# Computational challenges in RTP treatment planning: setting the stage

Joe Deasy, PhD deasyj@mskcc.org

#### Overview

- Treatment planning: underpinnings
  - Outcomes and models
- Treatment planning technology
  - The desirables: what we need
  - How can we get there?

# Treatment planning: underpinnings Outcomes and models

The 'Planning Target Volume': add margins around suspected disease for uncertainty with respect to anatomy + geometric motion + geometrical uncertainty + possible gross disease.

Give this the full dose right out to the edge.

Also stands for 'Poor Tradeoff Volume'

#### Problems with the PTV concept

- Large shifts (very unlikely) are treated the same as small shifts (very likely)
- assumes that normal tissue volume effects are much less important than target volume coverage
- The Gaussian dose tail will always invade the PTV; so full dose requires a big field

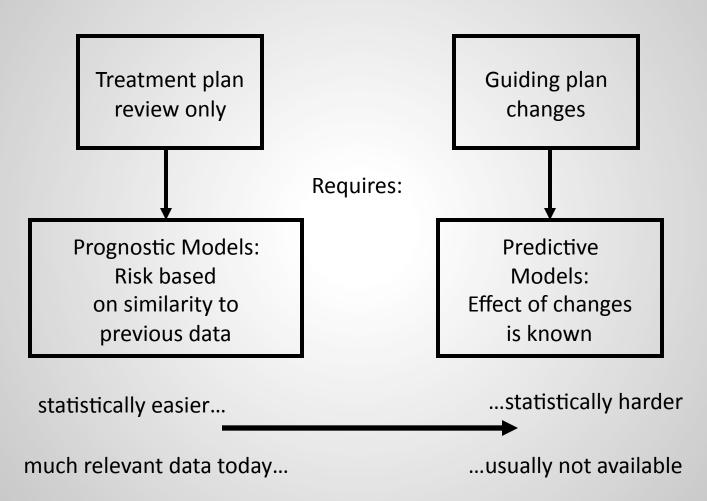
#### Instead...

- Treatment planning needs to be driven by the impact on outcomes.
- And different treatment philosophies will handle tradeoffs between local control and toxicity differently.

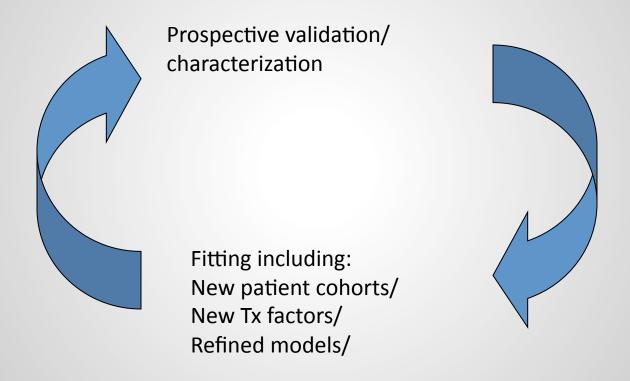
#### Planning objective functions

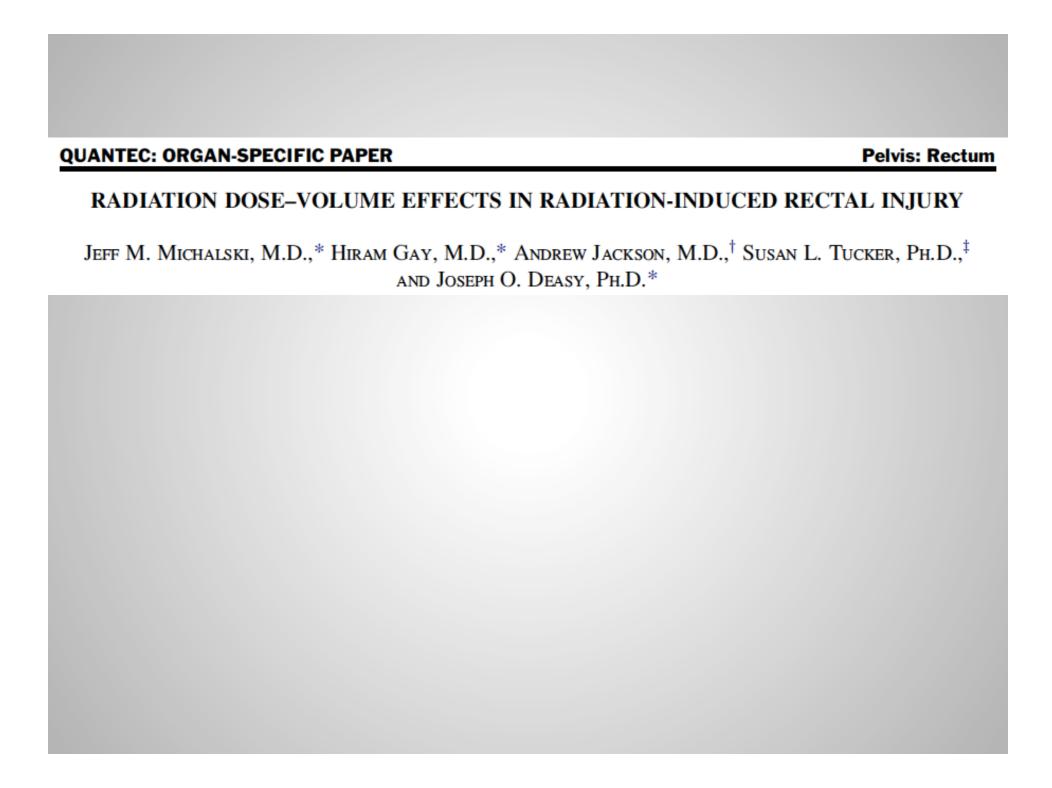
- Only just emerging...
- Usually numerically uncertain
- Often have shallow slopes instead of sharp separations between risk and non-risk

#### NTCP models used for:

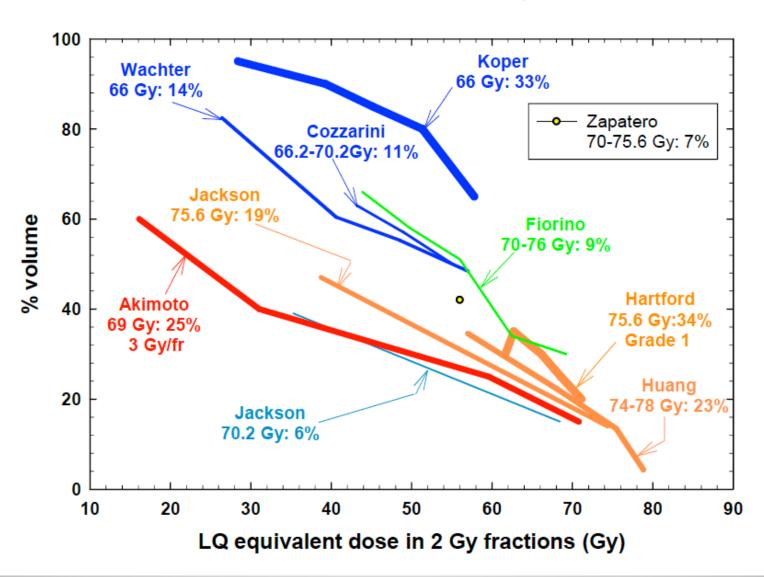


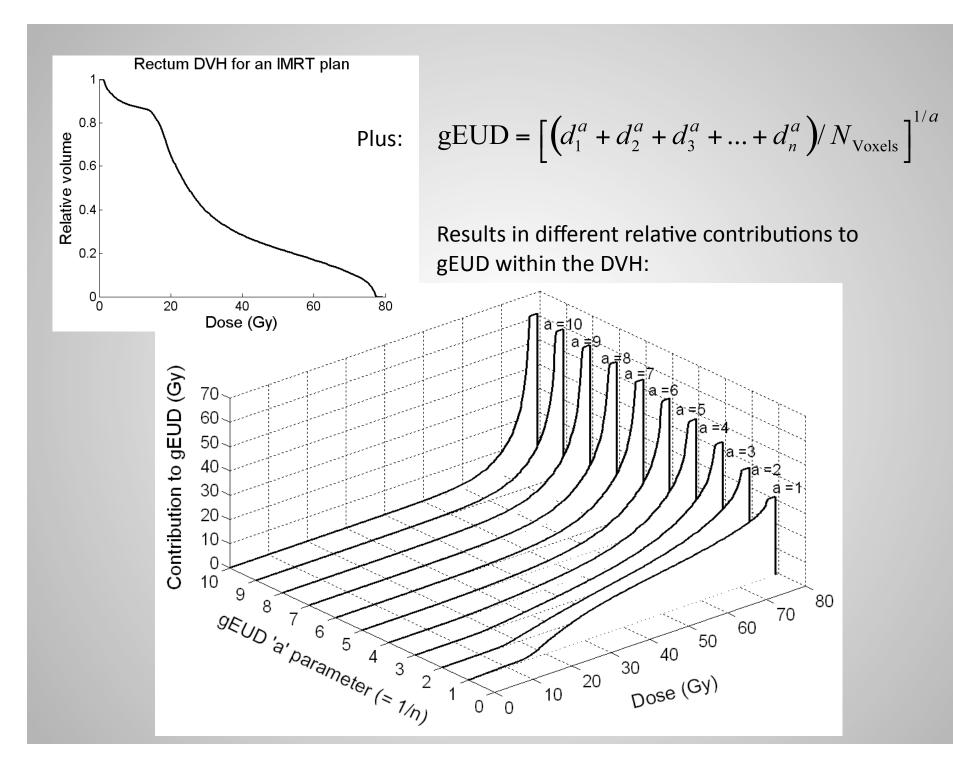
#### The NTCP/TCP development spiral

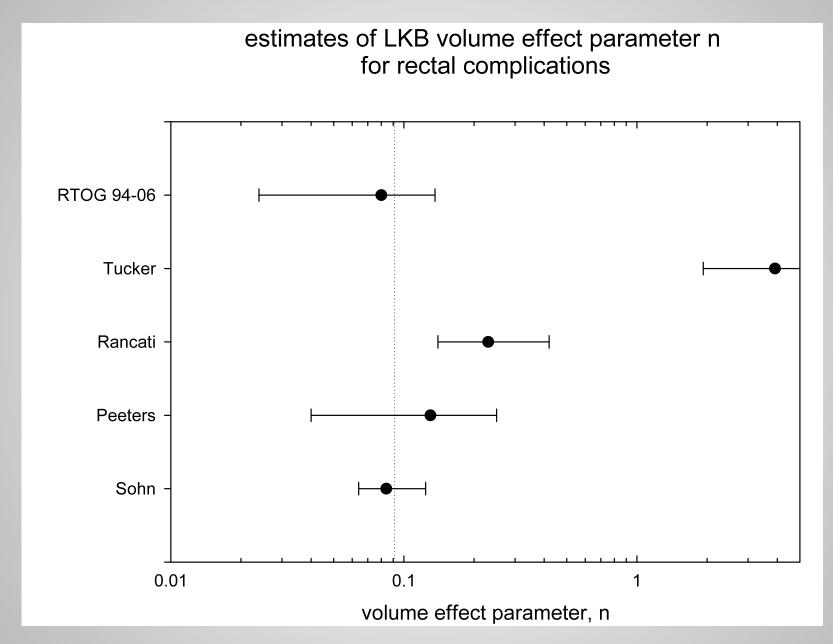




#### Dose-volume limits for >= grade 2 rectal toxicity with LQ corrected doses ( $\alpha/\beta$ = 3 Gy)







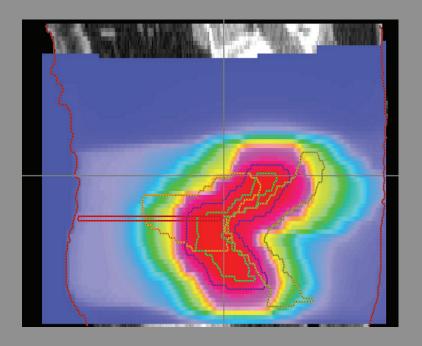
(Slide courtesy A. Jackson et al.)

(Note: a = 1/n)

# Are IMRT rectal dose-volume limits a problem?

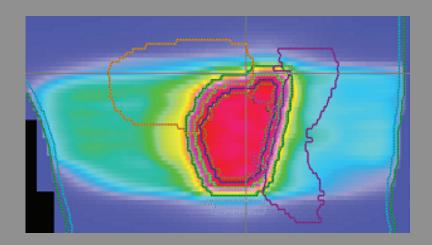
(a= 10)

**Typical 3DCRT plan (sagittal)** 



Rectal gEUD = 53 Gy

Typical IMRT plan (sagittal)



Rectal gEUD = 24 Gy

Radiotherapy and Oncology 74 (2005) 187-195

www.elsevier.com/locate/radonline

#### Rectal and bladder motion during conformal radiotherapy after radical prostatectomy

Claudio Fiorino<sup>a,\*</sup>, Franca Foppiano<sup>b</sup>, Paola Franzone<sup>c</sup>, Sara Broggi<sup>a</sup>, Pietro Castellone<sup>a,d</sup>, Michela Marcenaro<sup>c</sup>, Riccardo Calandrino<sup>a</sup>, Giuseppe Sanguineti<sup>e</sup>

\*Servizio di Fisica Sanitaria, IRCCS H. S. Raffaele, Via Olgettina 60, 20132 Milano, Italy bUnità di Fisica Medica, Istituto Nazionale per la Ricerca sul Cancro, Genova, Italy cDipartimento di Radioterapia, Istituto Nazionale per la Ricerca sul Cancro, Genova, Italy dDipartimento di Fisica, Università Federico II, Napoli, Italy cDepartment of Radiation Oncology, University of Texas, Medical Branch Galveston, TX, USA

Received 17 December 2003; received in revised form 7 September 2004; accepted 7 October 2004 Available online 28 October 2004

(Rad Oncol 2005)

#### Variation in rectal DVHs for daily Fx's

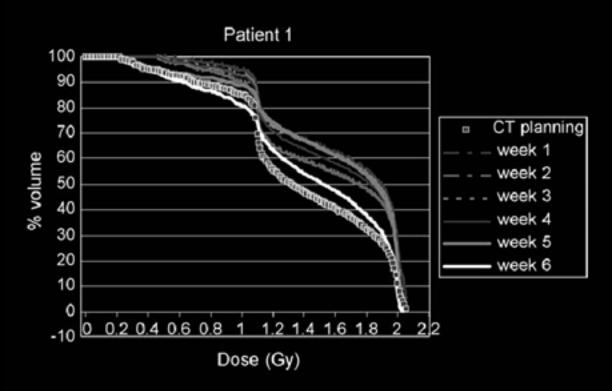


Fig. 3. Rectum DVH trend: the DVH at planning scan (CT planning) and the DVHs during the treatment (week 1...week 6) are shown for a patient with a large systematic component of the difference between planning and therapy (Patient 1).

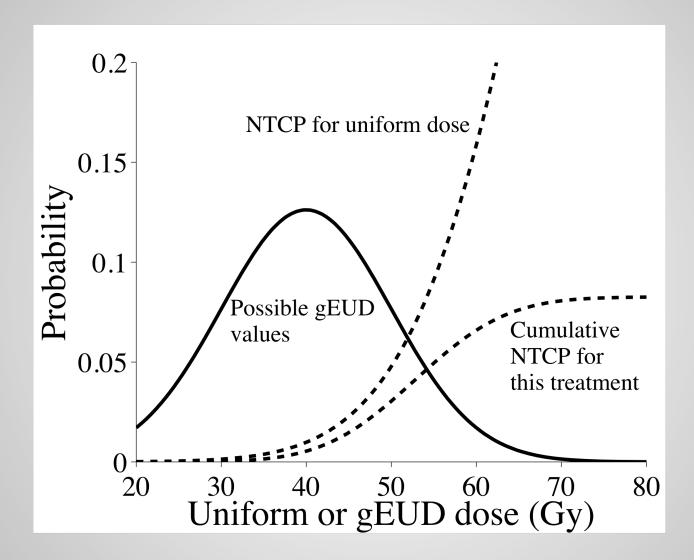


Fig. 9

Acta Oncologica, 2010; 49: 1040-1044

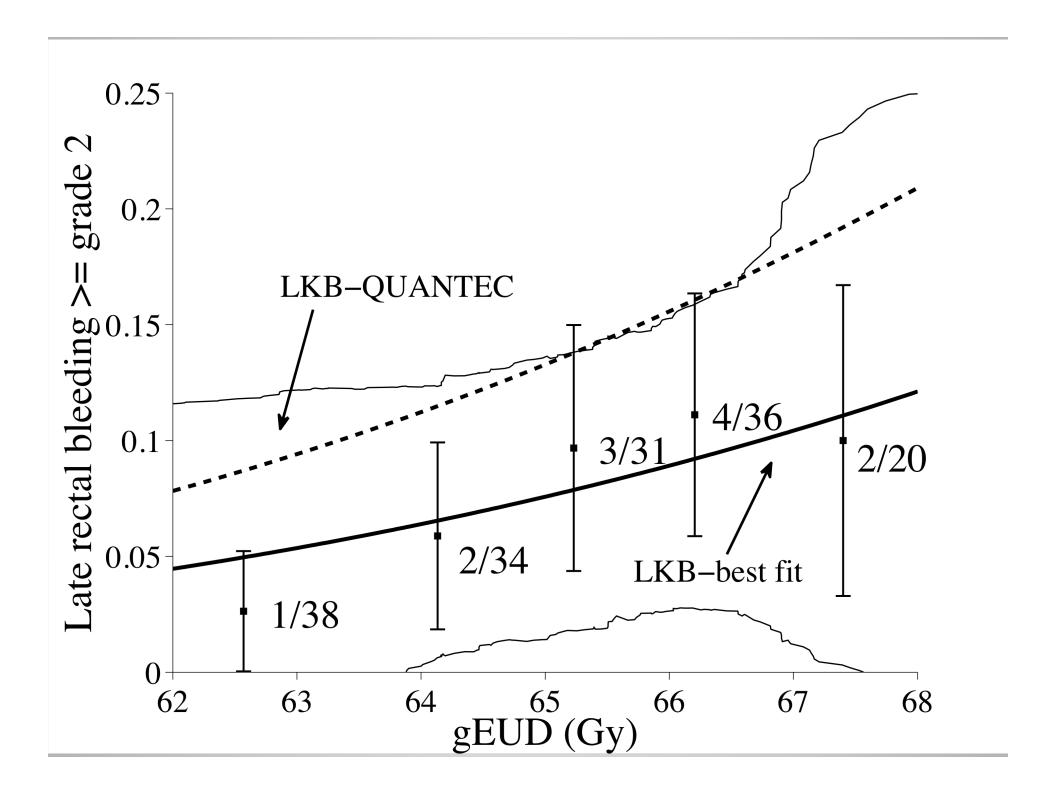


#### ORIGINAL ARTICLE

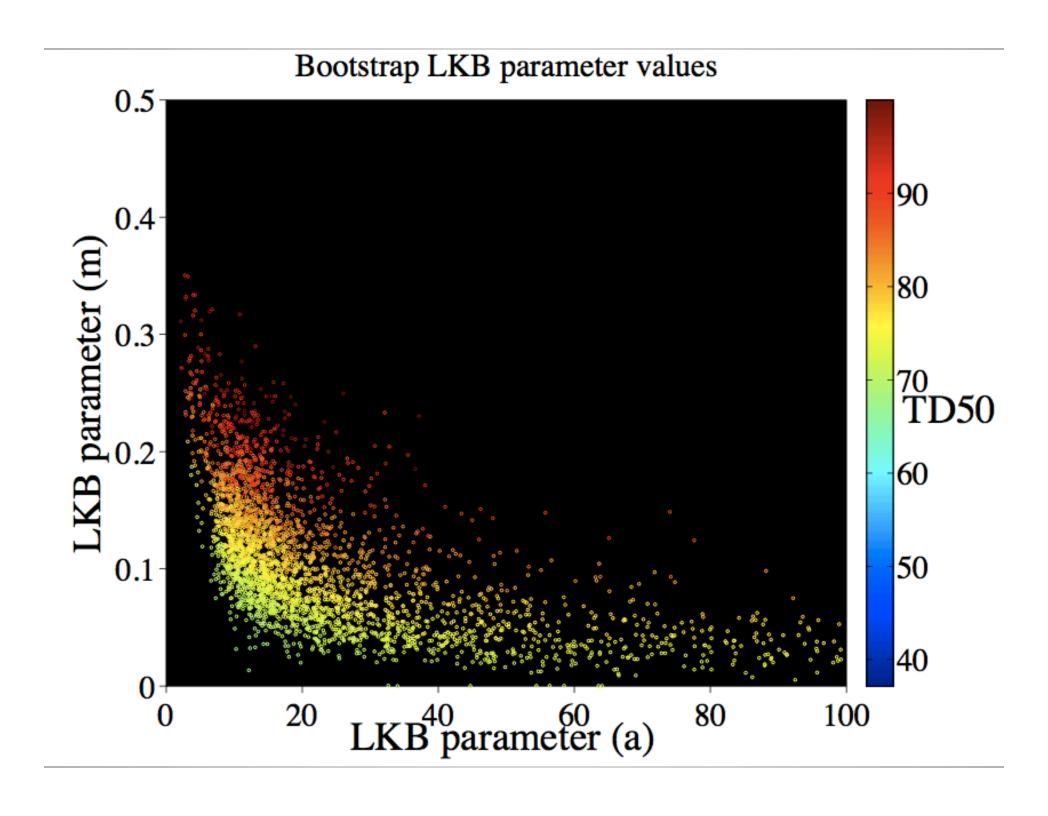
Normal Tissue Complication Probability (NTCP) modeling of late rectal bleeding following external beam radiotherapy for prostate cancer: A Test of the QUANTEC-recommended NTCP model

MITCHELL LIU<sup>1</sup>, VITALI MOISEENKO<sup>2</sup>, ALEXANDER AGRANOVICH<sup>1</sup>, ANAND KARVAT<sup>1</sup>, WINKLE KWAN<sup>1</sup>, ZIAD H. SALEH<sup>3</sup>, ADITYA A. APTE<sup>3</sup> & JOSEPH O. DEASY<sup>3</sup>

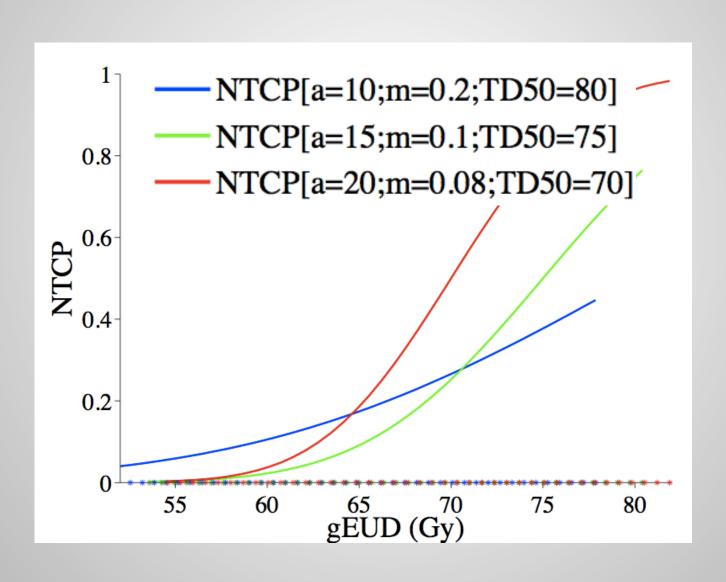
<sup>1</sup>Fraser Valley Centre, British Columbia Cancer Centre, Surrey, BC, Canada, <sup>2</sup>Vancouver Centre, British Columbia Cancer Agency, Vancouver, BC, Canada and <sup>3</sup>Department of Radiation Oncology and the Mallinckrodt Institute of Radiology, Washington University, St. Louis, MO

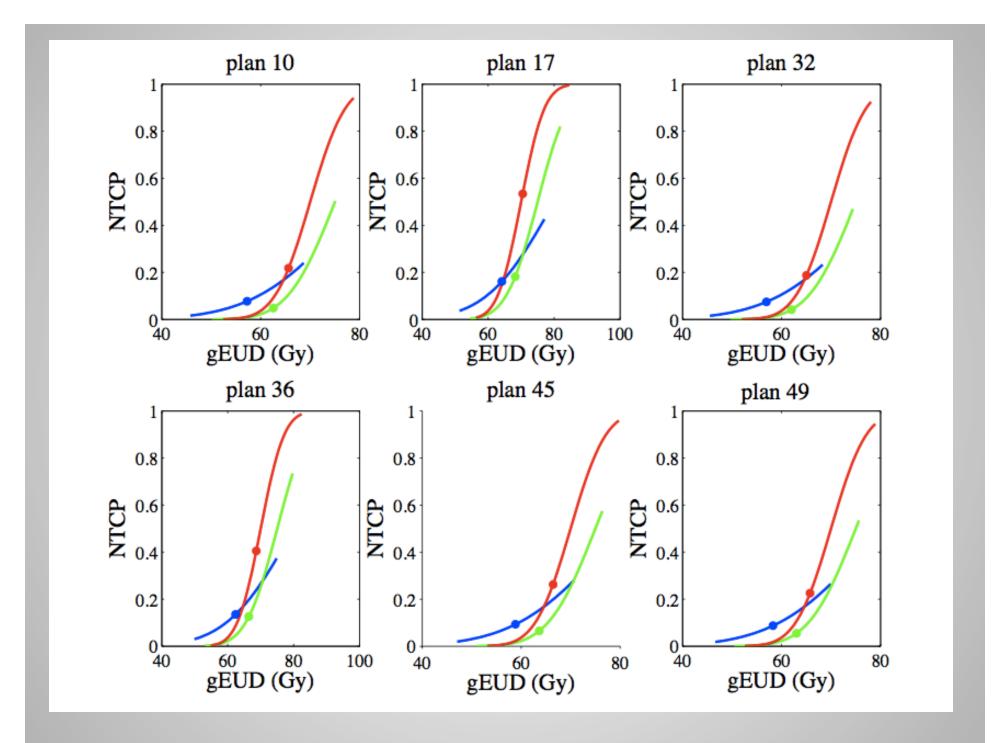


# The impact of model parameter uncertainties on NTCP predictions



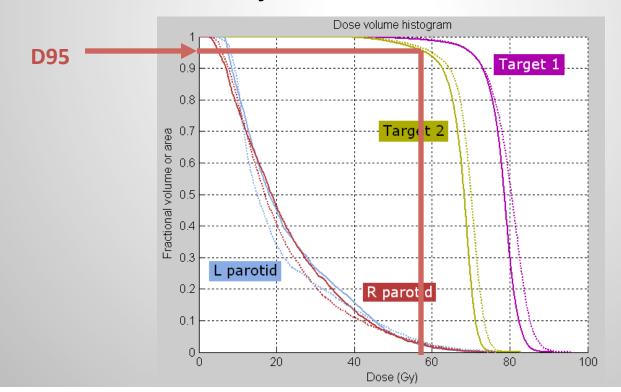
#### What is the true model?



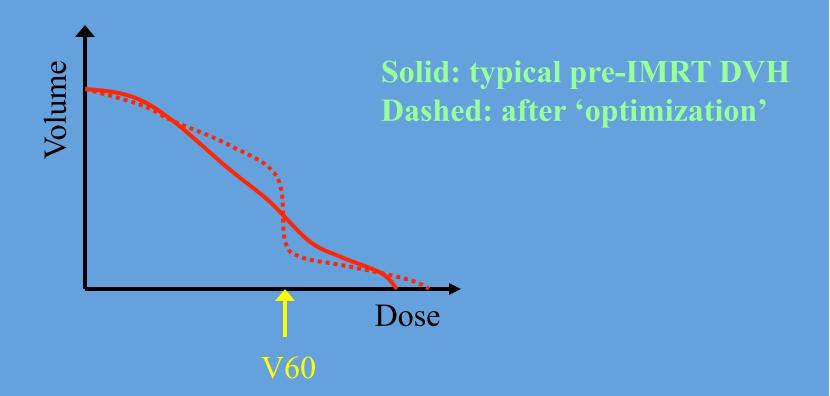


#### Dose-volume constraints

- Dx represents the minimum dose that at least x% of the volume receives
- Vx represents the minimum (percentage) volume that receives at least x Gy



#### Illustration of DVH 'warping' by a dosevolume constraint



Big problem when single DVH pts used for optimization: pinned DVH

# Additional problems with using traditional dose-volume constraints in IMRT optimization

- Non-convex
  - No guarantee of optimality with current solvers
  - Potentially slower than linear/quadratic objectives
- Difficult implementation. Must be:
  - Mixed-integer programming, or
  - Approximated

#### Solution: mean-tail-dose

- Definition
  - MOHx mean of hottest x%
  - MOCx mean of coldest x%
- Potentially more biologically relevant (smoother)
- Linear
- Convex
  - Faster
  - Guaranteed to be optimal if solution is found
- Found to correlate with traditional dose-volume metrics
- Suggested by Romeijn et al. (2003, 2006)

# Solution: generalized equivalent uniform dose (gEUD)

gEUD is the exponentially weighted dose value:

$$gEUD(\mathbf{d}, a) = \left(\frac{1}{N} \sum_{i=1}^{N} d_i^a\right)^{1/a}$$

- $-d_i$  represents the dose to voxel i,
- N is the number of voxels in the structure of interest, and
- − *a* is a variable exponent parameter.

#### Benefits of gEUD

- Radiobiologically relevant
  - Has parameter (a) which can be tuned to underlying data
  - Can track high doses (a large), low doses (a negative), or some average dose ( $a \sim 1$ )
- Convex (Choi 2002)
- Correlates well to dose-volume constraints
- Previously suggested (Niemierko 1998; Choi 2002; Wu 2002, 2003; Olafsson 2005; Thomas 2005; Chapet 2005)

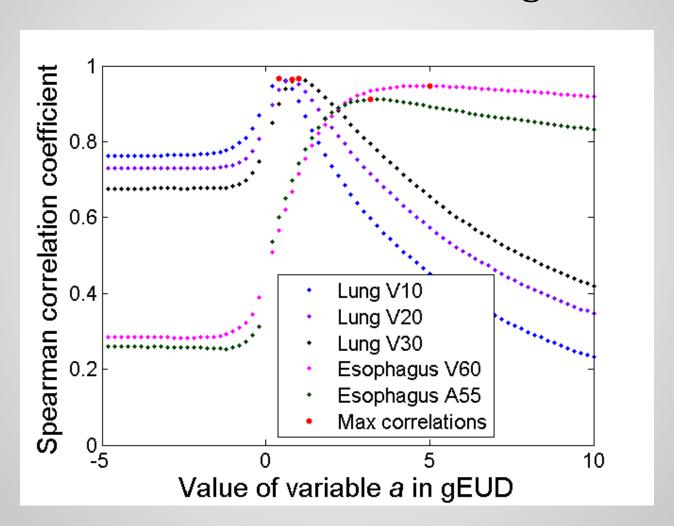
# Dissertation work: Determining value of replacement candidates

- Datasets of clinical treatment plans were used:
  - Lung cancer, 263 patients
  - Prostate cancer, 291 patients
- Calculated the correlation between gEUD and clinically applicable dose-volume constraints
  - Tested various values of the parameter a
  - Did the same for MOH/MOCx with values of x

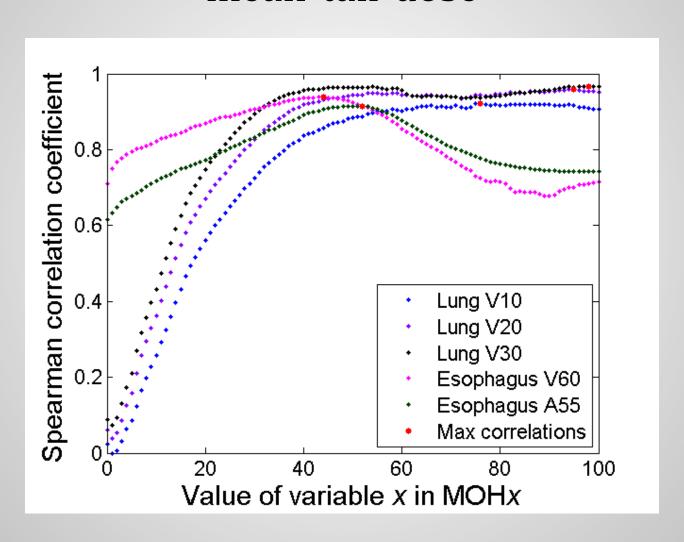
# Correlations between gEUD/MOH/MOC and DV metrics

- Spearman correlation coefficients
  - All together mean: 0.94, median: 0.97, range: 0.45 to 1.00
  - gEUD mean: 0.93, median: 0.96, range: 0.45 to 1.00
  - mean-tail-dose mean: 0.95, median: 0.98, range: 0.45 to 1.00
  - The 0.45 coefficient is an outlier
- 51 of the 60 correlations tested had a Spearman correlation coefficient greater than 0.90
- In general, mean-tail-dose had higher correlations than gEUD (27 of 30 correlations equal or greater), though gEUD and mean-tail-dose correlation coefficients were often similar.

# Correlation between dose-volume metrics and gEUD



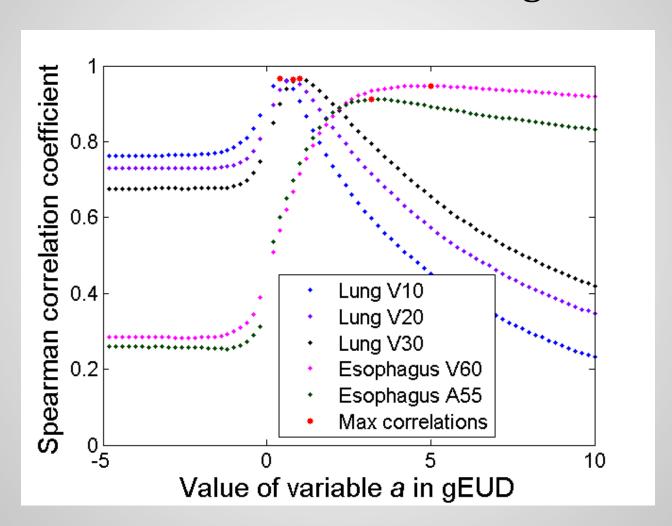
### Correlation between dose-volume metrics and mean-tail-dose



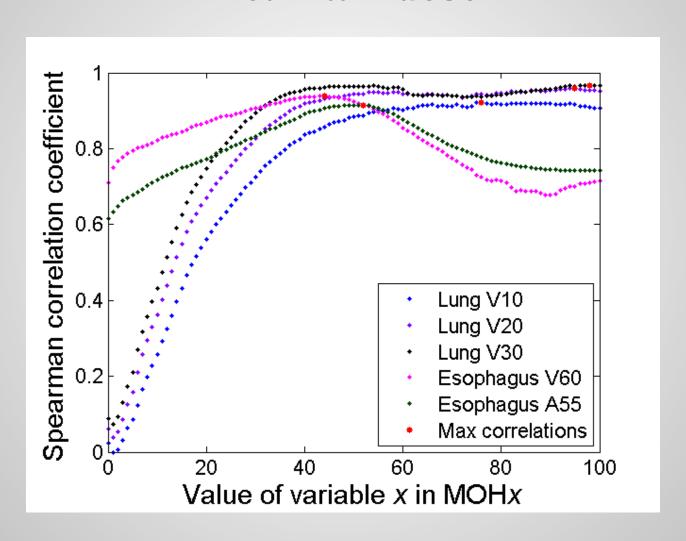
# Correlations between gEUD/MOH/MOC and DV metrics

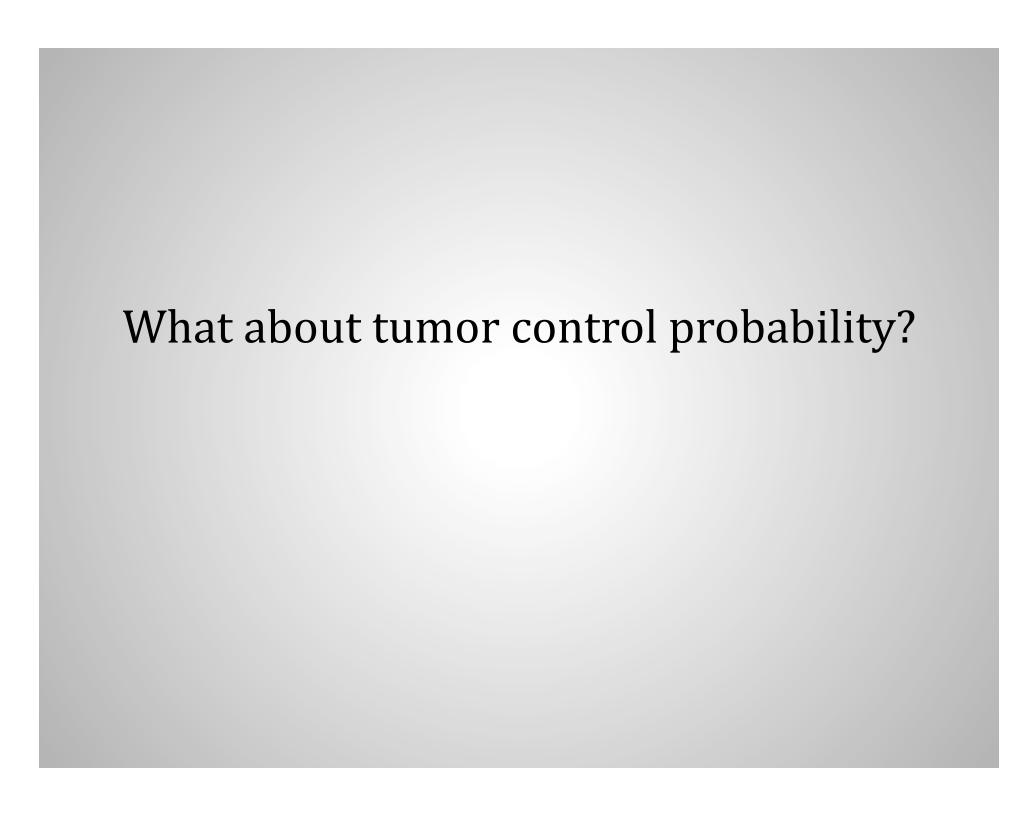
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# Correlation between dose-volume metrics and gEUD



### Correlation between dose-volume metrics and mean-tail-dose





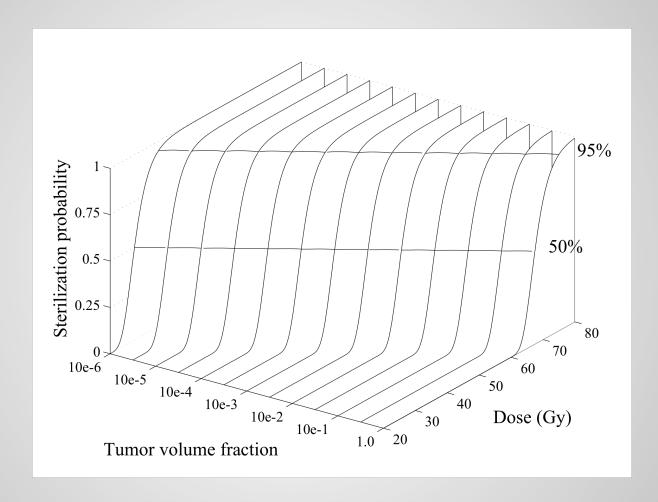


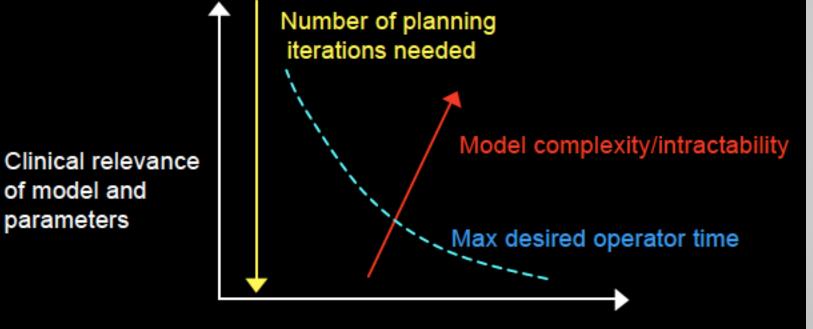
Fig. 1

How can we get there? The desirables: what we need

### IMRT software design goals

- Clinical relevance of parameters and plan model
- Steerability of plan changes in next plan iteration or at "tweaking phase"
- Efficiency/speed of feasible solution generation in a plan iteration
- Dosimetrically faithful and delivery-time-efficient leaf segmentation
- Accuracy, likelihood of arriving at the global optimimum to within a target %.

#### IMRT algorithm tradeoffs



Model iteration compute time

# How can we get there? Our approach

(many following slides from Vanessa Clark's dissertation with supervisor Yixin Chen)

#### Current paradigm

Input algorithm parameters:

Hard constraints, objective function weights.

Solve an optimization problem

Review: is this the best plan possible? & is it clinically acceptable? If "no" to either, change input and re-run.

# Prioritized prescription optimization

Input prioritized prescription planning objectives

Solve a series of optimization problems which add the next lower priority goal at each iteration. Higher priority goals are "constraints" in lower priority iterations.

Review: was the prescription statement appropriate? & is the resulting plan clinically acceptable? If "no" to either, change input and re-run.

# Previous work using hierarchical methods for IMRT optimization

- Jee et al. (Michigan 2007)
  - Lexicographic ordering
  - Goal programming, not slip
  - Nonconvex dose metrics
  - 2 cases (1 prostate, 1 head and neck)
  - No direct clinical comparison
- Spalding et al. (Michigan 2007)
  - Lexicographic ordering, gEUD, IMRT
  - 15 pancreatic cancer cases
- Wilkens et al. (WashU 2007)
  - 6 head and neck cases
  - No direct clinical comparison

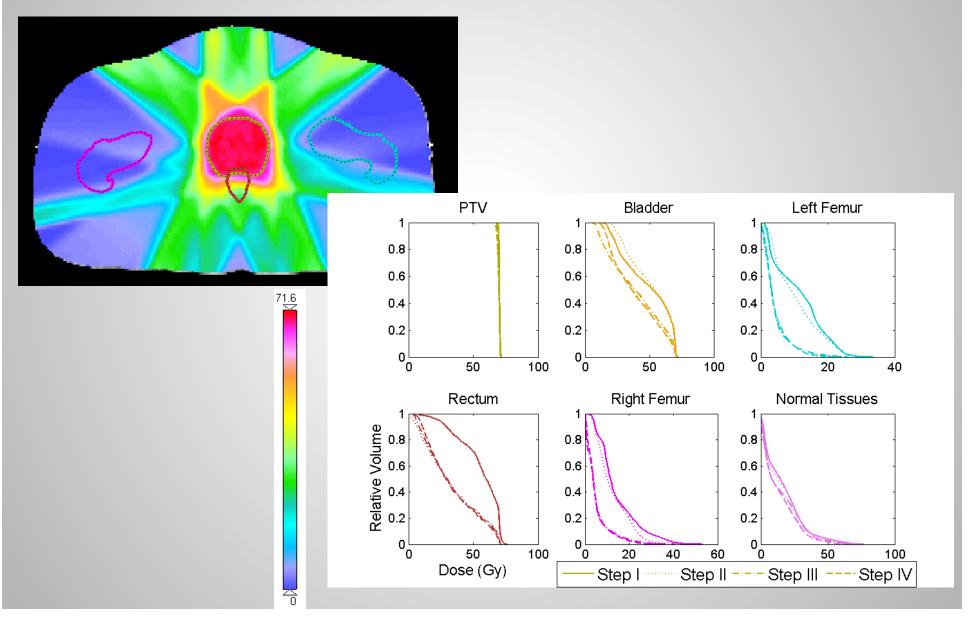
#### This work

- Hierarchical optimization
- 22 total prostate cases
- Convex objectives and constraints (for optimality)
- Direct clinical comparison (12 cases)
- First successful implementation of optimization for prostate using lexicographic ordering that is completely quadratic and linear (and convex)

Prioritized prescription optimization

	objectives	hard constraints
step I target coverage, high-priority OARs	minimize $F_l = \sum_{all \text{ voxels } j} (D_j - D_{pres})^2$ and maximize $D_{min}$ for <b>PTV</b>	<ul> <li>D<sub>max</sub> for <b>rectum</b></li> <li>W<sub>max</sub> for beam weights</li> </ul>
step II additional OARs	minimize MOHx for <b>rectum</b>	<ul> <li>as in step I and</li> <li>max value for F<sub>I</sub> for all targets</li> <li>D<sub>min</sub> and D<sub>max</sub> for target</li> <li>as achieved in step I</li> </ul>
step III dose falloff	minimize D <sub>mean</sub> in bladder, femurs, and normal tissue	as in step II and  ■ MOHx for rectum  as achieved in step II
step IV smooth dose	minimize $\Sigma_{ m all\ beamlets\ i}$ ${ m W_i}^2$	<ul> <li>as in step III and</li> <li>D<sub>mean</sub> for bladder, femurs,</li> <li>and normal tissue as achieved in step III</li> </ul>

### Prostate results



### Results summary

- Results appear to be comparable to clinical quality
- Average total runtime: 20 mins (prostate)



# Chapter 4: Comparison of prioritized prescription optimization with current clinical results

Showed that prioritized prescription optimization is comparable to clinical methods by direct comparison

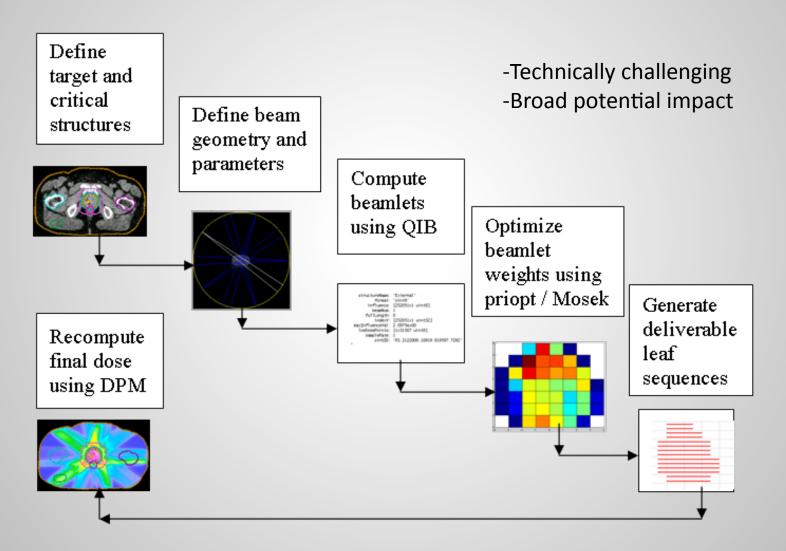
### Background: Clinical comparisons

- To translate this research to use in the clinic, comparisons with existing clinical methods are necessary
  - To show validity
  - To show usefulness
  - To be accepted by clinicians
  - To have commercial companies implement it in their clinical software
- Technically challenging
- No study has previously compared using prioritized optimization techniques directly to those in the clinic

### Formulation

- Based on physician preferences
- Guarantees (constraints):
  - Maximum PTV dose  $\leq$  75.6 Gy \* 1.10
  - Rectum V65 < 17% (MOH31 < 65)</p>
  - Rectum V40 < 35% (MOH98 < 30)
  - Bladder V65 < 25% (MOH35 < 64)</li>
  - Bladder V40 < 50% (MOH100 < 40)
- Objectives (optimized in the following order):
  - 1. Maximize PTV D98 (MOC10)
  - 2. Minimize Rectum V65 (MOH31) & V40 (MOH98)
  - 3. Minimize Bladder V65 (MOH35) & V40 (MOH100)
  - 4. Minimize Normal Tissues mean & max dose
  - 5. Smoothing and minimize PTV mean dose

### System Integration



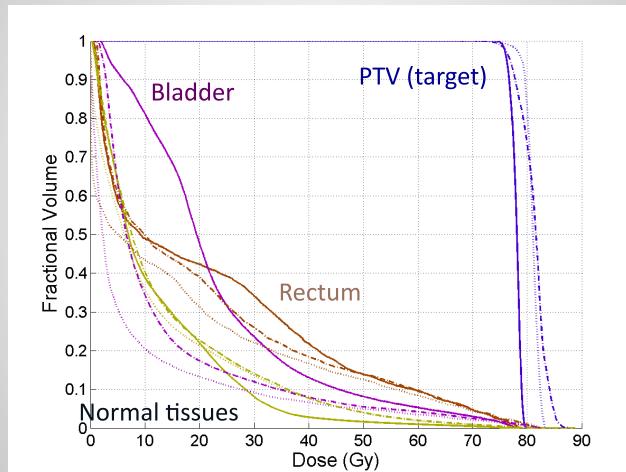
### Experiment datasets

- 6 definitive prostate (3 training)
  - 75.6 Gy
- 6 postoperative prostate (2 training)
  - -64.8 Gy
  - -79.2 Gy
- Anonymized
- Compared to results calculated using Pinnacle during normal course of clinical operations

\_\_\_\_ Clinical

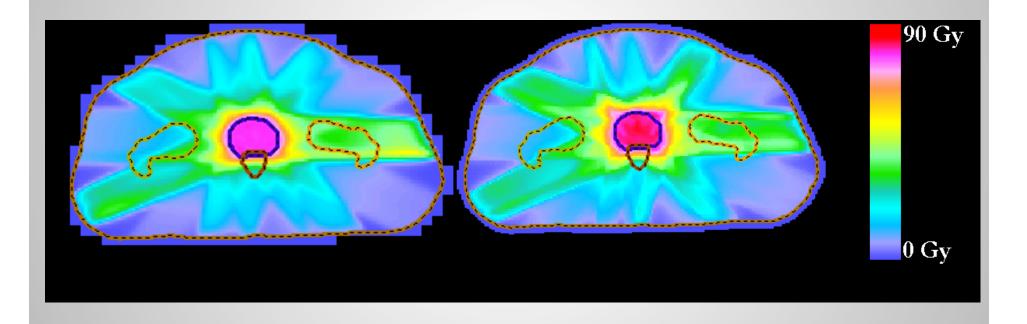
...... Priopt QIB

--- Priopt DPM



Example result

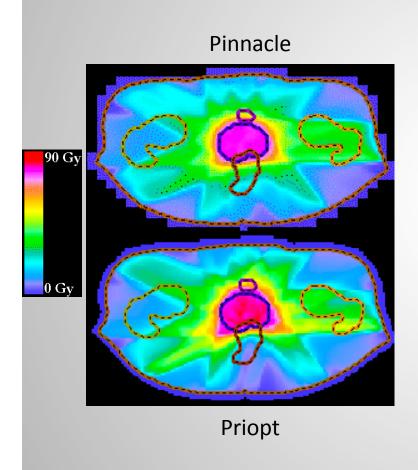
# Dose distributions for example result

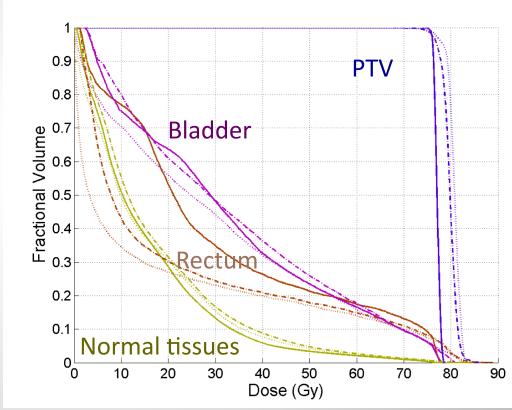


Clinical

Priopt with DPM

### Priopt vs. Pinnacle for one case



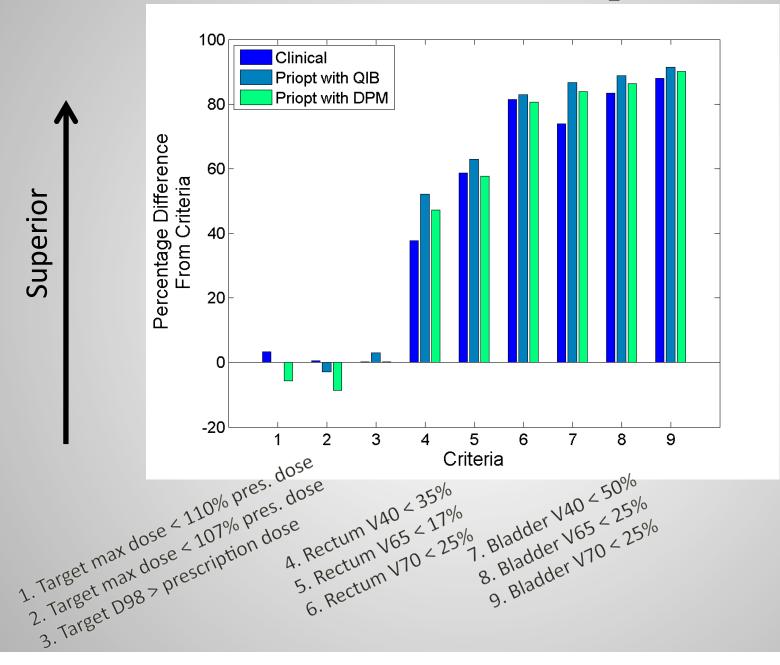


---\_Pinnacle

... Priopt QIB

- - Priopt DPM final

### Clinical criteria comparison



Superior

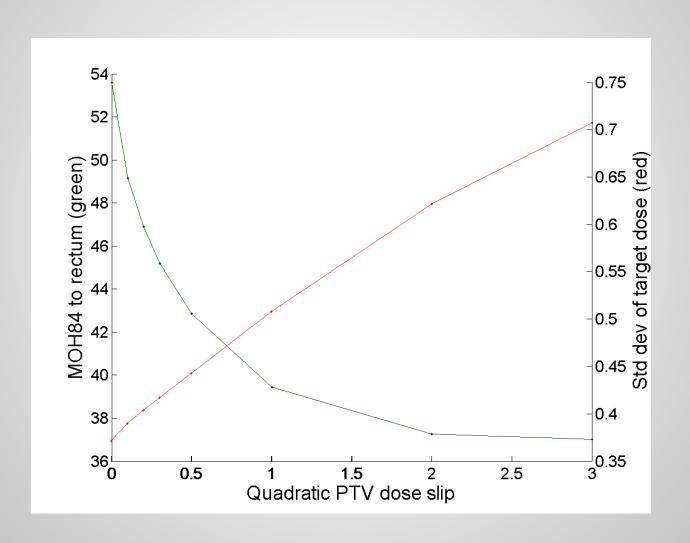
# Difference between our method and clinical for all cases

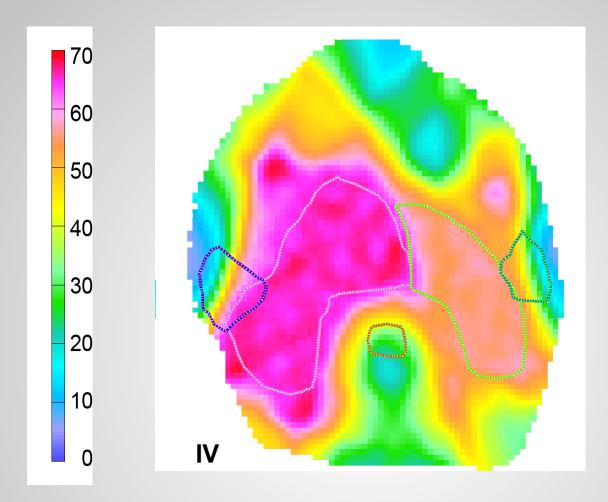
	Definitive Postoperative				
	Difference	Std Dev	Difference	Std Dev	
Target max dose < 110% pres dose	11.0	4.3	9.4	2.7	
Target max dose < 107% pres dose (preferred)	11.3	4.4	9.7	2.8	
Target D98 > prescription dose	0.0	0.0	0.0	0.0	
Rectum V40 < 35%	4.0	8.5	17.0	45.5	
Rectum V65 < 17%	1.7	9.0	0.0	20.0	
Rectum V70 < 25%	1.3	5.8	3.1	7.8	
Bladder V40 < 50%	3.8	6.5	1.8	25.9	
Bladder V65 < 25%	3.4	3.6	0.3	14.7	
Bladder V70 < 25% (unspecified criteria)	4.0	3.7	4.4	12.1	

### The concept of 'slip'

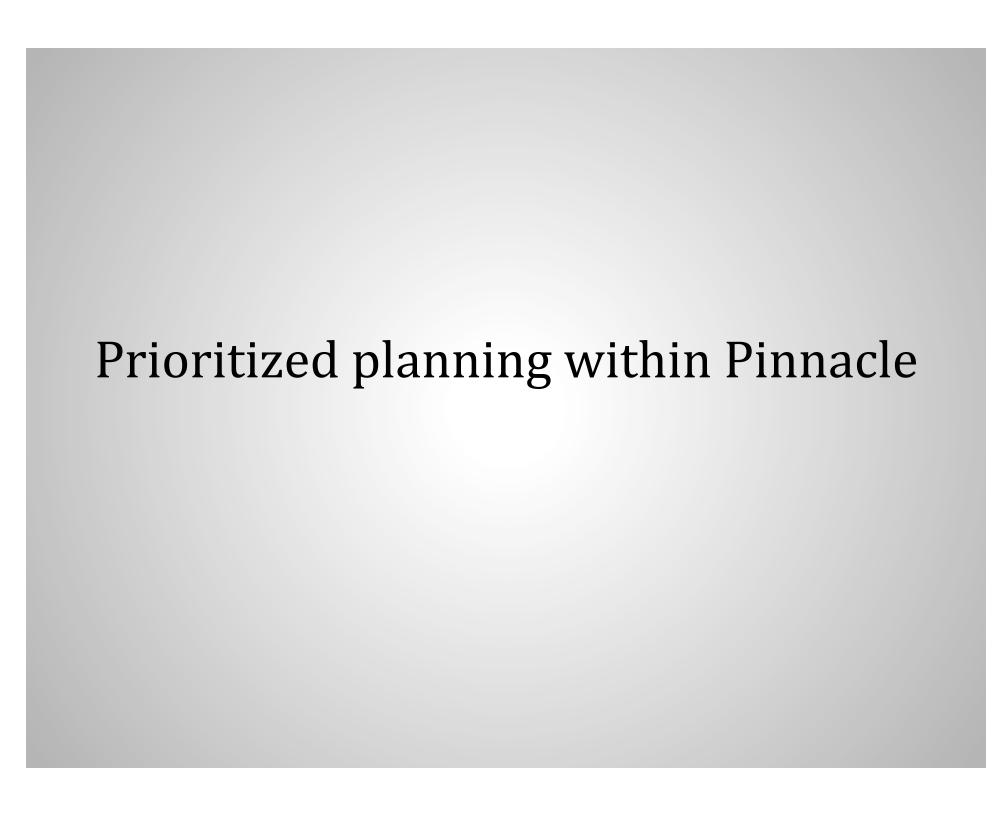
- Resulting dose distributions at each step may be quite similar (small search space)
- Need for expanding the possible solution space at each step
- Allow slight degradation of objective function from step to step: 'slip'
- Variable parameter what is best value?
- We found that a class solution works
- Automated solution may improve results

### Slip tradeoff graph



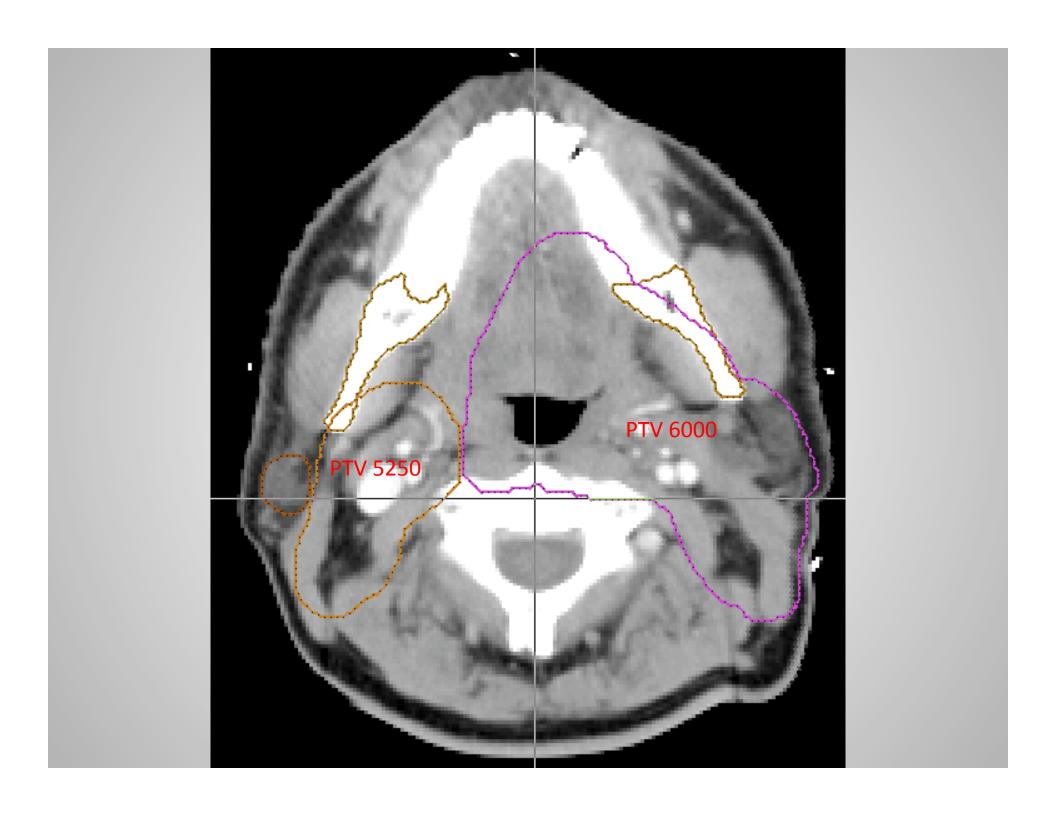


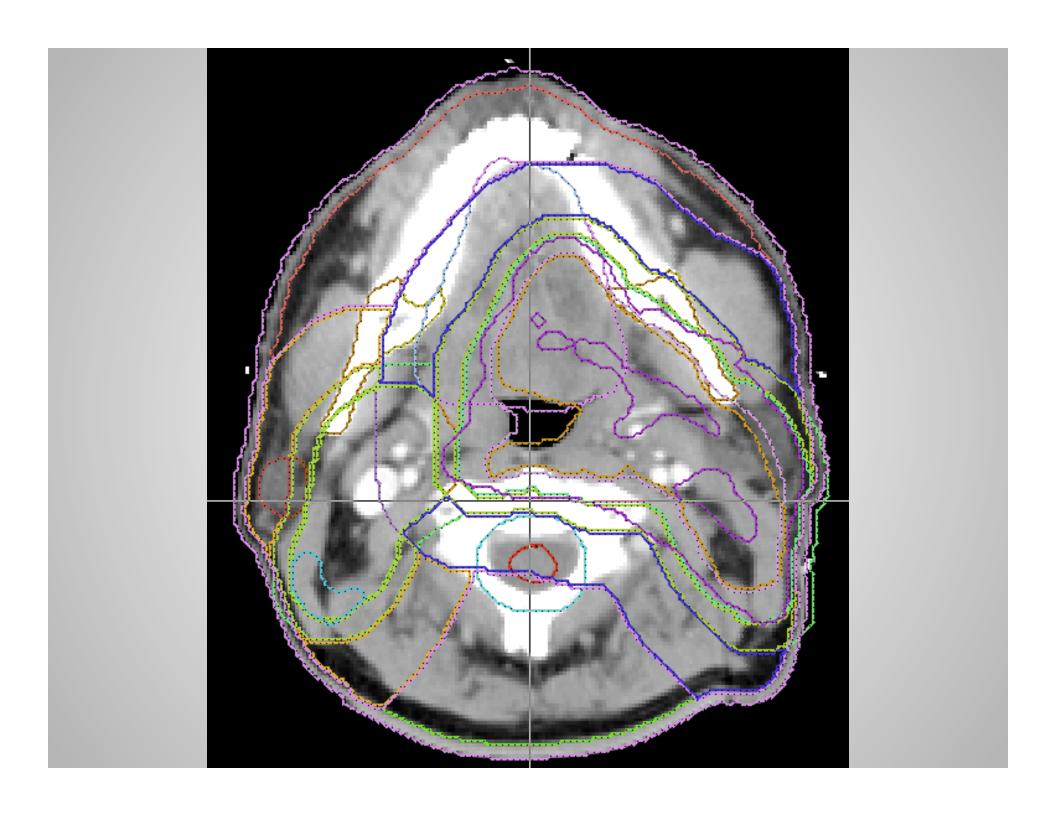
**Figure 4.** Transverse view of the dose distribution after step IV for a representative head and neck case. The outlined structures are (from left to right) right parotid gland, PTV1, spinal cord, PTV2, and left parotid gland. All dose values are given in Gy.



### Pinnacle has...

- Typical DVH constraints
- gEUD objectives
- Constrained optimization





### Required planning objectives [1/3]

ROI	Туре	Constrain	Target cGy	% Volume	% Variation	Weight	Objective Value	а
LTPAROT_OPT =	Max EUD		Ĭ 1950			<u> </u>		1
RTPAROT_OPT	Max EUD		Ĭ 1400			Ĭ 1		1
POST AVOID 🗖	Max Dose		¥ 4000			Ĭ 1		
2CM OUT_60	Max EUD	<b>ゴ</b> ロ	[3200			Ĭ 2		1
2CM OUTER RIN	Max EUD		Ĭ 2700			Ĭ 1		Ĭ 1
MAND_OPT =	Max EUD		Ĭ 2750			<u> </u>		Ĭ 1
ORALCAV_OPT -	Max EUD		Ĭ 3850			<u> </u>		Ĭ 1
2CM OUT_60 =	Max Dose	크 ㅁ	Ĭ 5800			<u> </u>		
2CM OUTER RIN	Max Dose	1 -	Ĭ 4600			Ĭ2		

## Required planning objectives [2/3]

ROI	Туре	Constrain	Target cGy	% Volume	% Variation	Weight
PTV60_OPT =	Max Dose ⊒		Ĭ 6150			Ĭ 41
PTV60_OPT =	Min Dose =		[6150			Ĭ 25
INNER RING 60 🔟	Min Dose		Ĭ6150			Ĭ 65
5MM OUT RING <b>□</b>	Max Dose =		Ĭ 6000			<u></u> 1
PTV5250_OPT 🖃	Min Dose =		Ĭ 5400			Ĭ 10
PTV5250_OPT 🖃	Max Dose =		Ĭ 5600			Ĭ 33
5MM OUT RING 🖃	Max Dose =		Ĭ 5250			Ĭ5
SHOULDER AVO	Max Dose =		Ĭ 1900			<u> </u>
CORD + MARGIN⊒	Max Dose ⊒		Ĭ 4150			Ĭ 35

## Required planning objectives [3/3]

ROI	Туре	Constrain	Target cGy	% Volume	% Variation	Weight	Objective Value	а
OUTER BODY RIF	Max Dose =		Ĭ 4500			Ĭ2		
OUTER BODY RIF	Max EUD =		[800			Ĭ2		Ĭ 1
LARYNX_OPT =	Max EUD		Ĭ 4200			<u> </u>		Ĭ 1
ESOPH_OPT =	Max EUD ⊒	]	[2000			<u> </u>		Ĭ 1
5MM OUT RING 🔟	Max EUD =		<u> </u> 5300			Ĭз		<u> </u>
AVOID STREAK 🔟	Max EUD ⊒	] 🗆	<u> </u> 4000			Ĭз		<u> </u>
AVOID STREAK 🔟	Max Dose =		Ĭ 5000			<u> </u>		
6900 (DMPO)_1 =	Max Dose 🗀	] 🗆	Ĭ6150			<u> 1</u> 10		
PTV60_OPT =	Uniform Dose 🖃		<u>¥</u> 6150			<u> </u>		
NEED 52	Min Dose =		Ĭ5400			<u> </u>		
6000 (DMPO)_1 🔟	Max Dose =		Ĭ 6150			<b>[</b> 55		
PTV5250_OPT 🔟	Max DVH □		Ĭ5700	<u> </u>		Ĭ 10		
INN RING 5250 🖃	Min Dose =		<u> </u> 5400			Ĭ 15		

### PriOpt, Step 1

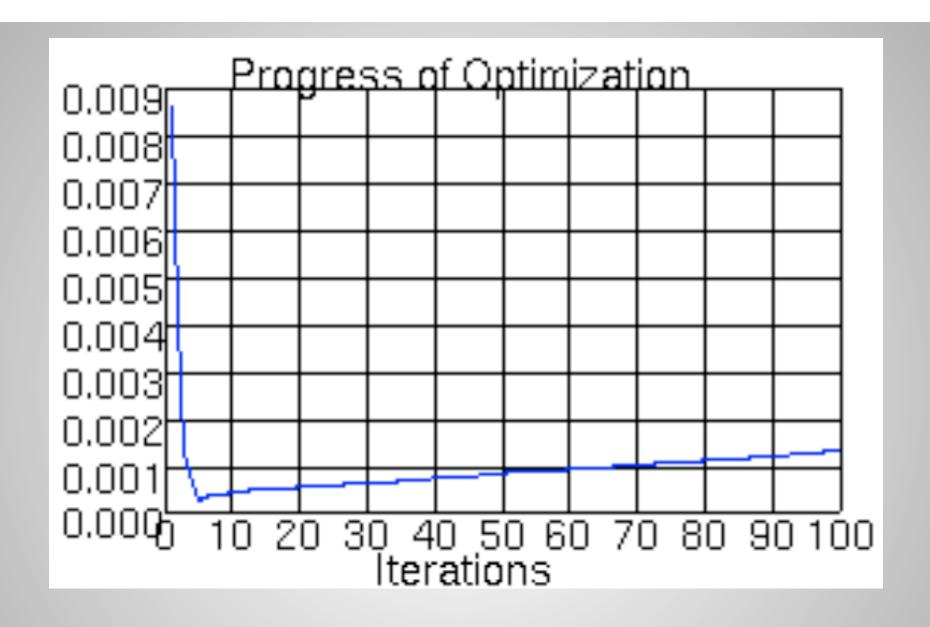
ROI	Туре	Constrain	Target cGy	% Volume	% Variation	Weight	Objective Value	а
CORD =	Max Dose 🗀	▼	Ĭ 4500					
BRAINSTEM =	Max Dose =	₹	Ĭ 5000					
PTV 6000 =	Min DVH =		[6000	Ĭ 98		Ĭ 1		
PTV 5250 🔟	Min DVH =		<u>[</u> 5250	Ĭ 98		Ĭ 1		
2CM OUT_60 =	Max EUD		Ĭ 4500			<u> 1</u>		Ĭ 10

# PriOpt, Step 2

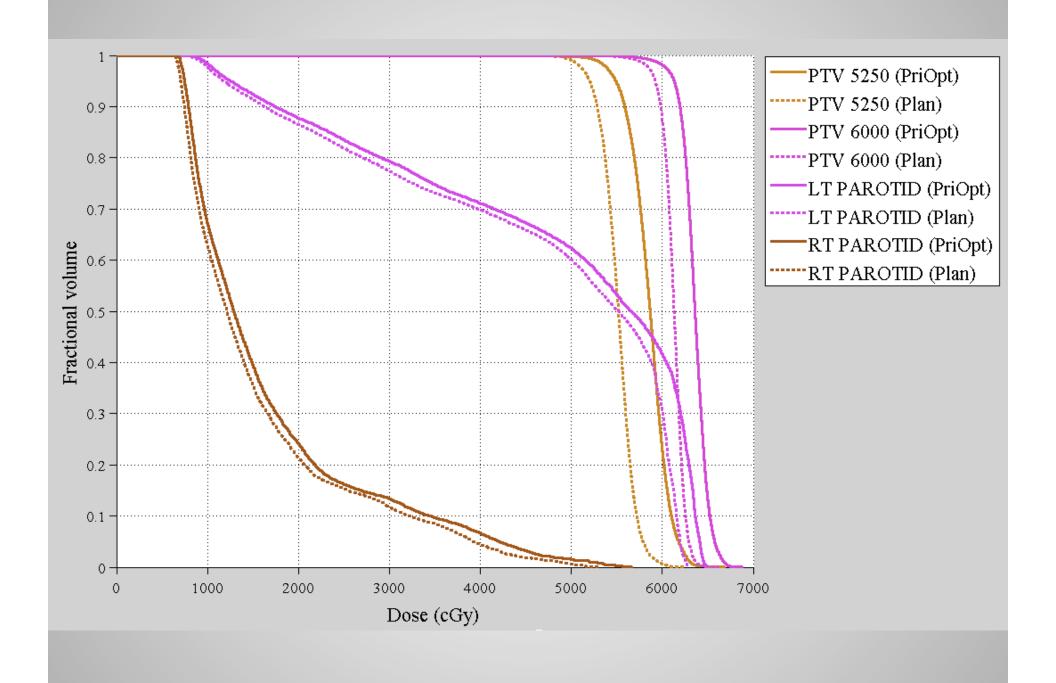
ROI		Туре		Constrain	Target cGy	% Volume	% Variation	Weight	Objective Value	а	gEUD
CORD	_	Max Dose	_	<b>       </b>	Ĭ 4500				1.30191e-07		
BRAINSTEM	_	Max Dose	_	<b>▼</b>	Ĭ 5000				2.43115e-09		
PTV 6000	_	Min Dose	_	<b>▼</b>	Ĭ 5580				8.78213e-07		
PTV 5250	_	Min Dose		<b>▼</b>	<u> 1</u> 4950				1.19567e-07		
2CM OUT_60	_	Max EUD	_	<b>▼</b>	Ĭ 4600				1.23133e-07	Ĭ 10	4603.61
LT PAROTID	_	Max EUD	_		Ĭ 4000			Ĭ 1	0.00270102	<u> 1</u>	4464.85
RT PAROTID	_	Max EUD	_		Ĭ 1200			Ĭ 1	4.14136e-05	<u> 1</u>	1217.27

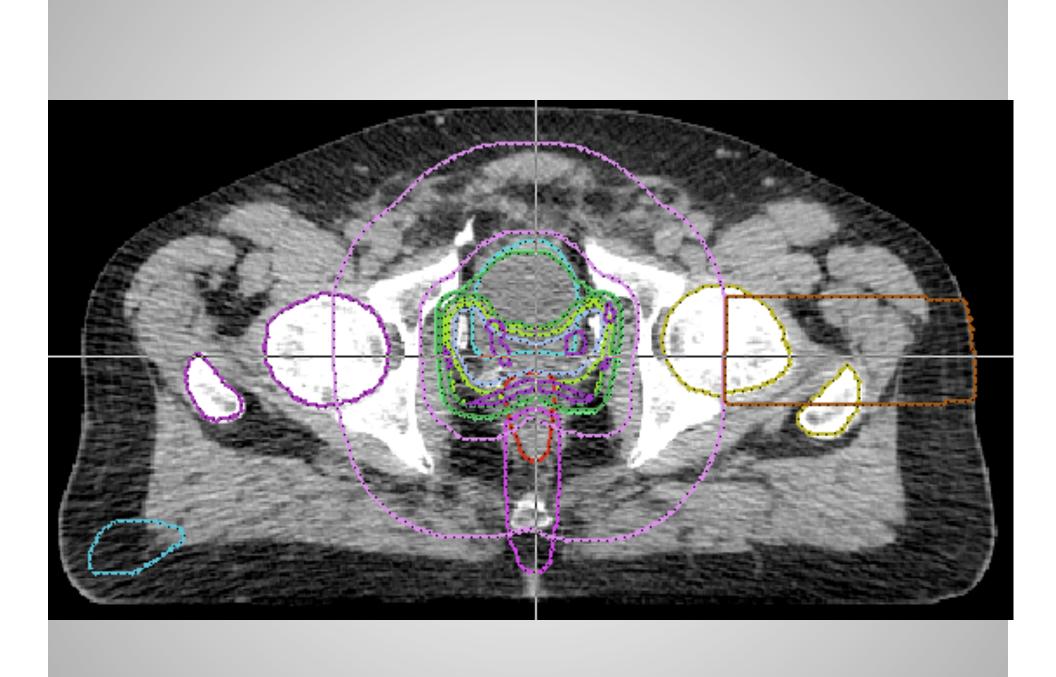
### PriOpt, Step 3 (last step)

ROI	Туре	Constrain	Target cGy	% Volume	% Variation	Weight	Objective Value	а
CORD =	Max Dose =	I▼	Ĭ 4500					
BRAINSTEM =	Max Dose 🖃	<b>▼</b>	Ĭ 5000					
PTV 6000 =	Min Dose =	<b>▼</b>	Ĭ 5580					
PTV 5250	Min Dose =	<b>▼</b>	<u> 1</u> 4950					
2CM OUT_60	Max EUD =	<b>▼</b>	Ĭ 4600					Ĭ 10
LT PAROTID 🔟	Max EUD =	<b>▼</b>	Ĭ 4500					Ĭ 1
RT PAROTID 🔟	Max EUD =	<b>▼</b>	Ĭ 1250					<u> 1</u>
2CM OUT_60	Max EUD =		Ĭ 3200			<u> 1</u>		Ĭ2
PTV 5250	Target EUD 🖃		Ĭ 5250			<u> 1</u>		Ĭ 1
PTV 6000 =	Target EUD =	] 🗆	Ĭ 6000			Ĭ 1		Ĭ 1



Slow!



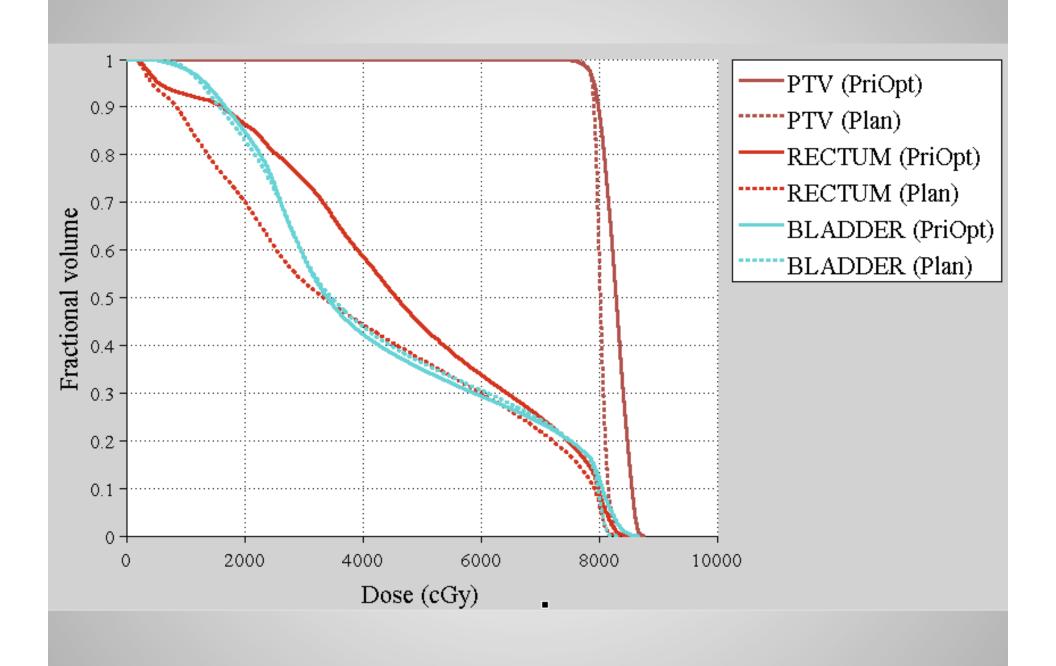


### Prostate objectives

ROI	Туре	Constrain	Target cGy	% Volume	% Variation	Weight	Objective Value	а
LTPAROT_OPT =	Max EUD =		Ĭ 1950			Ĭ 1		Ĭĭ 1
RTPAROT_OPT =	Max EUD =		Ĭ 1400			<u> </u>		1
POST AVOID 🗖	Max Dose =		Ĭ 4000			<u> </u>		
2CM OUT_60 =	Max EUD =		Ĭ3200			Ĭ2		<u> </u>
2CM OUTER RIN □	Max EUD =		Ĭ 2700			Ĭ 1		<u> </u>
MAND_OPT =	Max EUD =		[2750			Ĭ 1		1
ORALCAV_OPT □	Max EUD =		Ĭ 3850			Ĭ 1		<u> </u>
2CM OUT_60 =	Max Dose =		Ĭ 5800			Ĭ 4		
2CM OUTER RIN □	Max Dose =		Ĭ 4600			Ĭ2		
OUTSIDE RING 1=	Max Dose =		Ĭ6175			<u> </u>		
PTV	Max Dose =		[8662			<u> 1</u>		
PTV	Min Dose =		Ĭ 7875			<u> </u>		

### Prostate PriOpt Step 3 (final step)

ROI	Туре	Constrain	Target cGy	% Volume	% Variation	Weight	Objective Value	а
OUTSIDE RING 1=	Max DVH □	<b>▼</b>	<u></u> 6175	<u></u> 12				
PTV	Max Dose =	<b>▼</b>	Ĭ 8662					
PTV	Min DVH ⊐	<b>▼</b>	Ĭ 7875	<u></u> 98				
RECTUM =	Max EUD □	<b>▼</b>	Ĭ 6180					<u></u> 5
BLADDER =	Max EUD □	<b>▼</b>	Ĭ6180					Ĭ5
OUTSIDE RING 1=	Max EUD □		Ĭ 2000			Ĭ 1		Ĭ 2
OUTSIDE RING 8.	Max EUD □		Ĭ 5000			<u> </u>		<u> 10</u>
extern_minus_ptv=	Max Dose =	₹	Ĭ 4500					



### **Issues**

- Can be slow
- Automation (scripting) may be hampered by the need to keep the objective function from 'blowing up'. That is, objectives shouldn't be too far off what was available from earlier steps.

### Problems with 'PriOpt'

- What if there could be a BIG gain in a lower priority outcome for a small loss in a higher priority outcome?
- ...the urge to make small changes ('tweaking')
- Speed...