

## Agenda

- Background
  - IRCU 50
  - Uncertainties
  - Robust optimization
- Use Cases
  - Lung Robust 4D Optimization
  - Breast skin flash
  - Craniospinal Independent beam robustness
- Evaluation
- Future of Robustness

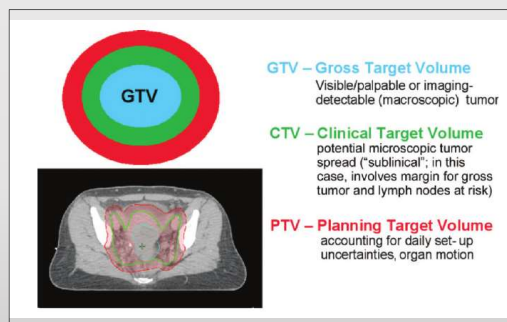
### - Disclaimer -

I am a RaySearch employee, images and tool sets presented are from RayStation.

I will attempt to provide a neutral perspective on the current and future status of Robust Optimization.

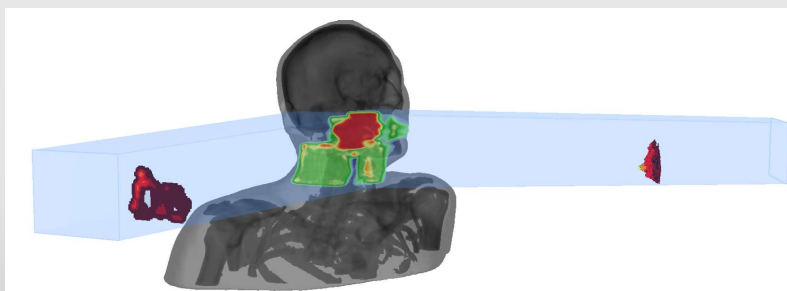
## ICRU 50

- Purpose: to promote the use of a common language for specifying and reporting the doses in radiation therapy, as well as the volumes in which they are prescribed
  - GTV & CTV are purely oncological concepts of actual disease or possible extension of disease
  - PTV takes into account the movement of the patient and the inaccuracies due to patient-beam positioning or related to the therapy equipment.

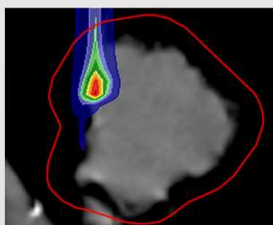


## Background

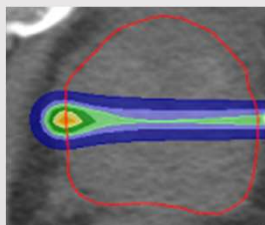
- Primary goal:
  - Dose coverage despite geometric uncertainty
- Conventional solution:
  - Planning margins
- Underlying assumption:
  - Changing the anatomy doesn't change the dose (the static dose cloud approximation)



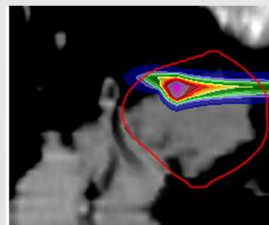
## EXAMPLES OF IMPT DOSES IN DIFFERENT GEOMETRIES



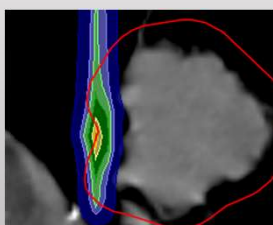
(a) Nominal setup



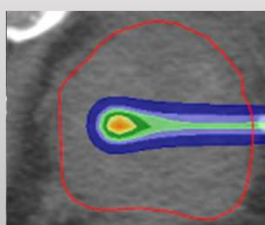
(c) Nominal density



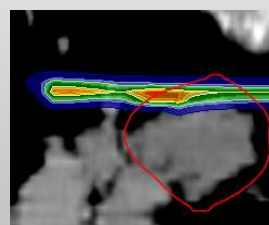
(e) Nominal tumor position



(b) Shifted setup



(d) Scaled density



(f) Shifted tumor position

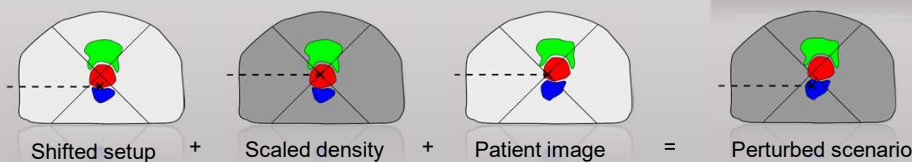
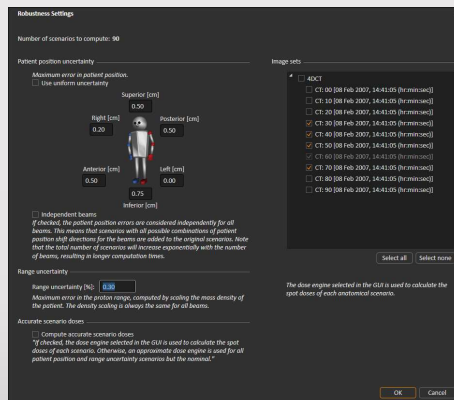
## What to do?

- The underlying hypothesis fails, so margins won't work as intended
  - Some heuristics (SFUD and possibly material overrides) may help
- Is Robust Optimization the answer?
  - Robust optimization is a field of optimization theory that deals with optimization problems in which a certain measure of robustness is sought against uncertainty that can be represented as deterministic variability in the value of the parameters of the problem itself and/or its solution.



## What is robust optimization

- Explicitly create scenarios
  - Positioning change
  - Density Error
  - Organ and target motion
- Optimize on covering all scenarios
- To do so, uses a new algorithm!



## Robust Optimization

- Calculate the the number of combinations
  - Number of combinations = (+,-,0 range shift) x (setup)x(number of images)
    - +5mm range with (3mm... -3mm... 0mm...)
    - 5mm range ...
- Minimize the worst case scenario (maximum)
  - Fluence - x
  - Scenarios - k
  - Dose - d

$$\underset{x}{\text{minimize}} \quad \max_k [f(d^k)]$$

## Range and Setup Uncertainty

### Setup uncertainty

The patient may not be placed in the correct position each day

There are mechanical tolerances that determine some physical limits

### Range uncertainty

The CT scan contains uncertainty as a model of the patient

Weight loss or weight gain

Conversion of Hounsfield units to stopping power

Imaging artifacts

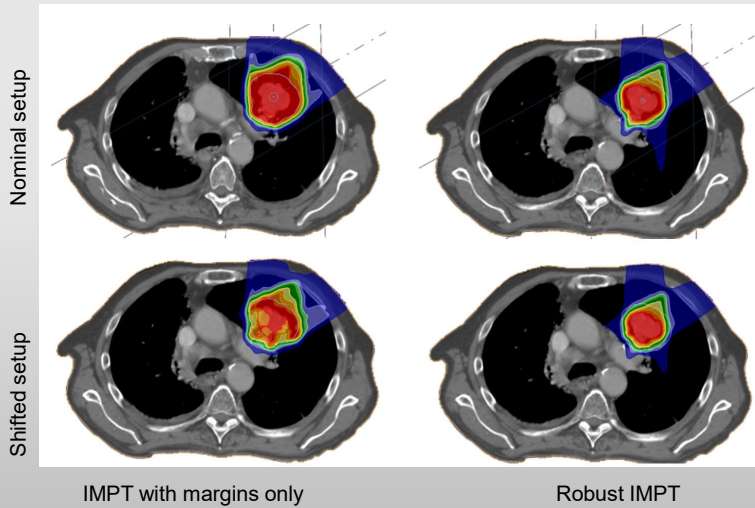
## MINIMAX OPTIMIZATION

- minimax optimization to achieve robustness

$$\begin{aligned} & \underset{x \in \mathcal{X}}{\text{minimize}} && f_0(d(x; s_0)) + \max_{s \in \mathcal{S}} f(d(x; s)) \\ & \text{subject to} && g_0(d(x; s_0)) \leq 0, \\ & && g(d(x; s)) \leq 0, \quad s \in \mathcal{S} \end{aligned}$$

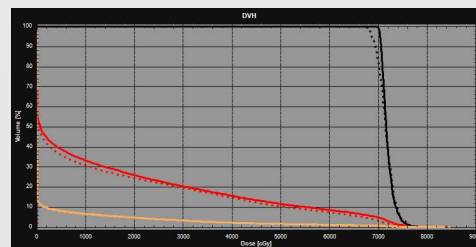
$f_0$	— nominal objectives
$f$	— robust objectives
$g_0$	— nominal constraints
$g$	— robust constraints
$s_0$	— nominal scenario
$\mathcal{S}$	— set of scenarios
$d(x; s)$	— dose as function of variables $x$ under scenario $s$

## RESULTS: Lung case

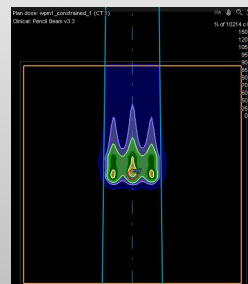


## SCENARIO selection

- How to select the scenario  $\max_{s \in S} f(d(x; s))$
- For small setup errors, use scenarios along main axes
- For larger errors, include additional directions
- Include intermediate setup and density scenarios



DVHs for plan using robust constraint and axis scenarios only.  
 Solid: nominal scenario Dashed: (0, 0.7, 0.7) cm setup shift



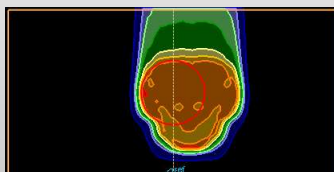
Small target, large uncertainty,  
 no intermediate scenarios

## SCENARIO selection

- Setup errors, density errors, and organ motion are generally independent



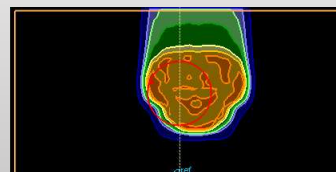
Example not using independent setup and density errors:



Setup error



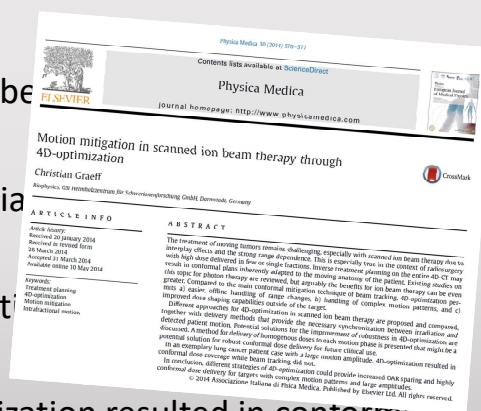
Density error



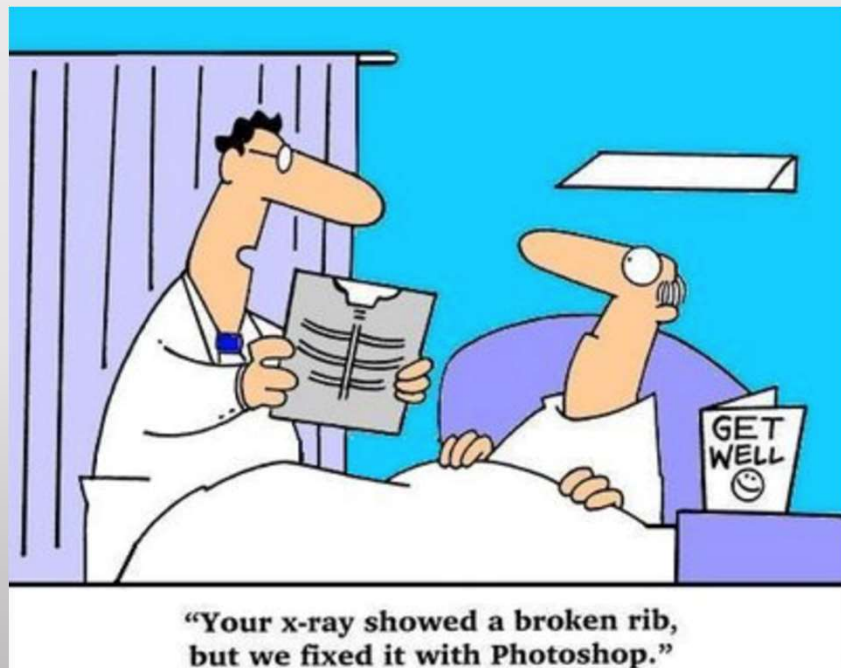
Setup and density error

## Lung: Robust 4D-CT optimization

- For lung cases, respiratory motion must be
- Conventionally: ITVs, average CTs, materia
- Robust optimization with respect to multi
- Independent study from GSI: “4D-optimization resulted in conformal dose coverage while beam tracking did not.”

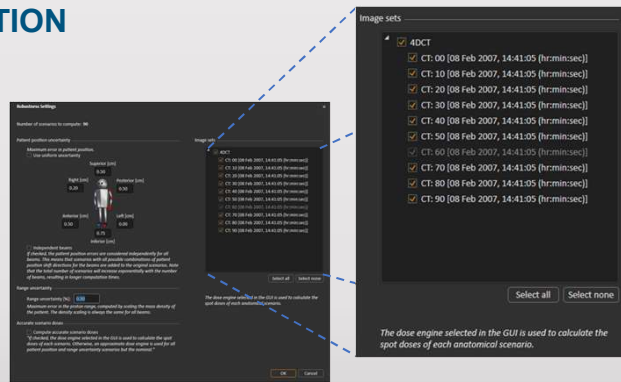


# Case Discussion

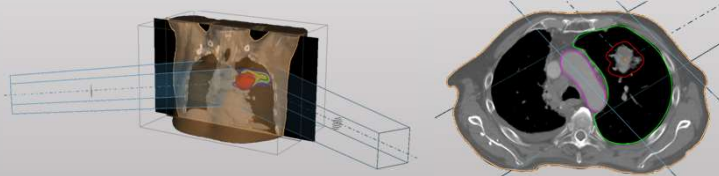


## LUNG: ROBUST 4D-CT OPTIMIZATION

- Select phases over which to optimize
- Combinable with setup and range robustness: allows for CTV planning
- Requires contours on all selected image sets



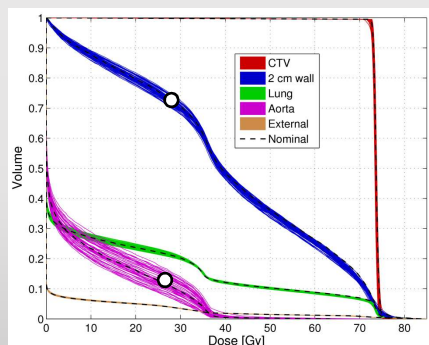
Example: IMPT for lung





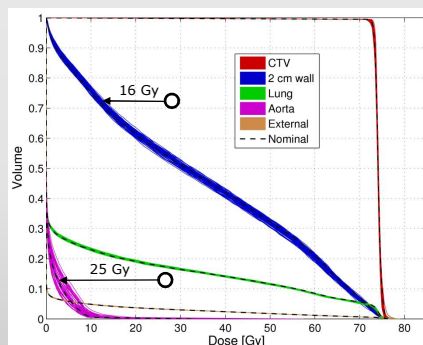
## LUNG: ROBUST 4D-CT OPTIMIZATION

Evaluation over 50 scenarios:



### Conventional planning

- Single field uniform dose
- Material override
- Margins for setup errors
- ITV for respiratory motion



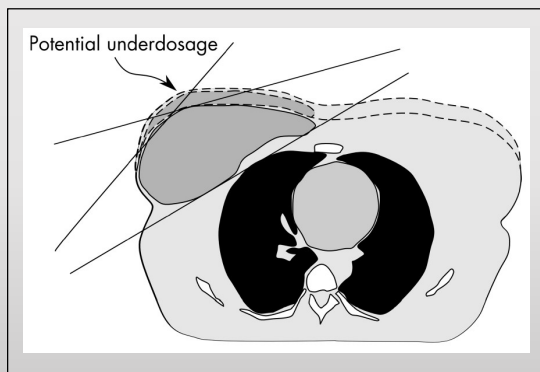
### Robust optimization

- 4D-CT and setup scenarios

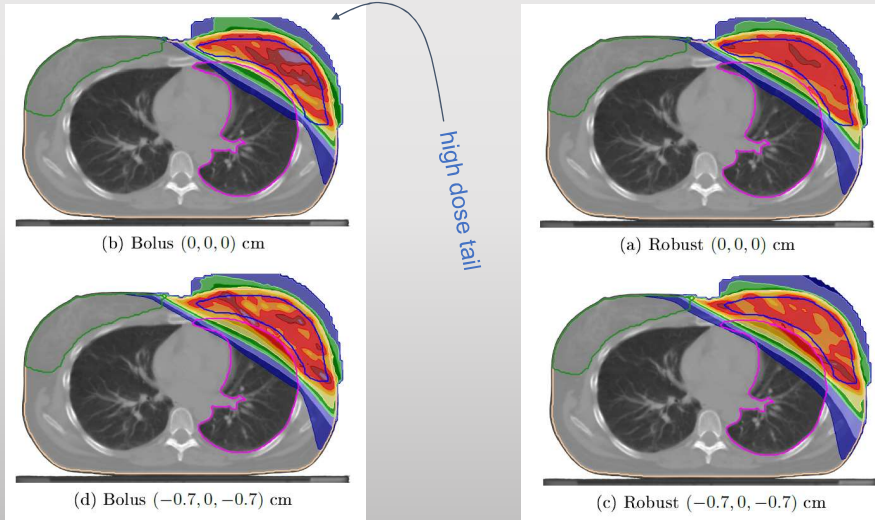
Albin Fredriksson and Björn Hårdemark, Robust optimization accounting for organ motion, range errors, and setup errors in IMPT, 11th Biennial On Physics & Radiation Technology For Clinical Radiotherapy, 2011

## Breast: skin flash

- In the skin flash region, the static dose cloud approximation doesn't even hold for photons
- Heuristic solutions
  - Expanding fields (3D-CRT)
  - Virtual Bolus
- Robust optimization solves this directly

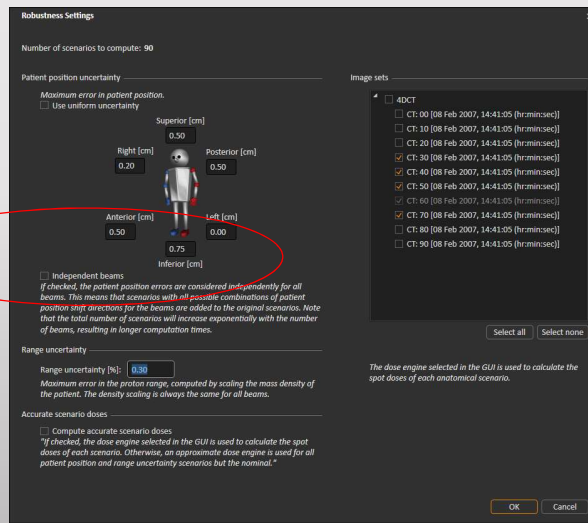


## BREAST: Skin flash



## Optimization of independent beams

- Individual beam dose

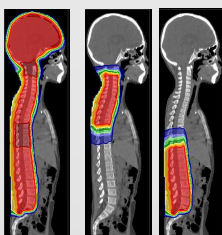


## Craniospinal: INDEPENDENT BEAMS

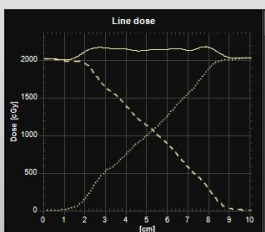
- What if beams shift relative to each other?
- Use beam dose ramps at field junctions
- Achieved by robust optimization with independent beams
- Ramp size can be controlled by a single structure for each junction
- Exponential growth of scenarios, but polynomially many beam doses



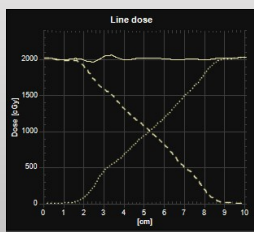
## Craniospinal: INDEPENDENT BEAMS



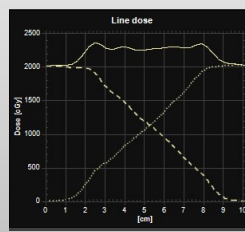
- Example: IMPT with independent beams for craniospinal
- Robust PTV constraint: min 20 Gy
- If nominal uniform dose is prioritized: robust PTV objective



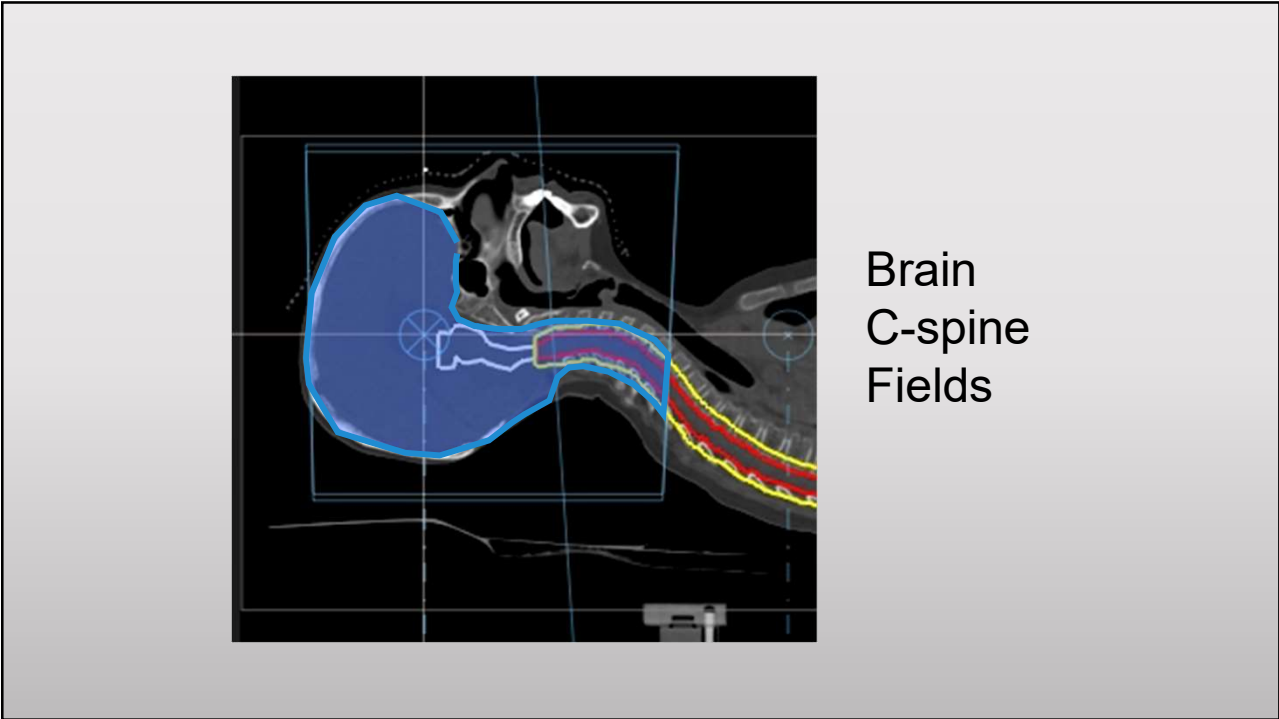
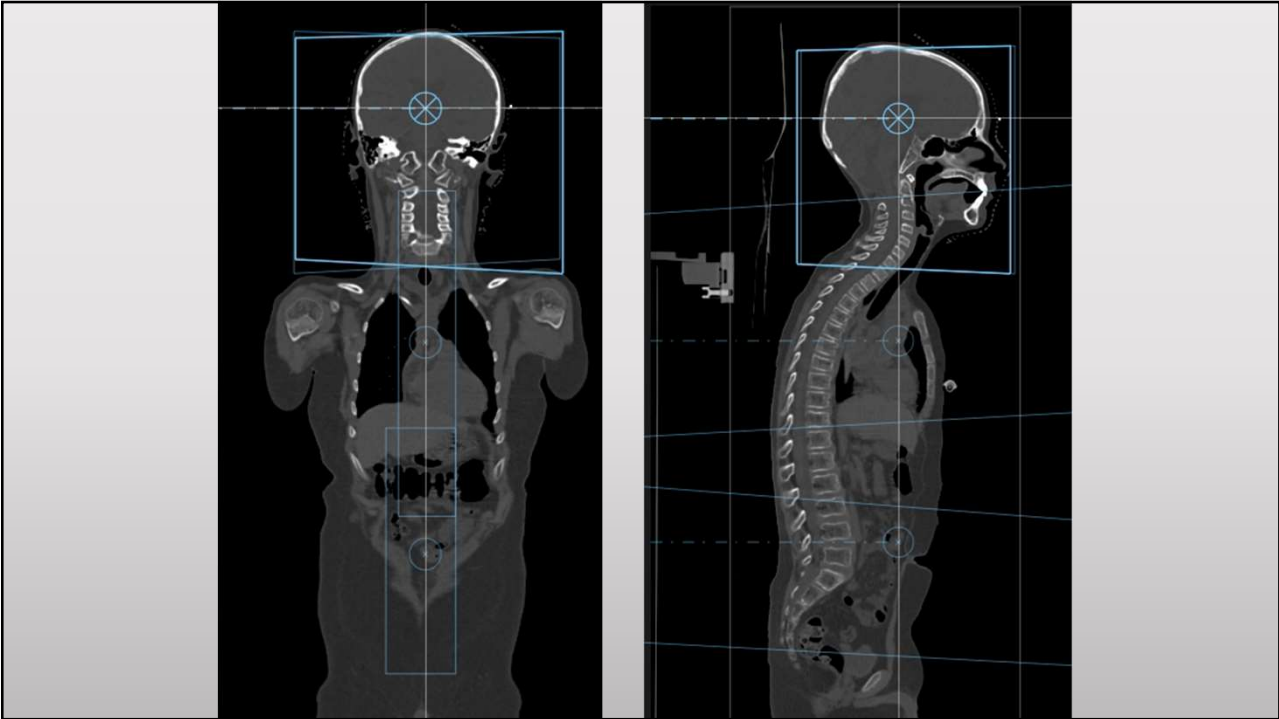
Planned dose

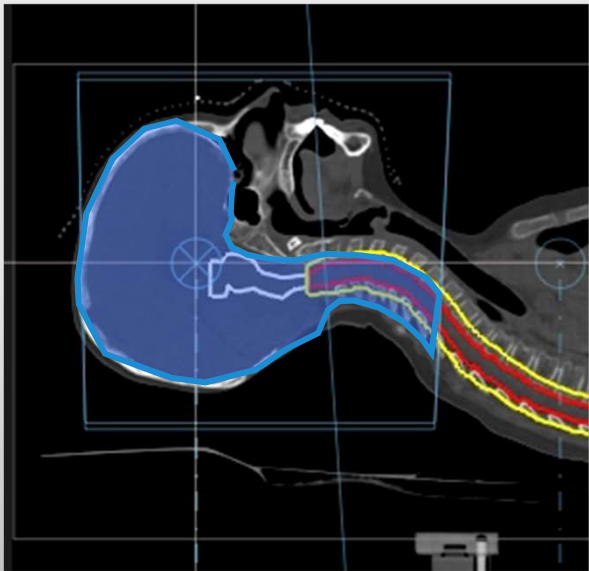


Beams moving apart

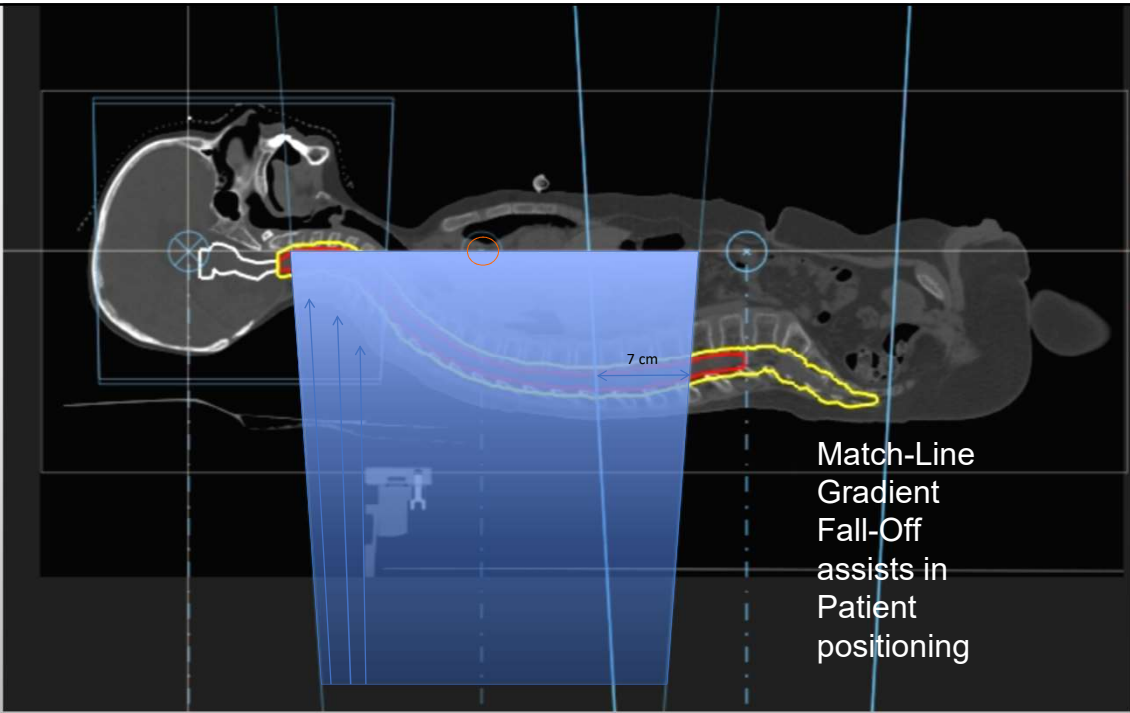


Beams moving together

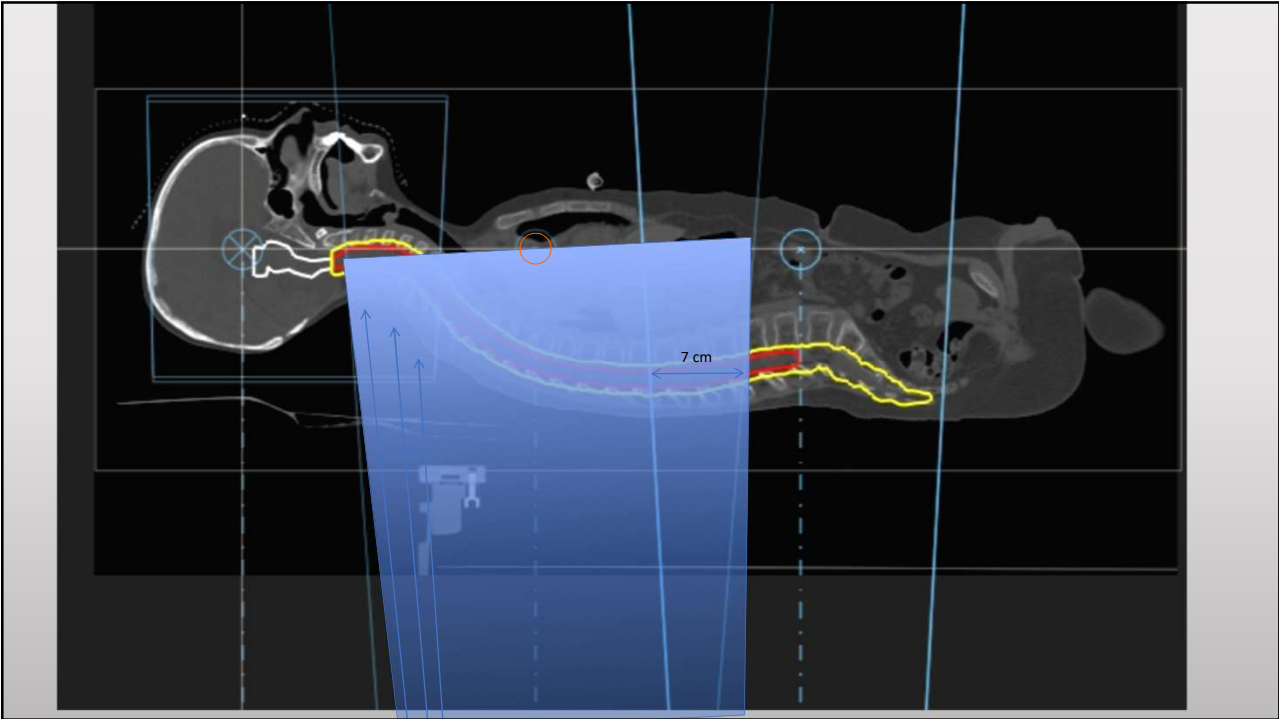
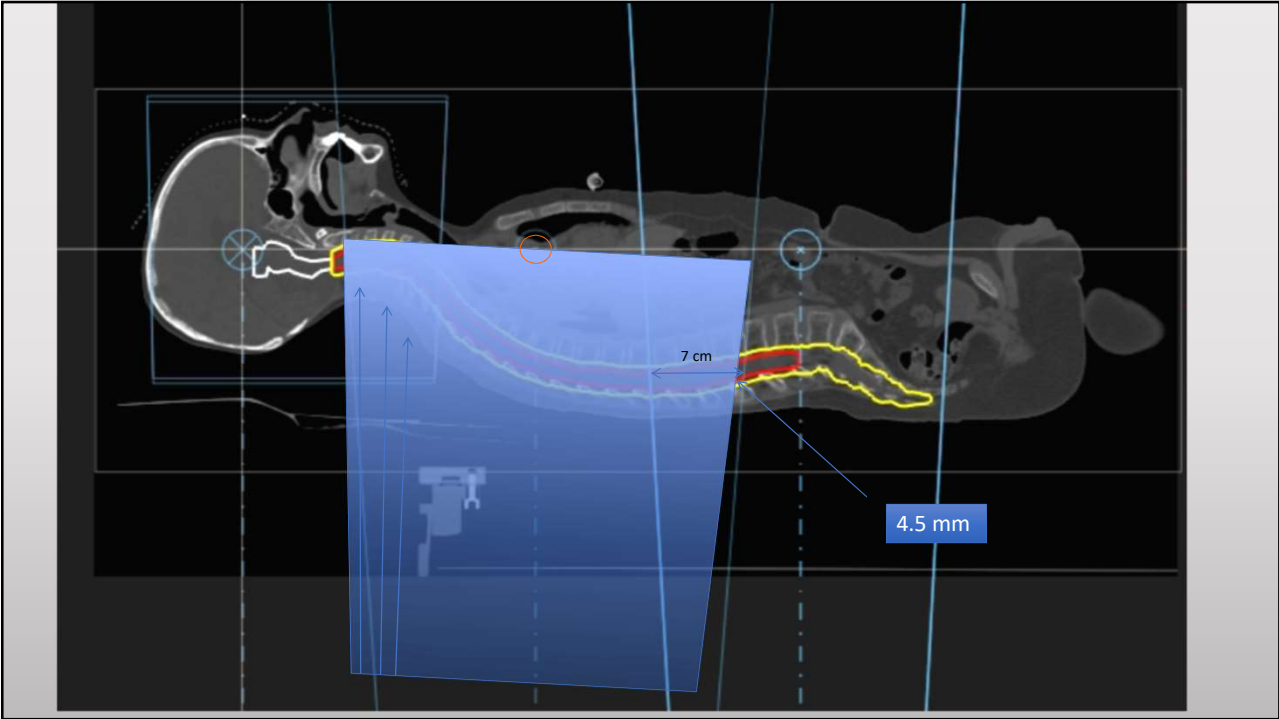




Planning  
to a  
Robust  
Pitch  
Set-up

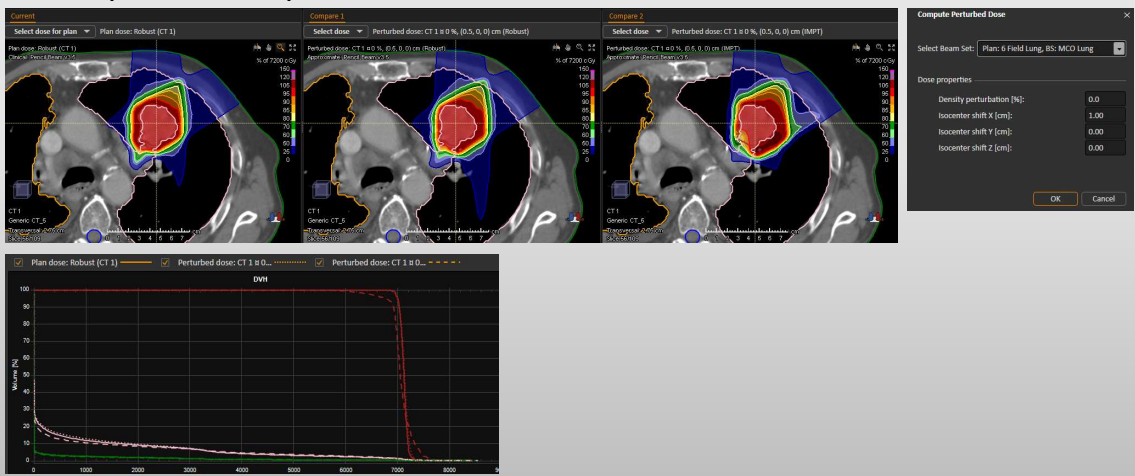


Match-Line  
Gradient  
Fall-Off  
assists in  
Patient  
positioning



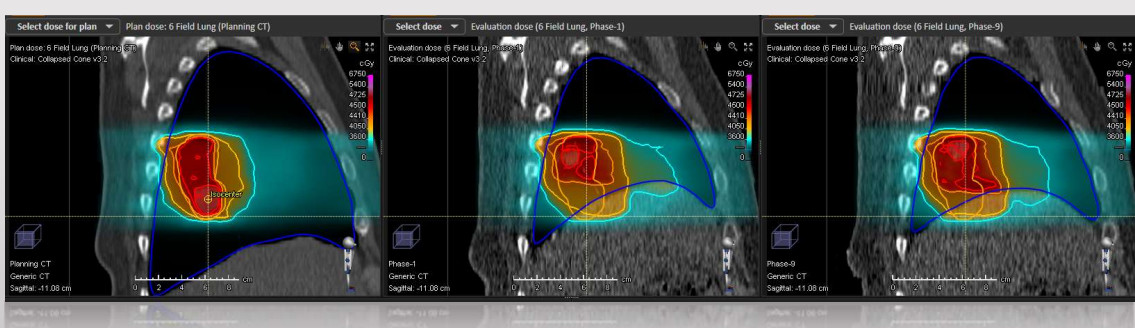
# Evaluation of robust plans

- Computation of perturbed dose



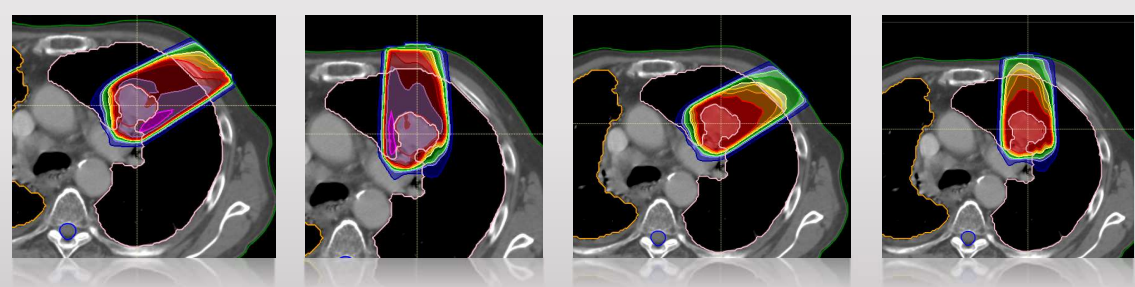
# Evaluation of robust plans

- Review dose on phases

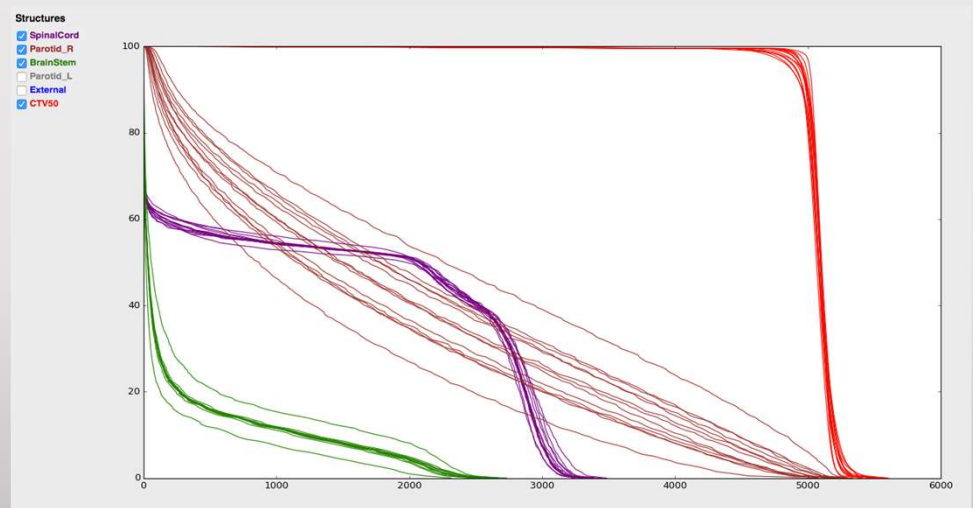


## Evaluation of robust plans

- Individual beam dose



## Pass Robust Evaluation





## What did we achieve?

- Robust Parameters
- Setup and range uncertainties
- Robust Optimization – min/max
- Planning to Pass Robustness Scenarios

## Mitigating Setup and Range Uncertainty

- Steep dose gradients cause hot and cold spots during perturbations
- Beam by beam uniform dose – Single Field Uniform Dose (SFUD)
  - Can still lead to hot and cold spots when perturbed because the patient is heterogeneous and the beams use intensity modulated fluence maps (even with SFUD)

## Robust observations

- Robust planning- with many scenarios can increase computation time.
- Plans look different
  - Cover multiple scenarios and data sets as opposed to just the one snapshot in time
- Proton planners understand the need for Robustness, using it in most cases!
- Photon planners are cautious as to how it will change or improve their planning process
  - Perhaps better understand the results?



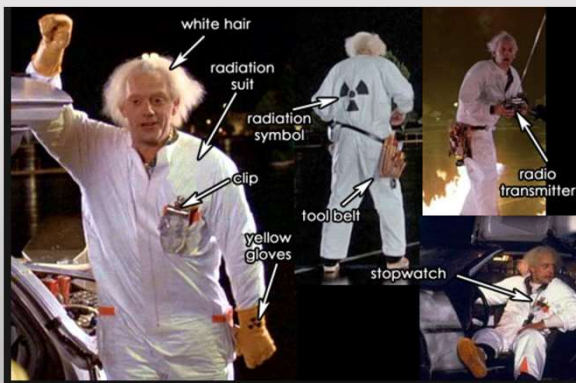
## Future of planning with robust optimization

- **Can we eliminate ICRU 50 definitions one day?**
  - Perhaps only the PTV, CTV must stay
- **Automation**
  - Scripting and protocols exist today but will robustness be inherent in all future plans?
- **Many more variables being accounted for in the optimization?**
  - More organ motion or prediction
  - May reduce the need for adaptive therapy?



## Future of planning with robust optimization

- **Future evaluation**
  - Simulated treatments
  - Generation of probability in fulfilling clinical goals
- **Knowledge building for future cases**
  - How many scenarios are needed?
  - Weight loss prediction?



## Conclusions

- Using planning margins is a heuristic solution that doesn't always work
- Explicitly incorporating the actual goal into the optimization is necessary for protons
- Robustness also solves other problems, such as field matching and flash in photon planning
- Many more scenarios can be included going forward, making Robustness applicable in other planning cases



