Dosimetry Knowledge Based Treatment

## Planning

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**DUKE University Radiation Oncology** 

## Acknowledgement

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## Acknowledgement

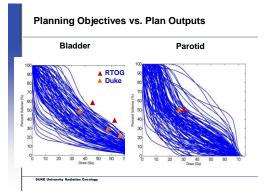
Funding from NIH, Varian Master Research Agreement

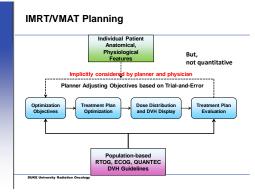
Technology licensed to Varian

 This study used the models from Duke, although the results are not identical to Varian's RapidPlan, they are indicative of RapidPlan's abilities.

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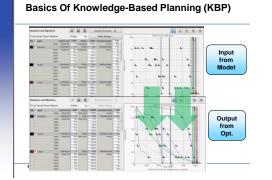


Basics Of Knowledge-Based Planning (KBP)

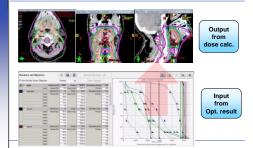
 Knowledge Models Provide Quantitative, Patient Specific, Optimization Parameters







# Basics Of Knowledge-Based Planning (KBP)



## Basics Of Knowledge-Based Planning (KBP)

Past experience based, anatomy driven

Identify anatomy and dosimetry features

- Machine learning to model correlations between anatomical and dosimetry features
- Predictions of new patient dose parameters are made based on the anatomy

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## 1. Identify anatomy and dose features

Site	OAR	Anatomical And Dosimetric Features
Prostate	Rectum	Distance to target histogram (DTH): PCS
	Bladder	Distance to OAR (DOH): PCS
		OAR volumes
HN	Parotids	PTV volume
	Oral cavity	Fraction of OAR volume overlapping with
	Larynx	PTV (overlap volume)
	Pharynx	Fraction of OAR volume outside the
	Spinal cord	treatment fields (out-of-field volume)
	Brainstem Mandible	Tightness of the geometric enclosure of
		PTV surrounding OAR
Spine	Spinal Cord	Curvature of specific OAR
SBRT	Spinal Cord + 2 mm	ourvature of specific OAR
		PTV dose homogeneity
		PTV hotspot

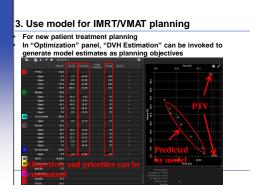
## 2. Generate a knowledge model

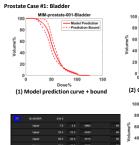


## 2. Generate a knowledge model - Model training

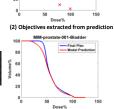
In "Model Configuration", select the plan to include in the training and then click "Train"

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(3) Eclipse optimization using objectives

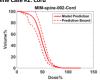


MIM-prostate-001-Bladder

× Objectives from Model

(4) Model prediction and plan DVH curve

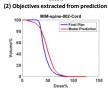




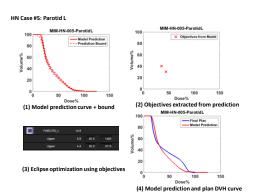
(1) Model prediction curve + bound



(3) Eclipse optimization using objectives



(4) Model prediction and plan DVH curve



#### Summary

- Modeling features
- Not limited to dosimetry features
- Beam angle, beam energy
- Trade-off preferences can be plan features too
- Presented prior to planning
- Decision support tool
- Post plan quality evaluation
- Clinical trials, protocol compliance
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