma_{T_1 \Theta_1} \\ \sigma_{T_1 \Theta_1} & \sigma_{\Theta_1} \end{bmatrix}$$
$$\vec{y}_2 = \begin{bmatrix} T_2 \\ \Theta_2 \end{bmatrix} \quad R_1 = \begin{bmatrix} 1 & U_2 - U_1 \\ 0 & 1 \end{bmatrix} \quad \Sigma_2 = \begin{bmatrix} \sigma_{T_2} & \sigma_{T_2 \Theta_2} \\ \sigma_{T_2 \Theta_2} & \sigma_{\Theta_2} \end{bmatrix}$$

Coordinate  
System  
Translation  
+ Rotation



$$\vec{y}_0 = \begin{bmatrix} t_0 \\ \theta_0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$R_0 \vec{y}_0 = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$u_2 = \cos \Theta_0 (U_2 - U_0) - \sin \Theta_0 (T_2 - T_0)$$

$$\vec{y}_2 = \begin{bmatrix} t_2 \\ \theta_2 \end{bmatrix} = \begin{bmatrix} \sin \Theta_0 (U_2 - U_0) + \cos \Theta_0 (T_2 - T_0) \\ \Theta_2 - \Theta_0 \end{bmatrix}$$

# Simplified MLP Equations

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$$\begin{aligned}\sigma_{t_1}^2(u_0, u_1) &= C(u_1) \int_{u_0}^{u_1} \frac{(u_1 - u)^2}{\beta^2(u)p^2(U) X_0} du \\ &= C(u_1)P_1(u_1)\end{aligned}$$

$$\begin{aligned}\sigma_{\theta_1}^2(u_0, u_1) &= C(u_1) \int_{u_0}^{u_1} \frac{1}{\beta^2(u)p^2(u) X_0} du \\ &= C(u_1)P_2(u_1)\end{aligned}$$

$$\begin{aligned}\sigma_{t_1\theta_1}^2(u_0, u_1) &= C(u_1) \int_{u_0}^{u_1} \frac{u_1 - u}{\beta^2(u)p^2(u) X_0} du \\ &= C(u_1)P_3(u_1)\end{aligned}$$

$$\sigma_{t_2}^2(u_1, u_2) = C(u_2 - u_1) [P_1(u_2) - u_2^2 P_3(u_1) + 2u_2 P_4(u_1) - P_5(u_1)]$$

$$\sigma_{\theta_2}^2(u_1, u_2) = C(u_2 - u_1) [P_2(u_2) - u_2 P_3(u_1) + P_4(u_1)]$$

$$\sigma_{t_2\theta_2}^2(u_1, u_2) = C(u_2 - u_1) [P_3(u_2) - P_3(u_1)]$$

where  $C(U) = E_0^2 \left[ 1 + 0.038 \ln \left( \frac{U}{X_0} \right) \right]^2$

$$\begin{bmatrix} P_1 \\ P_2 \\ P_3 \\ P_4 \\ P_5 \end{bmatrix} = \begin{bmatrix} 0 & 0 & \frac{a_0}{3} & \frac{a_1}{12} & \frac{a_2}{30} & \frac{a_3}{60} & \frac{a_4}{105} & \frac{a_5}{168} \\ 0 & \frac{a_0}{2} & \frac{a_1}{6} & \frac{a_2}{12} & \frac{a_3}{20} & \frac{a_4}{30} & \frac{a_5}{42} & 0 \\ a_0 & \frac{a_1}{2} & \frac{a_2}{3} & \frac{a_3}{4} & \frac{a_4}{5} & \frac{a_5}{6} & 0 & 0 \\ 0 & \frac{a_0}{2} & \frac{a_1}{3} & \frac{a_2}{4} & \frac{a_3}{5} & \frac{a_4}{6} & \frac{a_5}{7} & 0 \\ 0 & 0 & \frac{a_0}{3} & \frac{a_1}{4} & \frac{a_2}{5} & \frac{a_3}{6} & \frac{a_4}{7} & \frac{a_5}{8} \end{bmatrix} \begin{bmatrix} u \\ u^2 \\ u^3 \\ u^4 \\ u^5 \\ u^6 \\ u^7 \\ u^8 \end{bmatrix}$$

# MLP Implementation

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- Scattering matrices  $\Sigma_1$  &  $\Sigma_2$  are either:
  - calculated at run time each iteration
  - constructed from lookup table ( $\Sigma_1^{-1}$  &  $\Sigma_2^{-1}$  as well)
- Redundant computations involving fixed  $(t_2, u_2)$  precalculated for each MLP

# MLP Computational Efficiency

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Impact of MLP implementation on single intranode GPU reconstructions:

- ~70% reduction in compute operations
- achieved reconstruction < 10 min clinical feasibility threshold for 1<sup>st</sup> time
- allows reconstruction times ~6 min on single GPU, ~2 min on 4-GPUs

# TVS

Applying  
Modern  
Algorithm  
to pCT

## Total variation superiorization is...

- governed by the superiorization methodology (SM) w/ TV as cost function
- a perturbation method for reducing noise w/ minimal edge degradation
- interleaved between FS steps
- preserves convergence of underlying algorithm (e.g. feasibility-seeking) if perturbation resilience is obeyed
- Published: *IEEE Transactions on Medical Imaging* (2020)

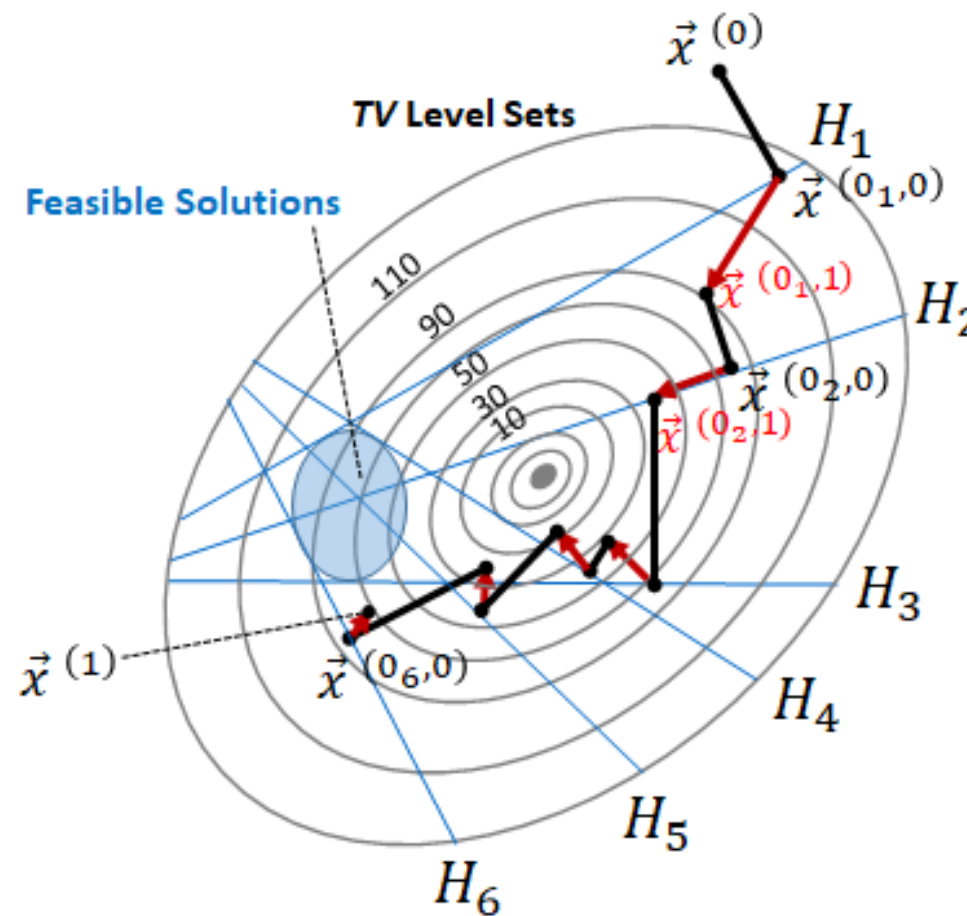
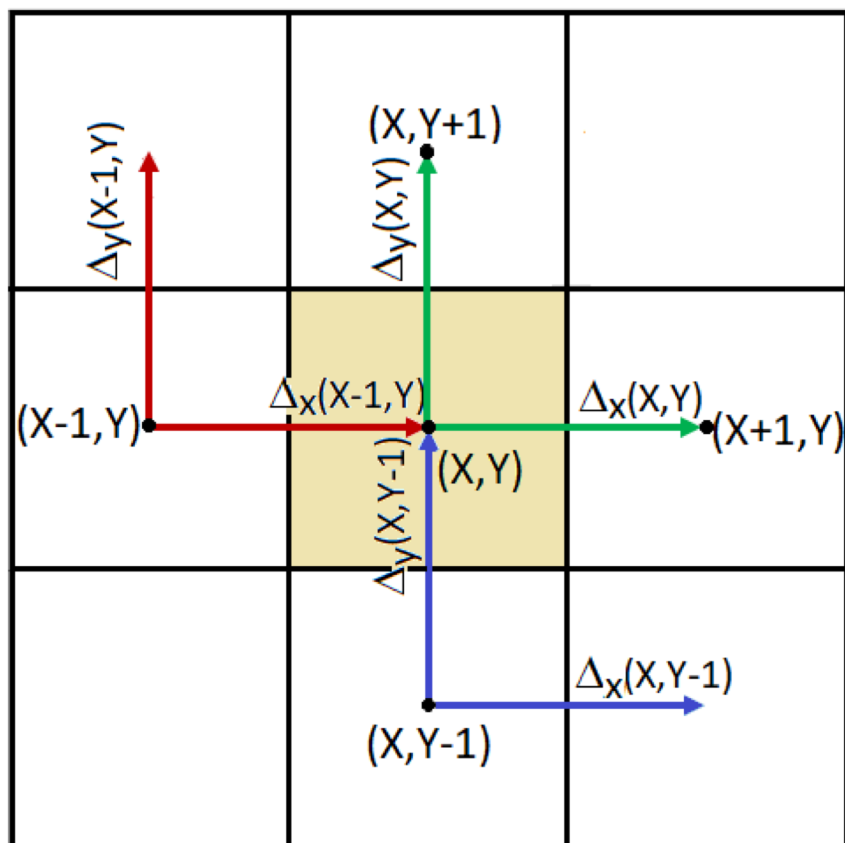
An Improved Method of Total Variation Superiorization Applied to Reconstruction in Proton Computed Tomography



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# TVS Perturbation Calculations & Application Impact





# Modernization of TVS Algorithm Used in pCT

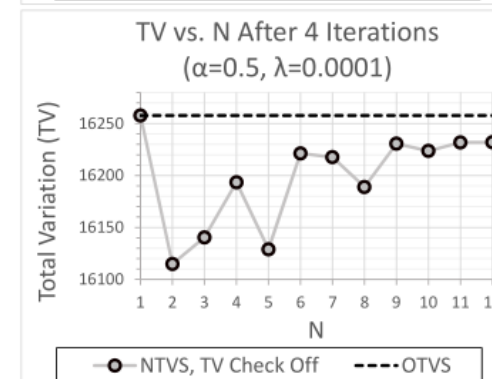
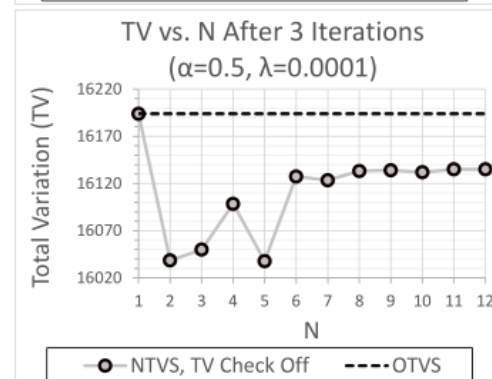
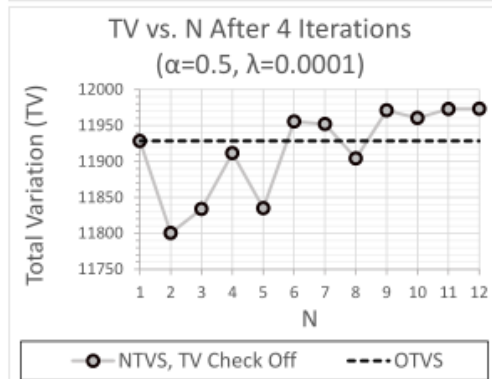
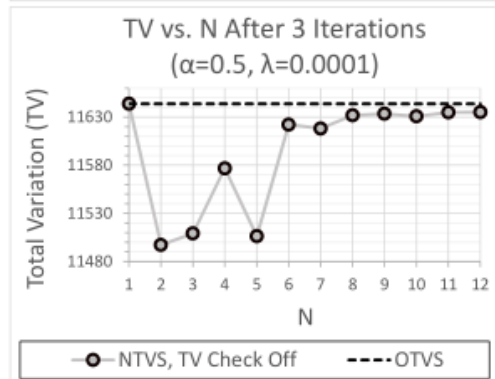
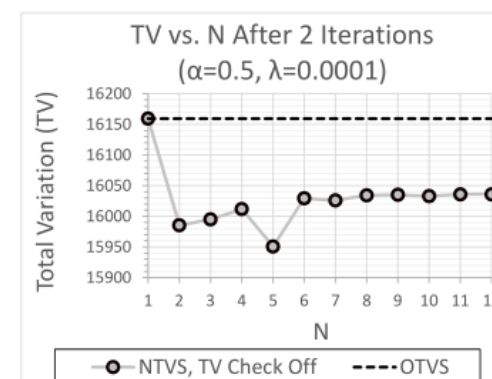
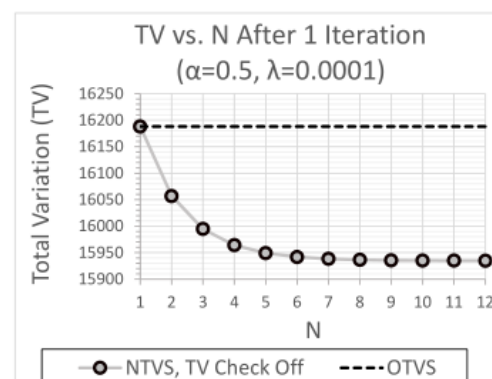
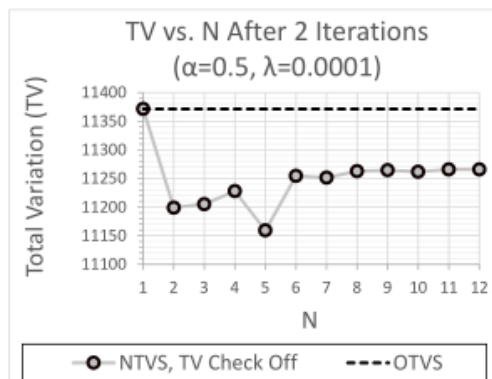
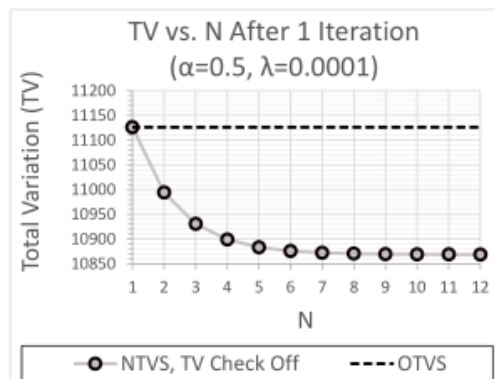
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- TVS had been shown to benefit pCT imaging but had not been modernized
- Advances in SM led to investigations of the following changes:
  - removal of total variation reduction verification step (TVRVS)
  - added  $N$  perturbations per FS iteration
  - perturbation step-size  $\beta \leftarrow \beta/2$  replaced by  $\beta = \alpha^\ell$ 
    - ( $0 < \alpha < 1$  &  $\ell \leftarrow \ell + 1$  after each perturbation)
  - added random decrease in  $\ell \leftarrow \text{rand}(k, \ell)$  at each FS iteration  $k$

# Experimental Results: Analysis of $N$

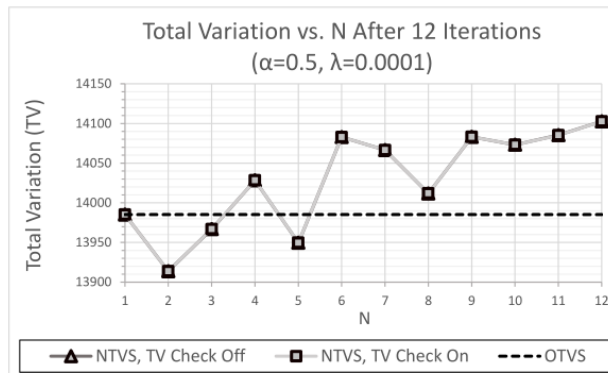
## CTP404 Module

## HN715 Head Phantom

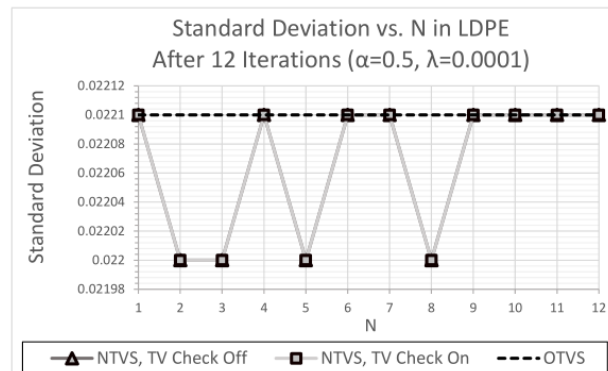


# Experimental Results: Analysis of TVRVS ( $\alpha = 0.5$ )

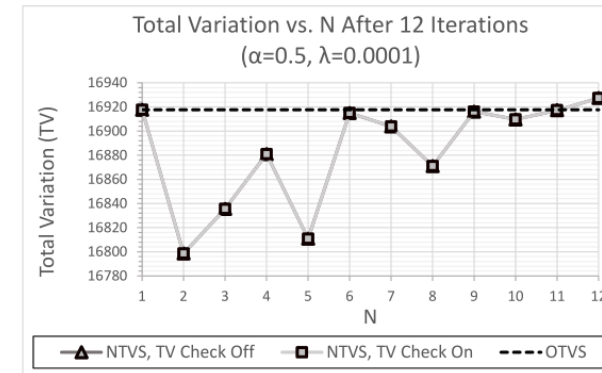
## CTP404 Module



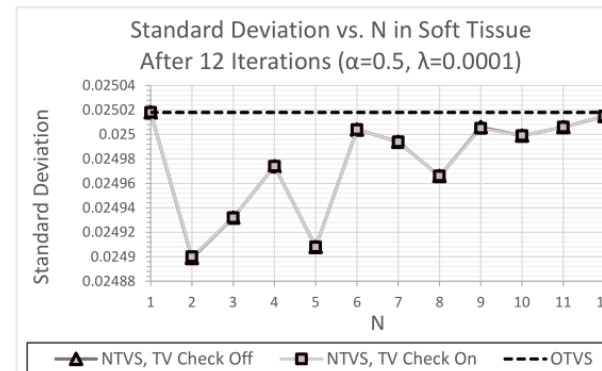
(a)



## HN715 Head Phantom

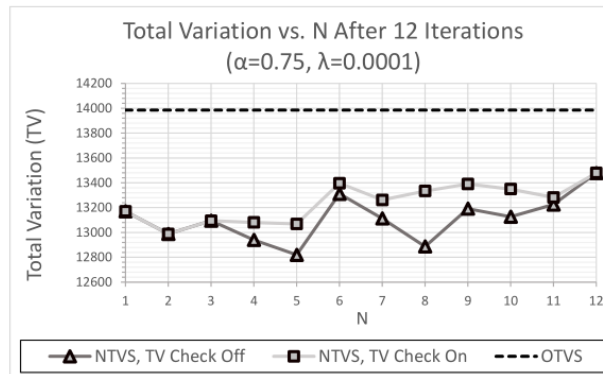


(a)

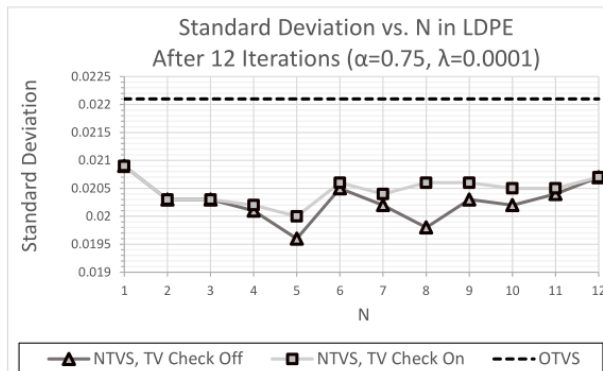


# Experimental Results: Analysis of TVRVS ( $\alpha = 0.75$ )

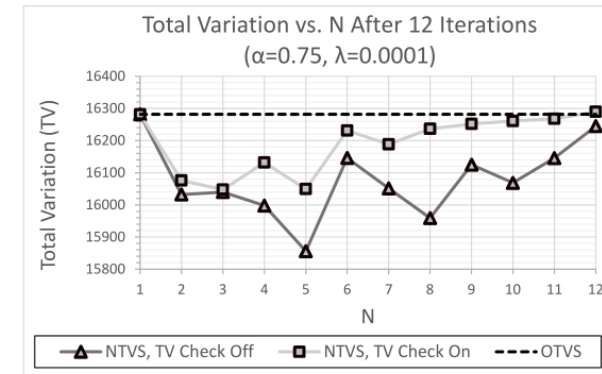
## CTP404 Module



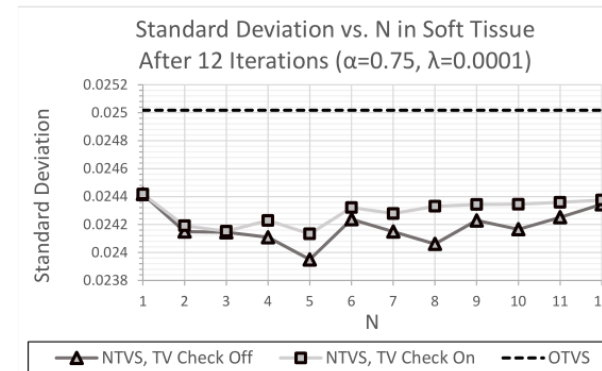
(a)



## HN715 Head Phantom

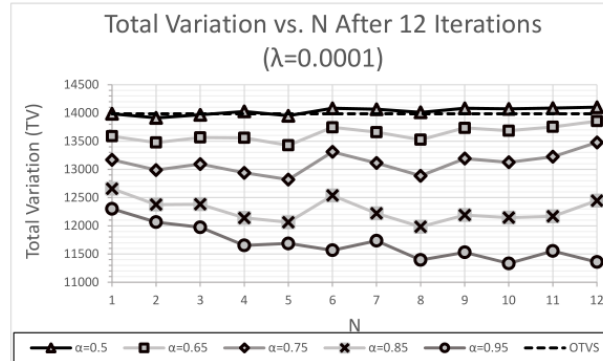


(a)

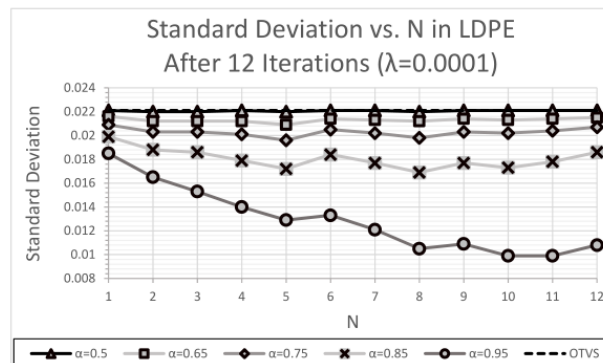


# Experimental Results: Analysis of Perturbation Kernel $\alpha$

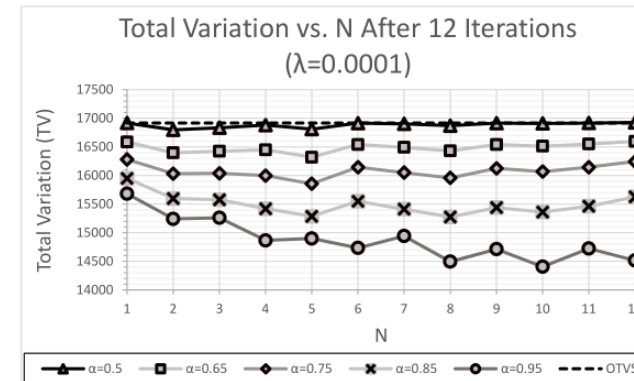
## CTP404 Module



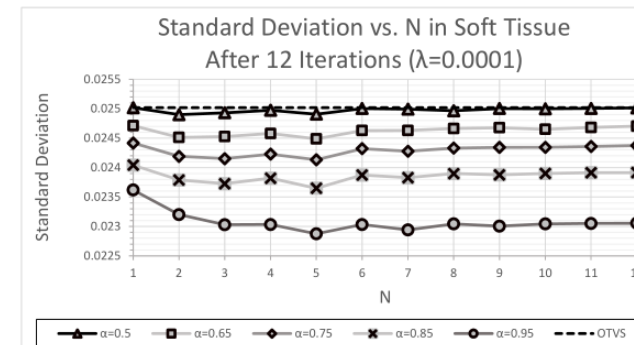
(a)



## HN715 Head Phantom

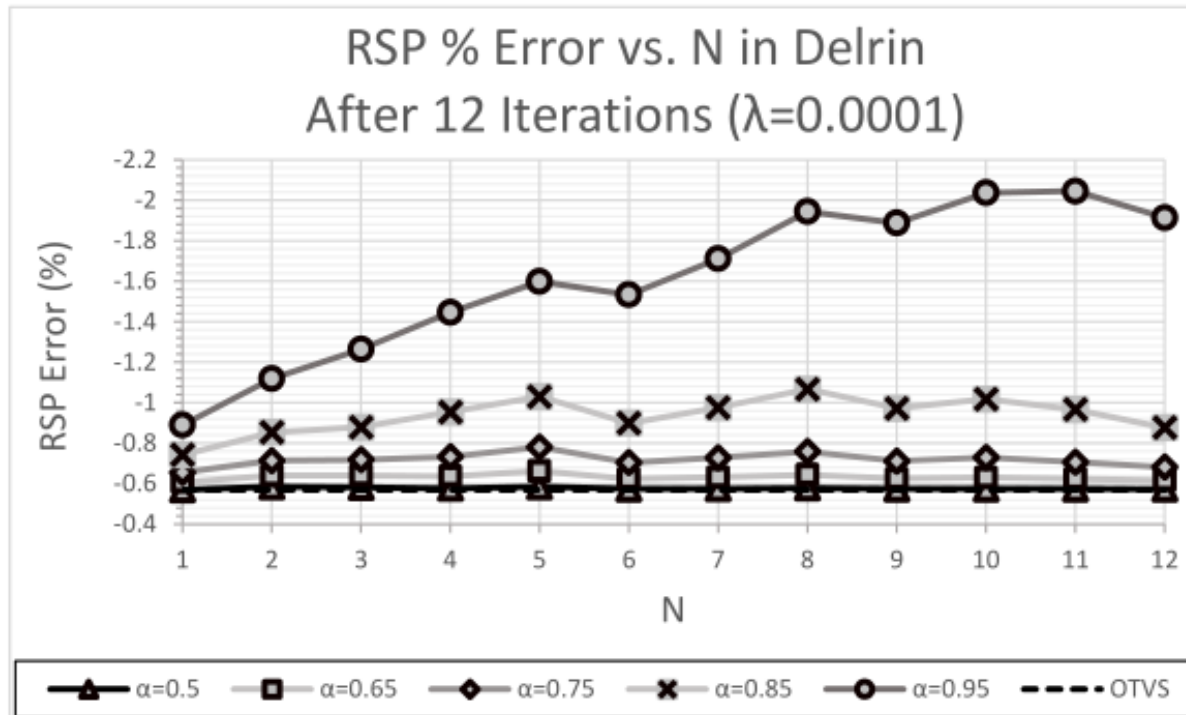


(a)

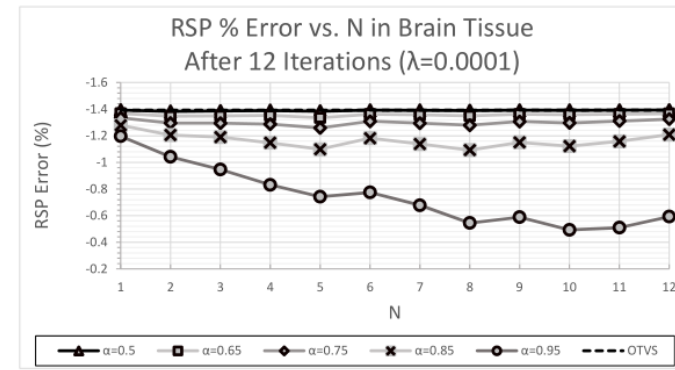


# Experimental Results: Analysis of Perturbation Kernel $\alpha$

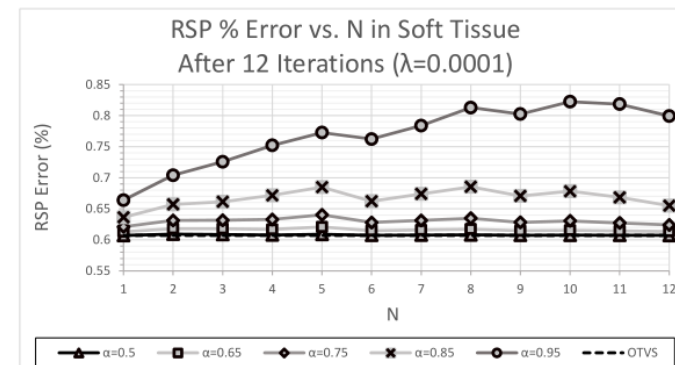
## CTP404 Module



## HN715 Head Phantom

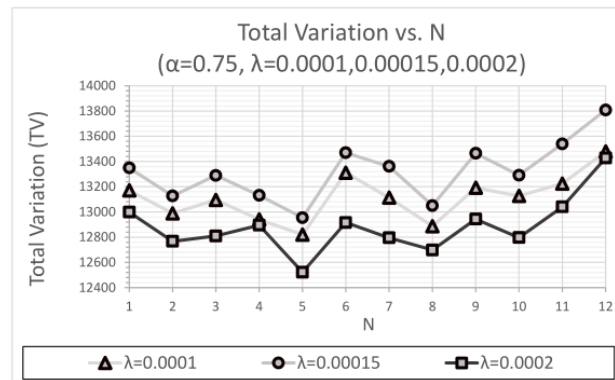


(b)

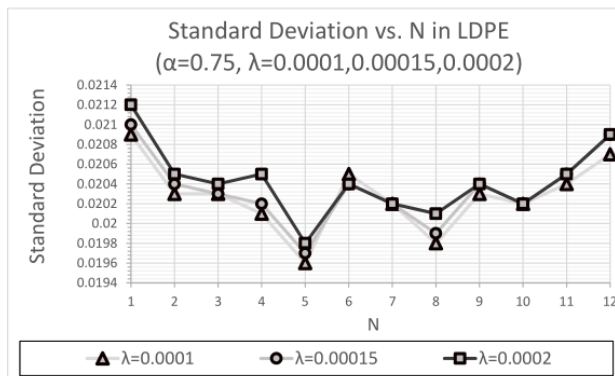


# Experimental Results: Analysis of FS Relaxation $\lambda$

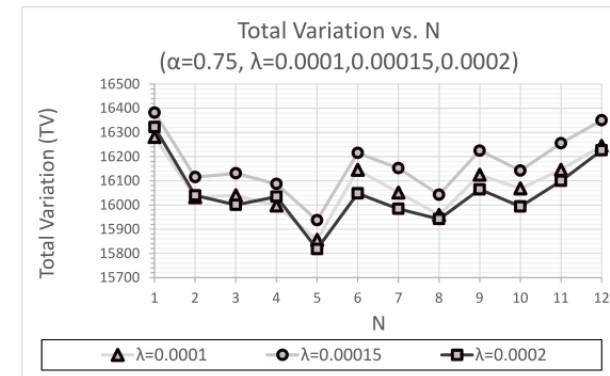
## CTP404 Module



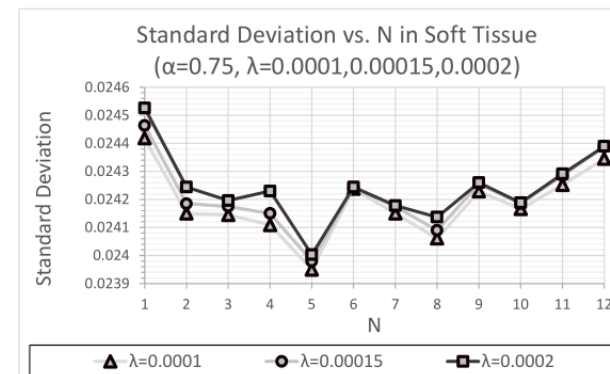
(a)



## HN715 Head Phantom



(a)



# Modern TVS Conclusions

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- Removal of TVRVS not harmful, actually often improves TV &  $\sigma_{RSP}$
- $N$  perturbations per FS iteration beneficial when  $3 \leq N \leq 6$ , w/  $N = 5$  optimal most often. Performance diminishes for  $N \geq 7$ , potentially from  $\ell$  increasing too quickly
- $\alpha > 0.5$  offers increasing benefits w.r.t. TV &  $\sigma_{RSP}$ , but  $\alpha > \sim 0.75$  increasingly and unpredictably affects RSP convergence,  $\Rightarrow \alpha = 0.75$  recommended
- $\lambda$  can potentially be increased w/o typically observed increase in noise
- future investigations of approaches to uncouple  $N$  &  $\beta$  (via  $\ell$ ) relationship are needed



# Questions or Comments?

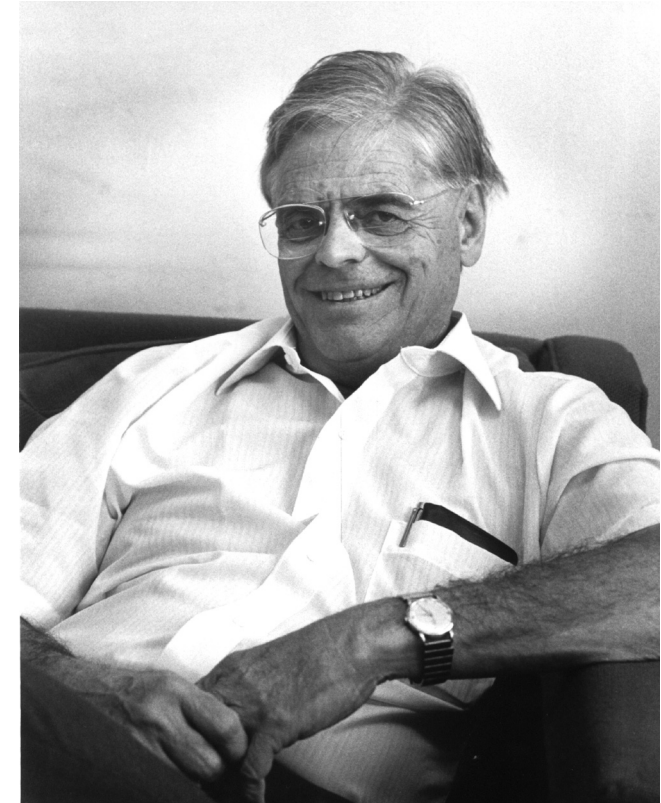
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# Acknowledgements

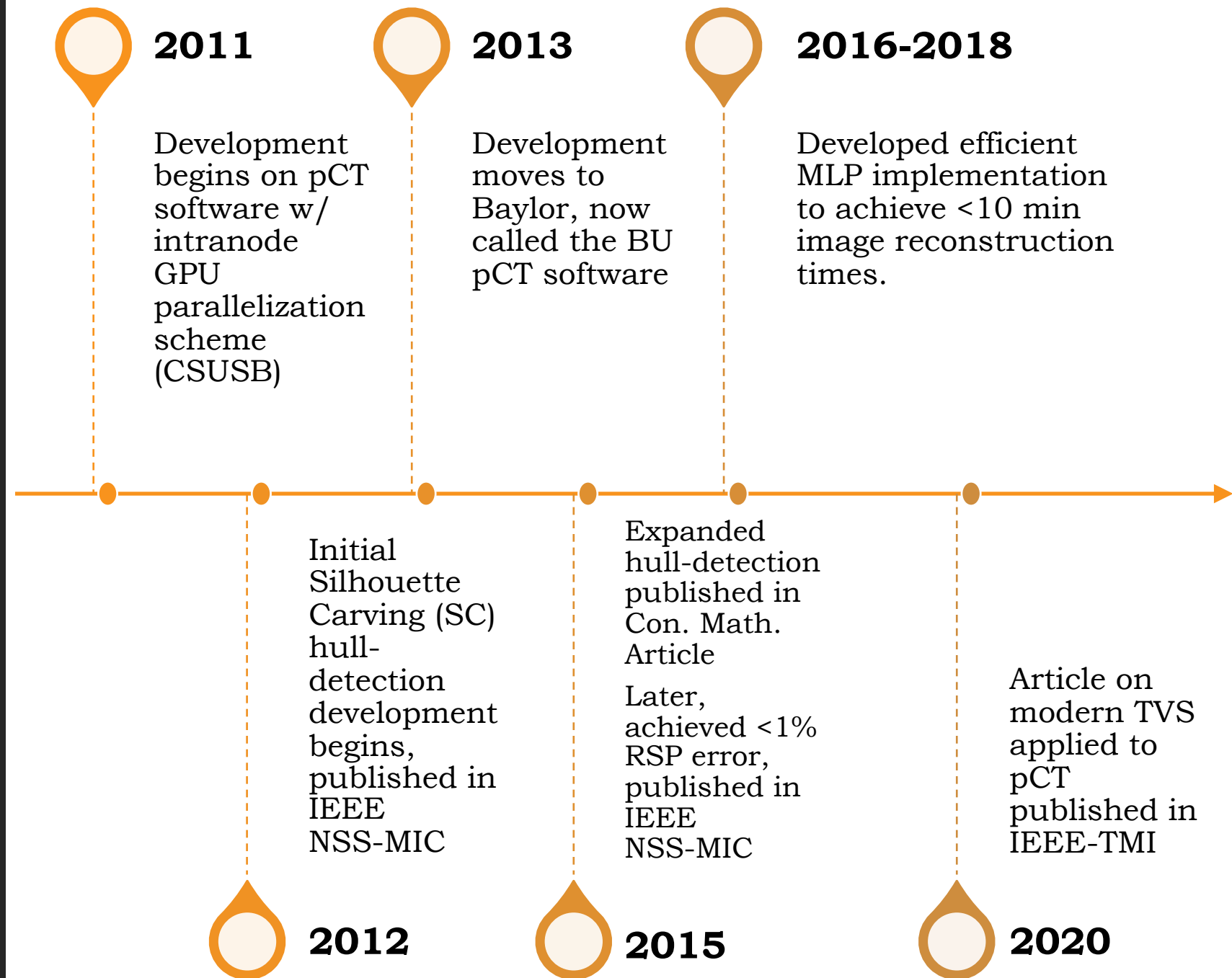
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- The content of this presentation is solely the responsibility of the authors and does not necessarily represent the official views of NIBIB, NIH and NSF.

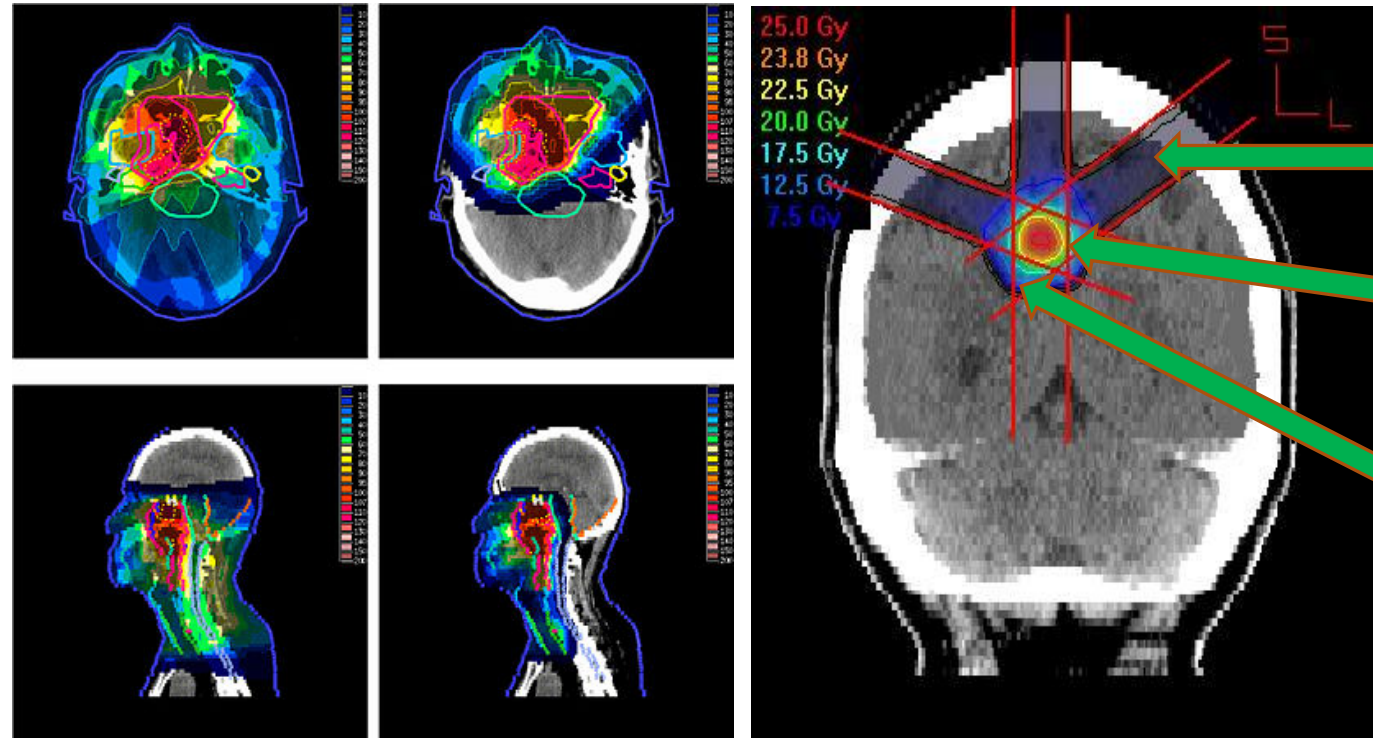
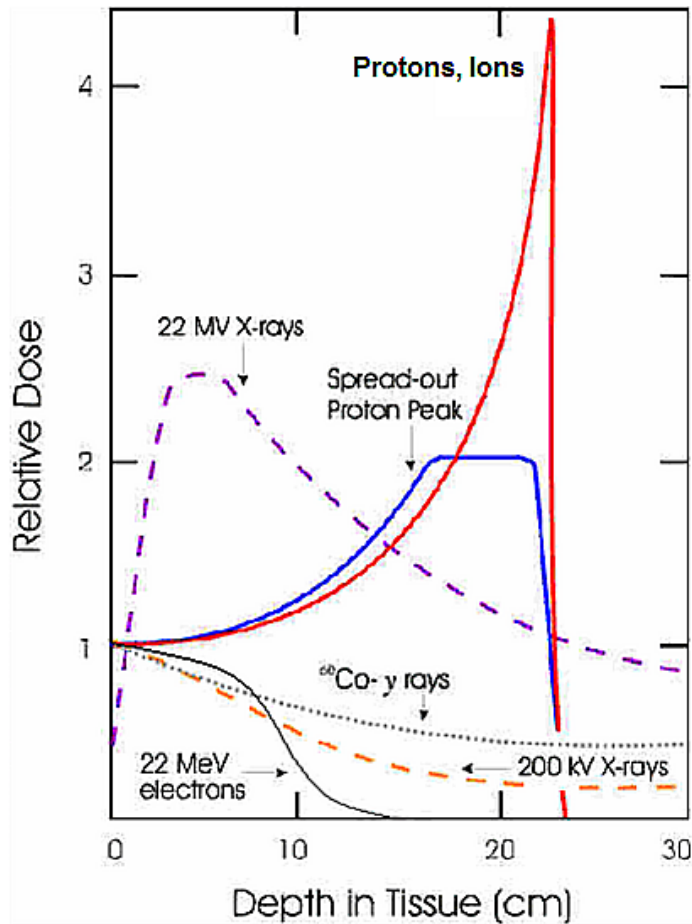


Robert R. Wilson, Ph.D., 1914-2000

# Timeline of pCT Work & Publications



# Advantages of Proton Therapy

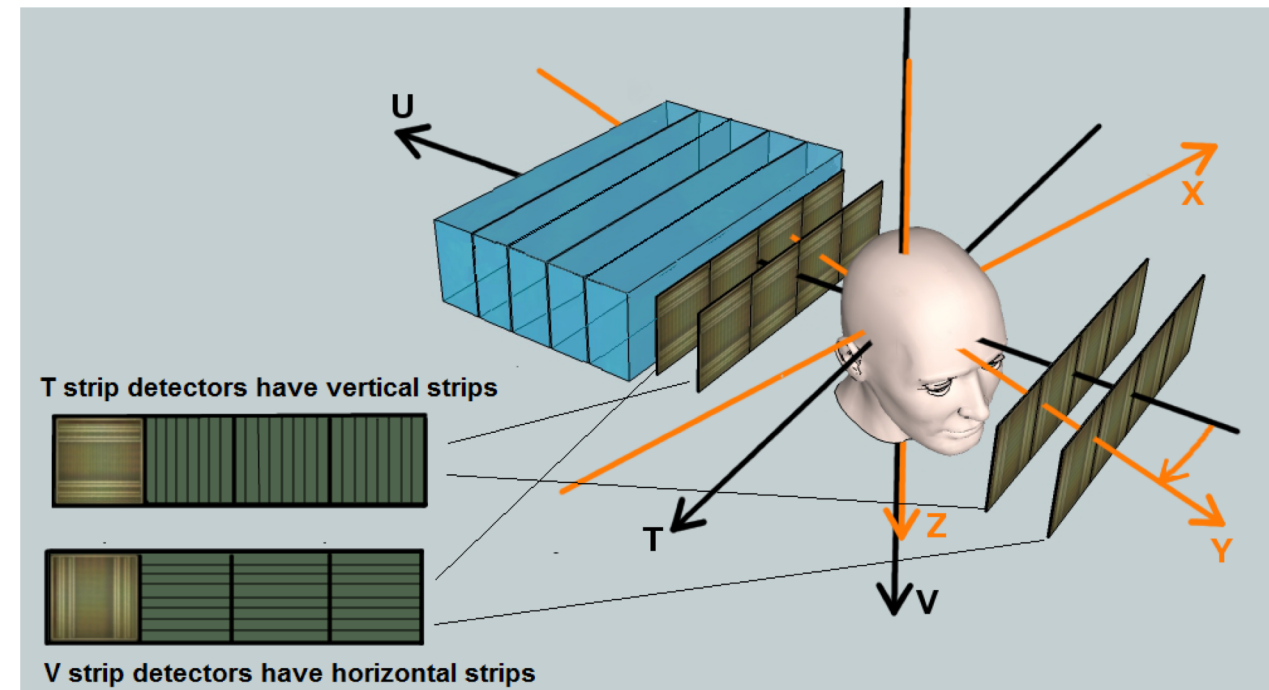
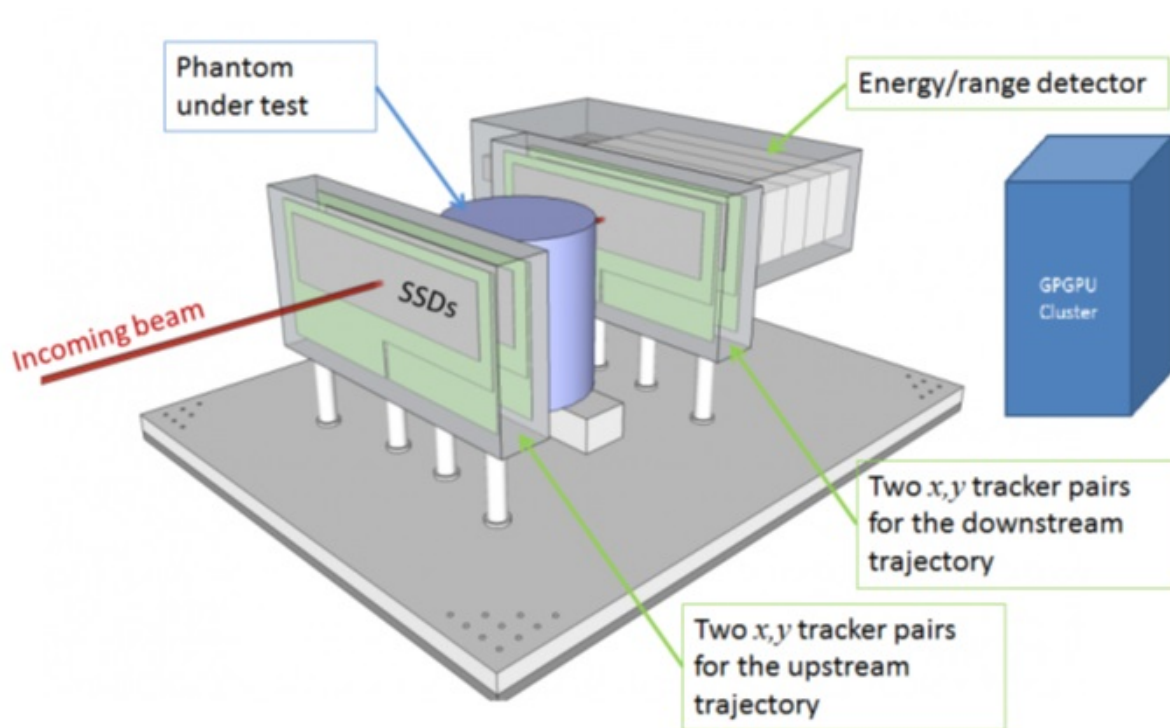


# Why pCT for Proton Therapy Treatment Planning

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- Conversion of Hounsfield units (HU) to relative stopping power (RSP) adds range uncertainty
- HU-RSP conversion is especially unreliable for some materials
- Common practice 3.5% + 1 mm additional margin to nominal range of proton beam exceeds clinical/planning uncertainty margins of photon therapy
- pCT offers ability to reconstruct RSP directly
- pCT imaging dose much lower than xCT, offering potential for weekly or even daily imaging

# pCT Scanner & Data/Image Coordinate Systems



# pCT

## Fundamentals: Measurement Data & Resulting Linear System

- pCT image reconstruction yields relative stopping power (RSP) map of object
- Energy measurements converted to water-equivalent path length (WEPL) for convenience since

$$\text{WEPL} = \int_{\text{entry}}^{\text{exit}} \text{RSP}(x, y, z) d\ell$$

$$\Rightarrow b_i = \sum_{\forall j} \text{RSP}_j \cdot a_{ij}$$

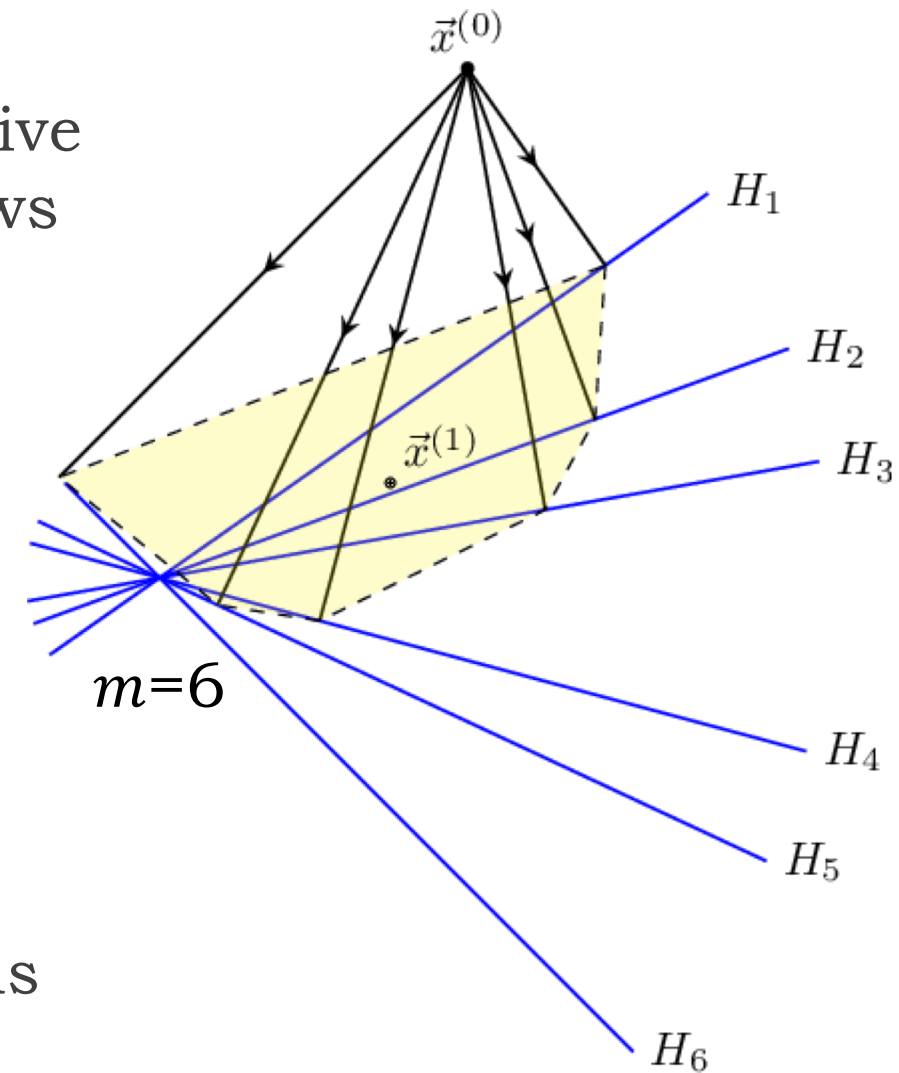
- Linear pCT system  $A\vec{x} = \vec{b}$  solved using feasibility-seeking (FS) algorithms

# Feasibility-Seeking (FS)

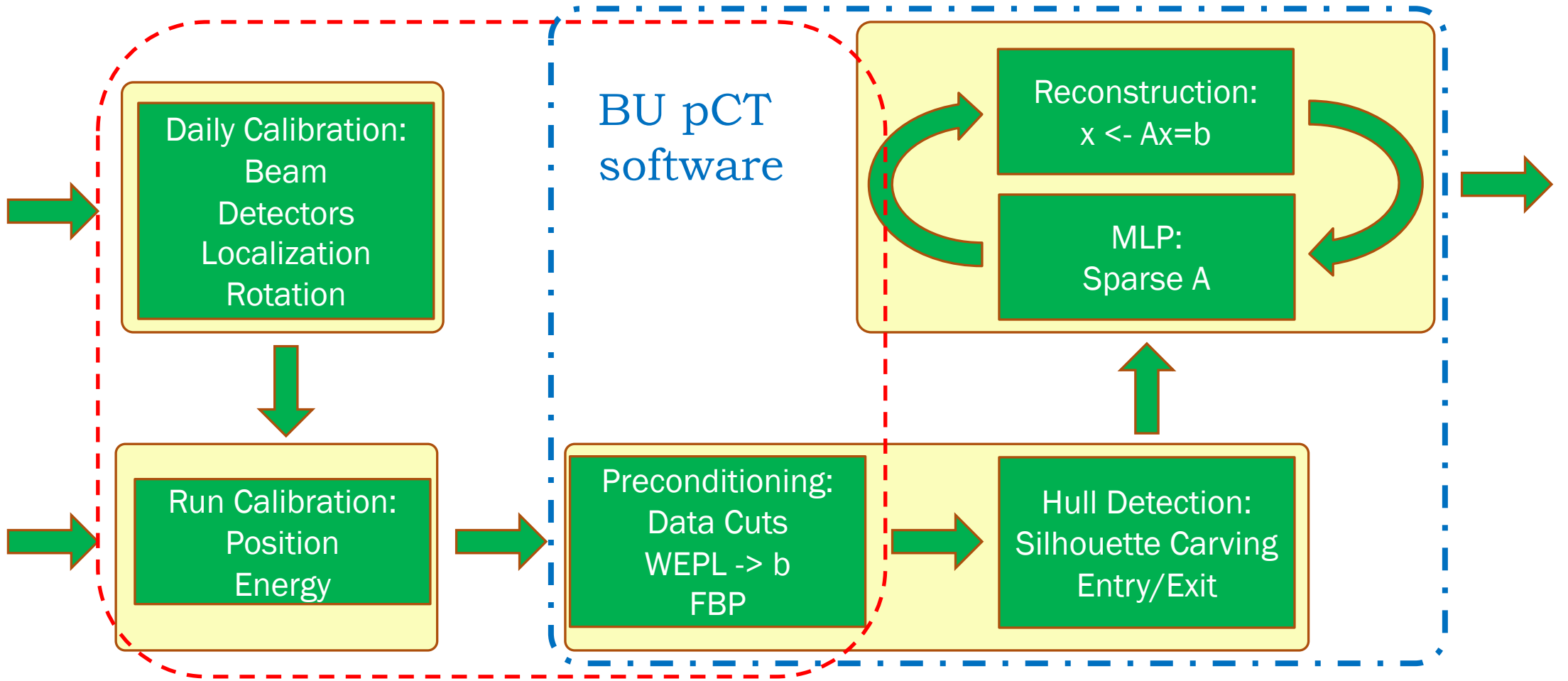
Fully Simultaneous  
(Cimmino type) FS  
Methods:

Fully Simultaneous  
DROP Algorithm

- FS performs iterative projections onto rows of  $A$ , advancing toward feasible (not exact) solution
- Cimmino type FS algorithms project onto all  $m$  hyperplanes simultaneously
- Fully Simultaneous DROP (FS-DROP) used herein







# Data Acquisition & Preprocessing

Detailed description published in

Insert	Predicted RSP	Mean RSP	% Discrepancy	Std. Deviation
PMP	0.8770	0.8788	0.1993	0.0201
LDPE	0.9973	0.9990	0.1721	0.0182
Polystyrene	1.0386	1.0390	0.0414	0.0178
Acrylic	1.1550	1.1635	0.7285	0.0199
Delrin	1.3560	1.3532	-0.2059	0.0193
Teflon	1.8280	1.8150	-0.7161	0.0212

# RSP Accuracy Achieved

FS-DROP w/ TVS

# Original Object Boundary Detection Scheme

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## Filtered Backprojection Thresholding

### Deficiencies of filtered backprojection (FBP) thresholding:

- FBP images noisy & artifact prone due to inappropriate SLP projections
- difficult to select RSP threshold
- produces object interior “holes” where material RSP < threshold
- not useful for excluding voxels outside object
- slow & cannot be performed until FBP complete

# Silhouette Carving (SC) Method

- 3D-DDA  
Development

- 3D-DDA based on 2D ray marching method digital difference analyzer (DDA)
- Fundamentally different than binary decision based steps of 2D algorithms
- Developed numerically stable 3D-DDA, was initially prone to numerical error
- Efficiently & accurately identifies intersected voxels and intersection coordinates