

Challenges in the Omics Landscape

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Disclosures



• Nothing to disclose

Radiomics Publications







Limitations to Radiomics Expansion

- Image acquisition settings
- Radiomics software
- Patient artifacts
- Validity of statistical analysis
- Multi-institutional data

Garbage in Garbage **OUT**



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Image Acquisition Settings: Tube Current Feature mA·s Dependence b Cork Acrylic

- Images acquired on GE and Toshiba at 10-12 different mAs settings
- Variability compared to inter-patient variability of **106 NSCLC patients**
- Tube current unlikely to significantly affect radiomics features for patients



Mackin, Dennis, et al. "Effect of tube current on computed tomography radiomic features." Scientific reports 8.1 (2018): 1-10.



Image Acquisition Settings: Reconstruction Kernel

- Lung phantoms¹ and patients^{2,3} analyzed using different kernels
- Significant difference between sharp and smooth or standard and lung reconstructions



¹Zhao, Binsheng, et al. "Exploring variability in CT characterization of tumors: a preliminary phantom study." Translational oncology 7.1 (2014): 88-93. ²Zhao, Binsheng, et al. "Reproducibility of radiomics for deciphering tumor phenotype with imaging." Scientific reports 6.1 (2016): 1-7. ³Lu, Lin, et al. "Assessing agreement between radiomic features computed for multiple CT imaging settings." PloS one 11.12 (2016): e0166550.



Image Acquisition Settings: Voxel Size

- Phantom studies^{1,2} with CCR phantom rubber cartridge
- Patient study³ with 8 lung cancer patients reconstructed with 5 FOVs
- Resampling reduces impact of voxel size
 - Butterworth smoothing needed in patient study

¹Shafiq-ul-Hassan, Muhammad, et al. "Intrinsic dependencies of CT radiomic features on voxel size and number of gray levels." Medical physics 44.3 (2017): 1050-1062.

²Larue, Ruben THM, et al. "Influence of gray level discretization on radiomic feature stability for different CT scanners, tube currents and slice thicknesses: a comprehensive phantom study." Acta oncologica 56.11 (2017): 1544-1553.

³Mackin, Dennis, et al. "Harmonizing the pixel size in retrospective computed tomography radiomics studies." PloS one 12.9 (2017): e0178524.





Image Acquisition Settings: Inter-scanner Variability

- Scanned updated CCR phantom on 100 CT scanners using local head protocol, local chest protocol, and control protocol (specific acquisition and reconstruction settings to minimize vendor differences)
- Total Variability = $\sqrt{\sigma_{manufacturer}^2 + \sigma_{scanner}^2 + \sigma_{residual}^2}$
- Total variability in control 57% less than total in chest
- Total variability in control 52% less than total in head
- Control protocol reduce variability >50%





Image Acquisition Settings: Summary

- Tube Current: not likely to significantly impact features
- Reconstruction Kernel: do not use dissimilar kernels together
- Voxel Size: resampling and smoothing reduce impact
- Control Protocol: reduces variability >50% compared to local head and chest protocols



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Radiomics Software: Intersoftware Variability

- Comparison of 2 in house algorithms, IBEX, and MaZda for 39 HN patients
- Significant differences but high ICC values on first order features
- Package defaults for GLCM parameters very different





Radiomics Software: Intersoftware Variability

- Analyzed 105

 esophageal cancer
 patients with in house,
 IBEX, PyRadiomics with
 8 features
- Logistic regression to classify radiation pneumonitis
- Differences in classification ability



Foy, Joseph J., Samuel G. Armato, and Hania A. Al-Hallaq. "Effects of variability in radiomics software packages on classifying patients with radiation pneumonitis." Journal of Medical Imaging 7.1 (2020): 014504.

Radiomics Software: Initiatives ^A

- Image Biomarker Standardization
 Initiative
- 25 research teams with different radiomics software
- Iteratively analyzed features from a digital phantom and CT scan of lung cancer patient
- Final data set: TCIA data set of multi-modality imaging of 51 sarcoma patients
 - 167 features good reproducibility





Zwanenburg, Alex, et al. "The image biomarker standardization initiative: standardized quantitative radiomics for high-throughput image-based phenotyping." Radiology 295.2 (2020): 328-338.



Radiomics Software: Initiatives

- Feature definitions published
- Compliance with IBSI can be checked against reference values



Zwanenburg, Alex, et al. "The image biomarker standardization initiative: standardized quantitative radiomics for high-throughput image-based phenotyping." Radiology 295.2 (2020): 328-338.



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Patient Artifacts



- Streak artifacts
 - Features values robust to removing up to 50% of volume
- Presence of bone
 - Affects HU but difference minimal compared to variability among patients



Ger, Rachel B., et al. "Practical guidelines for handling head and neck computed tomography artifacts for quantitative image analysis." Computerized Medical Imaging and Graphics 69 (2018): 134-139.



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JOHNS HOPKINS

Validity of Statistical Analysis ⁴

- Feature selection before cross validation can cause large positive bias
- 10 datasets
 - 8 high dimensionalfewer samples than features
- Datasets with higher dimensionality more prone to positive bias



Validity of Statistical Analysis:

- Multiple hypothesis testing correction¹
- Independent validation dataset¹
- 22 checklist items from Transparent Reporting of multivariable prediction model for Individual Prognosis or Diagnosis²



¹Chalkidou, Anastasia, Michael J. O'Doherty, and Paul K. Marsden. "False discovery rates in PET and CT studies with texture features: a systematic review." PloS one 10.5 (2015): e0124165. ²Collins, Gary S., et al. "Transparent reporting of a multivariable prediction model for individual prognosis or diagnosis (TRIPOD): the TRIPOD statement." Journal of British Surgery 102.3 (2015): 148-158.

Validity of Statistical Analysis:

Used Prediction model Risk Of Bias Assessment Tool (PROBAST) for prognostic models using machine learning in oncology



Dhiman, Paula, et al. "Risk of bias of prognostic models developed using machine learning: a systematic review in oncology." Diagnostic and Prognostic Research 6 (2022).



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Multi-institutional Data



- Most studies are single institution data
- Multi-institutional data incorporate many of the differences in acquisition covered and differences in patient demographics
- TCIA
 - 151 human collections, 10 phantom collections

Questions Remaining



• What about other modalities?

	Robustness		Reproducibility	Classification performance		
	Image acquisition	Reconstruction	Segmentation	Post-processing	Feature extraction	JOHNS HOPKINS MEDICINE
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MRI	 Field strength Sequence design Matrix size (acquired) Field of view Slice thickness Acceleration techniques Vendor Contrast timing Movement 	 Matrix size (reconstructed) Reconstruction technique 	 Manual 2D Manual 3D Semi-automated 2D Semi-automated 3D Automated 2D Automated 3D Automated 3D Size of the ROI 	 Image interpolation ('resampling' / 'rescaling') Grid alignment Pixel sizing Intensity discretisation ('rebinning') Normalisation 	 Mathematical formula Post-processing platform 	-
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PET	 Field of view / pixel spacing Slice thickness Injected activity Acquisition time Scan timing Duty cycle Vendor Movement 	 Reconstruction matrix Slice thickness Reconstruction technique (algorithm, PSF, FOV, subsets, iterations, FWHM) Attenuation correction 	 Manual 2D Manual 3D Semi-automated 2D Semi-automated 3D Automated 2D Automated 3D Automated 3D Size of the ROI 	 Image interpolation ('resampling' / 'rescaling') Grid alignment Pixel sizing Intensity discretisation ('rebinning') Normalisation 	 Mathematical formula Post-processing platform 	-

Van Timmeren, Janita E., et al. "Radiomics in medical imaging—"how-to" guide and critical reflection." Insights into imaging 11.1 (2020): 1-16.

Questions Remaining



- What about other modalities?
- What about segmentation variability?

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Questions Remaining



- What about other modalities?
- What about segmentation variability?
- What about deep learning radiomics?

Garbage in Garbage **OUT**



Summary

- Resample images, use only similar reconstruction kernels, consider radiomics control protocol
- Different software may not be compatible, consider following IBSI
- Be careful in statistical analysis or false positive results may occur



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Thank You!