

Automatic Identification of Critical Areas of the Head and Neck for Refined Dose-Toxicity Analysis in Radiotherapy

- Design, implement, and evaluate an algorithm that creates spatially dependent dose features at the inter-organ level to identify specific areas of the head and neck that are more or less critical and sensitive to radiation damage.
- **What Students Will Do:**
 - Work with an existing database of over 900 radiation oncology patients
 - Develop a method to normalize the full anatomy based on standard set of commonly contoured regions
 - Develop an algorithm to identify non-contoured anatomical regions based existing contoured anatomy to find specific locations and spatially dependent dose features related to radiation induced toxicities
 - Generalize the model to enable the creation of spatially dependent features of the dose such as principle component analysis or gradient based features of the dose
 - Evaluate the impact of these features on taste, xerostomia and dysphagia toxicities with existing analysis tools.
- **Deliverables:**
 - An atlas normalization method that maps all patients to a common geometry
 - A method of defining a spatial region within the normalized atlas
 - An algorithm for creating spatially dependent features of the radiation dose in regions of the head and neck relative to existing contoured anatomy
 - Evaluated toxicity models for taste, xerostomia and dysphagia
- **Size group:** 1-3
- **Skills:**
 - 3D shapes, Volumetric Image Segmentation
 - Programming experience (SQL, C, C#, python)
- **Mentors:** Todd McNutt (tmcnutt1@jhmi.edu) Sierra Cheng (zcheng4@jhmi.edu)

Error Correction for Treatment Planning and a Learning Health System in Radiotherapy

- The goal is to improve the integrity of the anatomical 3D contours used in a learning health system with tools that can identify potentially erroneous data.
 - A second goal is to detect errant longitudinal measures such as weight.
 - The tools will tag or correct them and provide feedback in the clinical workflow.
- **What Students Will Do:**
 - Develop framework to run on the database that respond to errant data
 - Detect errant contours utilizing the database of hundreds prior patient's anatomical contours as a norm
 - Detect errant longitudinal data such as excessive changes in clinical status (e.g .weight)
 - Allow for customized data integrity checks to be added as needs are identified.
 - The system will be constructed to report findings in 3 ways
 - Listed report of all detected errors
 - Ability to tag data as suspect
 - Provide real time check on single data point entry when it is possible
- **Deliverables:**
 - The overall framework for errant contour detection
 - Documented API for developing new integrity checks
- **Size group:** 1-3
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Which patient will do better?

69-year-old man with Stage Squamous cell carcinoma, NOS of the Right Malignant neoplasm of tonsil

HPV	1+Yes	HPV high risk positive
p16	1+Yes	p16 Positive
Tobacco Use	0	Quit Smoking
Smoking Hx Starting Age	17	Quit Smoking
Smoking Hx Cigarettes/Day	20	
Smoking Hx Total Yrs	0	
Mucositis CTC3	0	None (mucositis)
Stom - Axile	0	No change over baseline
Stomary eye	0	None
Nose Bleeding	0-No	0-No
Xerostomia	1	Mild mouth dryness, slightly thickened saliva
Salivary Gland	1	Mild
Saliva - Hopkins	0	Normal
Vomiting CTC4	0	No vomiting (none)
Dysphagia	0	None
Dysphagia	0	None
Aspiration	0	None
Taste - Hopkins	1	Taste changes/loss not affecting ADLs
Percent Taste (%)	80	
Taste changes/loss	0	post surgical taste changes
Voice Changes	0	Normal
Larynx	0	None
Larynx RTOS	0	None
Bone Changes	0	None
Bone RTOS	0	None
Observance of Jaw	0	None
Dental teeth	0	None
Tongue	0	None
Pain NCI	0	None
Tinnitus CTC4	0	None
Cough CTC	0	absent

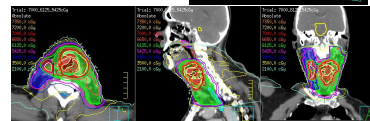
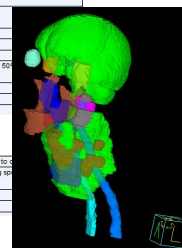
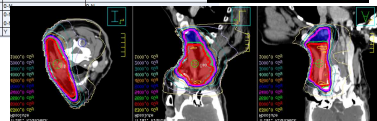
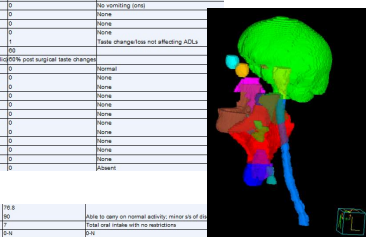
Weight (kg)	78.5	
KPS Score	90	able to carry on normal activity, minor loss of
ECOG	0-1	total oral intake with no restriction
PEG Tube (Placed)	0-1	0-1
PEG Tube (Used)	0-1	0-1
PEG Tube (Used)	0-1	0-1
ECOG	0-1	0-1

February 1, 2017

63-year-old man with T3 N2b M0 Stage IVA Squamous cell carcinoma, NOS of the Malignant neoplasm of larynx

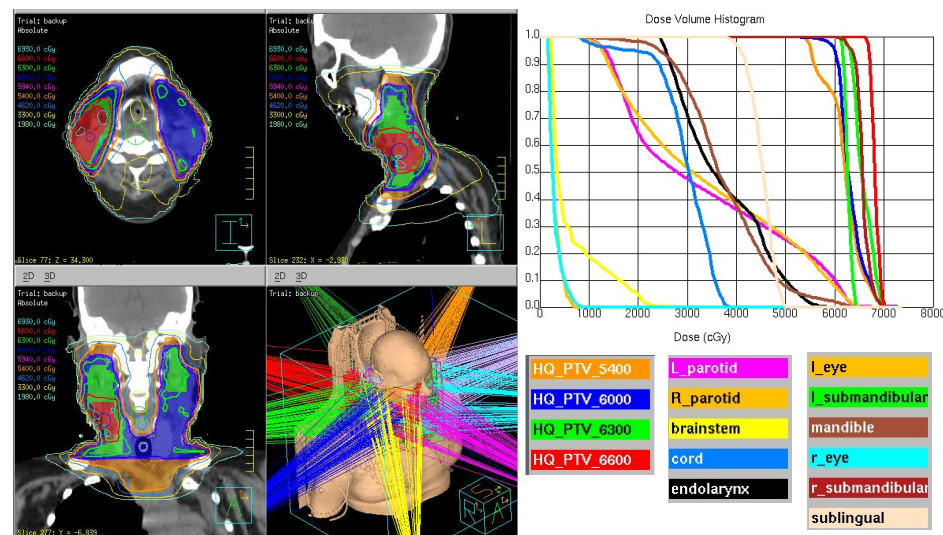
Tobacco Use	2	Quit Smoking
Smoking Hx Starting Age	23	
Smoking Hx Cigarettes/Day	40	
Smoking Hx Total Yrs	40	
Mucositis CTC3	0	None (mucositis)
Stom - Axile	0	No change over baseline
Stomary eye	0	None
Nose Bleeding	0-No	0-No
Xerostomia	0	None (xerostomia)
Salivary Gland	0	Normal
Saliva - Hopkins	0	Normal
Dysphagia	0	None
Dysphagia	0	None
Aspiration	0	None
Taste - Hopkins	1	Taste changes/loss not affecting ADLs
Percent Taste (%)	80	
Voice Changes	4	Disabling, non-understandable, requires voice aid > 50
Larynx	4	Disabling, non-understandable, requires voice aid > 50
Tinnitus	0	None
Pain NCI	1	Mild pain not interfering with fx
Cough CTC	0	absent

KPS Score	70	Cares for self, unable to carry on normal activity or to
ECOG	0	total oral intake of multiple consistencies requiring sp
PEG Tube (Placed)	1-1	1-1
PEG Tube (Used)	1-1	1-1
ECOG	1	1

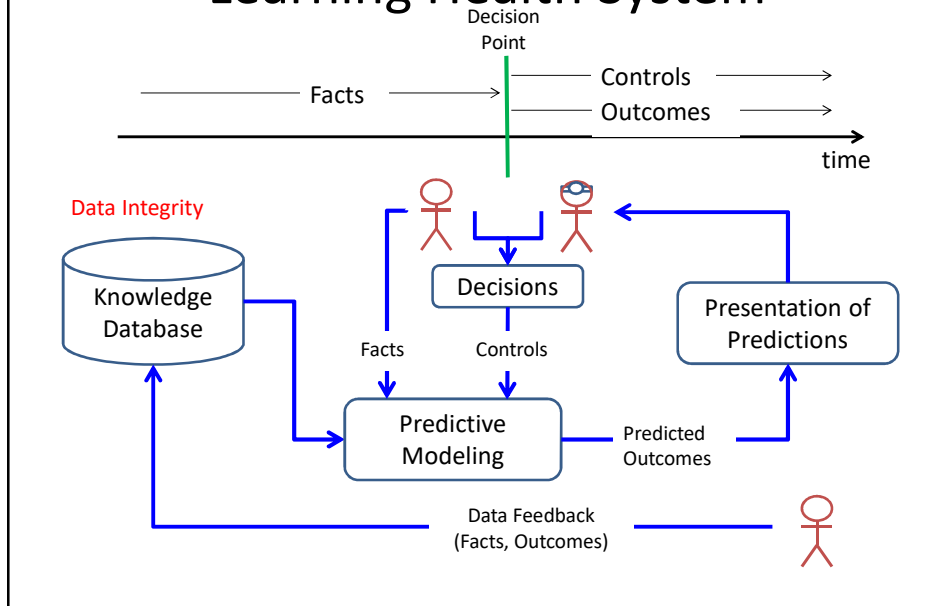


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Precision Radiotherapy Treatment Planning

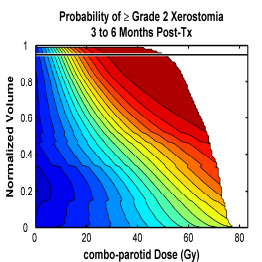
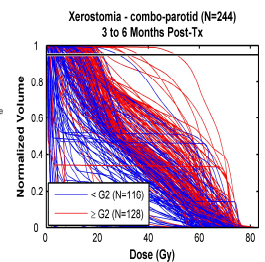
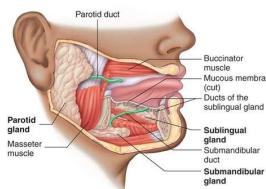
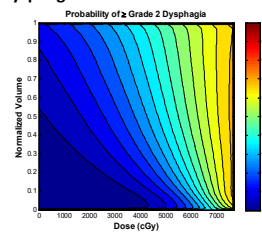
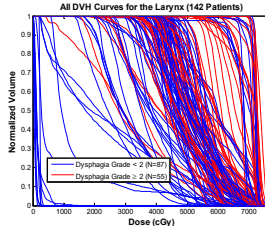
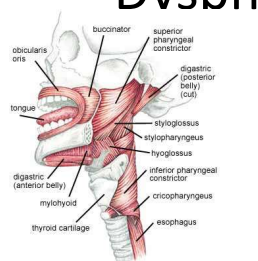


Learning Health System



Dysphagia and Xerostomia

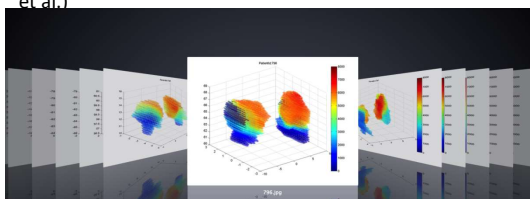
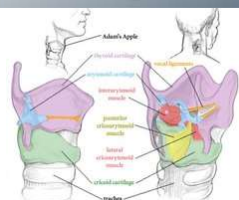
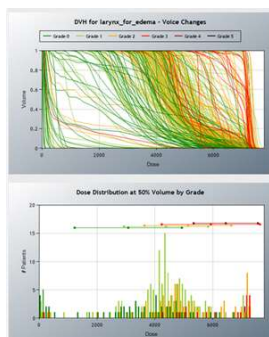
Larynx vs Grade ≥ 2 Dysphagia



Spatially dependent features of dose in the structures

(F. Marungo

et al.)



Method	Voice dysfunction n=99, n ₁ =8, n ₂ =91	Xerostomia n=364, n ₁ =275, n ₂ =89
Bagged Naïve Bayes (1000 iterations)	0.915	0.743
Bagged Linear Regression (1000 iterations)	0.905	0.737
Naïve Bayes	0.900	0.734
Linear Regression	0.896	0.731
Random Forest (1000 trees)	0.724	0.683
NTCP _{LKB}	0.596	0.700