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Robust Optimization of Intensity Modulated Particle (Hadron) Therapy Dose Distributions

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## Robustness of a Radiation Dose Distribution

- Resilience of a dose distribution in the face of factors that cause uncertainty
- Degree of confidence that dose distribution seen on a treatment plan is delivered

## **Robustness Can be Improved**

- Through reduction in uncertainties
- Through margin assignments for 3DCRT with photons or particles and for IMRT
- Through robust optimization for Intensity-modulated particle therapy (IMPT):
  - Incorporates in the optimization process of factors that cause uncertainty in dose distributions

## **Robust Optimization**

- A field of optimization theory evolving since the 1950's
- Applications in statistics, finance, manufacturing engineering, chemical engineering, medicine, ...
- Relatively new to radiation therapy

## Sources of Uncertainty in Dose Distributions

- Approximations in dose calculation algorithms
  - e.g., due to passage of beams through heterogeneities
- Uncertainties in beam characteristics
- Uncertainties in CT data and in factors to convert them to numbers for dose calculations

Sources of Uncertainty in Dose Distributions (cont'd)

- Inter-fractional variations
  - Set up variability
  - Anatomy changes (tumor shrinkage, weight gain or loss, ...)
- Intra-fractional motion
- Relative biological effectiveness (RBE)

# Particle Therapy is More Vulnerable to Uncertainties

# Physical characteristics that make particle therapy attractive are also its Achilles heel



Intensity-Modulated Particle Therapy is the Most Powerful Tool in Radiation Therapy

# But it is also the most vulnerable to uncertainties (> PSPT, >> IMRT)

## Multi-Field Optimized IMPT













## Strategies for Robust Optimization are Required

# ProbabilisticWorst case analysis

## **Probabilistic Approaches**

- Probability distribution for each uncertainty is assumed to be known
- A sufficiently large set of uncertainty scenarios are sampled randomly
- IMPT optimization process minimizes the expectation value of the objective function
  - Finds the dose distribution that aims to satisfy the criteria for every uncertainty scenarios
    Unkelbach, PMB

2007, Med Phys 2009

## Probabilistic Robust Optimization – Illustrative Example

Conventionally Optimized



## "Worst Case" Approaches

- Analogous to PTV and PRV margins-based approach in IMRT
- Two strategies
  - minimax
  - Voxel-by-voxel

## Worst Case Analysis – minimax

- Multiple scenarios of maximum range and set up uncertainties considered
- Example:
  - $\pm \delta x$ ,  $\pm \delta y$ ,  $\pm \delta z$  shifts (= PTV margins)
  - ±δr for range uncertainty
  - Nominal
- One dose distribution computed for each uncertainty scenario in each optimization iteration
- The scenario with the worst (maximum) score selected in objective function minimization process
- Criteria are intended to be met under the worst case
   Fredrikson (2011) and Chen (2012)

## Worst Case Analysis – Voxel-by-Voxel

- Multiple scenarios of maximum range and uncertainty considered (same as for minimax)
- One dose distribution computed for each uncertainty scenario in each optimization iteration
- Worst case dose in <u>each voxel (minimum in</u> <u>target voxel and maximum in normal tissue</u> <u>voxel</u>) selected to compute objective function and minimize the objective function
- More conservative than minimax (uncertainty combination unrealistic), but ...

Pflugfelder (2008) and Liu (2012)

## Voxel-by-Voxel Worst Case Robust Optimization

## A Lung Example

### Robustness Quantification and Robust Optimization Example A recent lung case from P01 2008-0133 protocol



The patient was randomized to IMRT. Both PSPT and IMRT plans could deliver only 66 GyR limited by MLD constraint (22 GyR).

74 GyR could be designed for this patient using IMPT with MLD=17 GyR.

Zhang, Liu and Mohan





Liu, Zhang and Mohan



**PTV-based** optimization

### **Robust optimization**

## Voxel-by-Voxel Worst Case Robust Optimization

## **A Base of Skull Example**

Nominal dose distribution

> 7800.0 7250.0 6000.0 5400.0 4000.0

Moved inferiorly 3 mm





PTV-based optimized plan

**Robustly optimized plan** 

# PTV-based Optimized Base of Skull IMPT





# Robustly Optimized Base of Skull IMPT

## Area under an RMSD-Volume Histogram (RVH) curve may be a good metric of plan robustness



## H&N IMPT Robust Optimization -10 patient planning study



## **Robust Optimization**

## **PTV-Based Optimization**



## **Normal Tissue Sparing**



## Why Does Robust Optimization Lead to Superior Robustness AND Superior Quality?

Other Factors That Affect IMPT Robustness

- Number of beams
- Spot sizes,
- Energy spacing
- Spot density



- Significant progress made in understanding of the issues
- Approaches to robust optimization have been developed
- Only set up and range uncertainties considered

## What more needs to be done ...

- Comparative effectiveness of alternative approaches to robust optimization
- Consideration of intra-fractional and inter-fractional anatomy variations
- Incorporation of RBE uncertainties
- Quantitative criteria of robustness and threshold of acceptable robustness

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## Summary

- A robustly optimized plan was generated using in-house system for patient David (1033946)
- The plan was imported into Eclipse to have the final dose calculations.
- The robustly optimized plan has worse target coverage (mainly CTV57) than clinically approved plan in nominal scenario (shown in Eclipse) but has better coverage in the worst case scenario using robust analysis.
- The robustly optimized plan is more robust for target coverage.
- The robustly optimized plan also achieved better normal tissue sparing for majority OARs such as parotids, oral cavity.
- The worse CTV coverage of CTV57 are mainly due to the two different dose calculation algorithms used for inhouse system and Eclipse.

## DVH comparison between robust plan (triangle) and clinical approved plan

<del>(square)</del>



### Selection $\lambda$ Registration $\lambda$ Contouring $\lambda$ Field Setup $\lambda$ Plan Evaluation

Fields Dose Prescription Dose Statistics Plan Sum

View	DVH Line	Structure	Approval Status	Plan	Course	Volume (cm <sup>3</sup> )	Dose Cover.[%]	Sampling Cover.[%]	Min Dose (cGv)	Max Dose (cGv)	Mean Dose (cGv) 🛛 🔿
~		Whole Brain	Unapproved	MFORbst	BOT_70CGE	1522.6	100.0	100.0	0.0	4501.8	82.1
~		Whole Brain	Unapproved	APVD_SJF	BOT_70CGE	1522.6	100.0	100.0	0.0	3605.5	124.7
~	A	Brain Stem	Unapproved	MFORbst	BOT_70CGE	25.0	100.0	100.0	0.0	1449.9	206.5
~		Brain Stem	Unapproved	APVD_SJF	BOT_70CGE	25.0	100.0	100.0	4.4	1075.8	269.0
~	A	Spinal Cord	Unapproved	MFORbst	BOT_70CGE	27.6	100.0	100.2	0.0	2593.5	799.7
~	<b>A</b>	Oral Cavity	Unapproved	MFORbst	BOT_70CGE	204.0	100.0	100.0	0.0	5478.7	892.3
~		Oral Cavity	Unapproved	APVD_SJF	BOT_70CGE	204.0	100.0	100.0	0.0	5207.4	987.8
~		Spinal Cord	Unapproved	APVD_SJF	BOT_70CGE	27.6	100.0	100.2	0.0	3376.0	1112.8
~	<b>A</b>	Lt Parotid	Unapproved	MFORbst	BOT_70CGE	24.3	100.0	100.0	242.0	5596.8	1904.3
	<u>^</u>	Rt Parotid	Unapproved	MFORbst	BOT_70CGE	27.0	100.0	100.0	475.7	5187.4	1999.4
~	<mark>.</mark>	Rt Parotid	Unapproved	APVD_SJF	BOT_70CGE	27.0	100.0	100.0	487.0	5708.8	2120.8
		Lt Parotid	Unapproved	APVD_SJF	BOT_70CGE	24.3	100.0	100.0	724.6	6102.9	2495.0
~	<b>_</b>	Esophagus	Unapproved	MFORbst	BOT_70CGE	1.6	100.0	100.3	754.6	3919.0	2496.7
		Ecophogue	Unannrovad	ADI/ID OIE	DOT 70000	1 6	100.0	100.0	0 61 0	4802.4	1 0330

## Comparing of DVH and DVH bands for robustly optimized plan and clinically approved plan





Robust ly optimi zed plan

> Clinical ly approv ed plan

# Dose Distributions for robustly optimized plan



# Dose Distributions for robustly optimized plan



# Dose Distributions for robustly optimized plan



## **Target Coverage**





The robust plan has slight worse coverage in terms of D99 and D95 for CTV70, CTV63 and CTV57

 The robust plan is more heterogeneous as indicated by the HI index and D1





The larger, the better for

The smaller, the better for D1 and

## Target Robustness (1)









The robust plan has better coverage in terms of D99 and D95 for **CTV70**, **CTV63** and CTV57 in worst case scenarios The robust plan is more homogeneous as indicated by the HI index and D1 in worst case scenarios

The larger, the better for

The smaller, the better for D1 and HI

# Target Robustness: variations of D99, D95, D1 and HI







The robust plan has narrower variations between worst case and nominal variations for the D99, D95, D1 and HI



## **Critical Organ Sparing**



The robust plan achieved the better normal tissue sparing in terms of mean dose except for larynx which is 0.9 Gy higher than clincal plan.

# Clinical Organ Sparing in worst case scenarios



In worst case scenarios, the robust plan achieves better normal tissue sparing except Brain Stem which both plans are well below tolerance and Rt SubMandibular.

## Not accountable for with PTV margins





Tsunashima

## **Rationale for Particle Therapy**





## Physical properties

- Low dose at the entrance
- Low or no dose beyond the range
- Biological properties
  - Higher relative biological effectiveness (RBE)
  - Lower oxygen enhancement ratio







## Particles - Sources of their Strength

- Particles loose energy continuously as they penetrate the medium
- Linear energy transfer (LET) increases as they slow down
  - Most energy deposited near the end of the range
  - Have a finite range they stop
- Biological effectiveness is a function LET, so it increases correspondingly

## **Radiobiological Effectiveness of Protons**

- Proton RBE is assumed to be 1.1
- Claim: Clinical data do not suggest that RBE is different from 1.1
- In reality, RBE is a complex function of
  - Energy (LET)
  - Dose per fraction
  - Tissue/cell type, alpha/beta ratio
  - End point
- Another claim: Proton RBE is high in very narrow region and, thus inconsequential

## Variable RBE-Weighted Dose Effect for a CNS Patient

## 13 year old male with malignant meningioma



## Possible Effect of Variable RBE-Weighted Dose - Brain Necrosis in CNS Patients





## **So** ...

## WYS ≠ WYG

## **Final Words**

- Particle therapy is the future of RT
- Despite considerable superiority on paper of particle therapy (vs. photon therapy), evidence of superiority of particle therapy in clinical practice so far is unclear
- It is essential to understand the underlying physical, biological and clinical reasons for such lack of evidence through case-by-case and population-based analyses of clinical data

## **Final Words**

- Particle therapy is more complex, more vulnerable to physical and biological uncertainties
- Particle therapy is more costly
- Enormous opportunities for additional R&D to
  - Minimize uncertainties and their impact
  - Improve understanding of biology of particles
  - Identify optimum particle(s)
  - Reduce the cost of particle therapy
- Engineers, computer scientists, physicists, etc. in India can play an important role in many of these areas

## **Free Breathing and Gated PSPT Plans**



Non gated: Free breathing



Gated on 40~60% expiration phase

(Yoshikazu Tsunashima)

## **Eclipse Dose for Beam 1**



Beam 1 Eclipse Dose, axial view at isocenter

## MC Dose for Beam 1



Beam 1 MC Dose, axial view at isocenter



## Nasopharynx Treatment Plans IMRT vs. IMPT

**IMRT IMPT** Dose Gy Dose Gy 82.4 80.5 73.0 73.0 61.4 61.4 53.8 53.8 46.1 46.1 38.4 38.4 30.7 30.7 23.0 23.0 0.0 0.0

Lomax - PSI, Smith - MGH



Protons Through Base of Skull: 90 to 20% fall of increases from 6 to 32 mm

Urie, et al, Phys. Med. Biol., 1986, Vol. 31, No. 1, 1-15.











## Effect of Motion of IMRT vs. PSPT 4D - Static Dose Distributions

- Motion < 5 mm</p>
- For photons Difference < Less 5 Gy (RBE)</p>
- For protons
  - Difference ~ 10 Gy contralateral lung
  - Difference ~ 15 Gy near spine & ribs





# Why robust optimization leads to more robust plans?\*

- Robust optimization considerably reduces high dose gradients within each individual field, which helps improve the plan robustness. *This mechanism mostly happens within the targets (between the green dash lines)*
- Robust optimization leads to a dose distribution, which can be perturbed to follow the change of anatomy. This helps to improve the plan robustness while maintaining



\*(1) *Liu et al.*, Med. Phys. 39(6):3089-4001, 6/2012