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Bio-X & Stanford Cancer Institute Stanford Biomedical Informatics Graduate Program 07/31/2018 <u>Stanford University</u>

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### Role of Radiologic Imaging in Oncology

- · Diagnosis, staging, response evaluation
- Routinely used, noninvasive, repeatable
- Image entire tumor & surrounding tissue
- Current radiology interpretation is limited.
   Subjective: inter-/intra-observer variations
- Diagnostic: few semantic features (typically <10)</li>
   Response criteria: tumor size, SUVpeak
- We need to do better.
  - Quantitative
  - Comprehensive



### Radiomics: the Rationale and Process

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 Imaging phenotypes are driven by underlying pathophysiology – FDG-PET SUV and metabolism.

- Tumor segmentation, high-throughput feature extraction – Convert radiologic image to quantitative 'omic' data
- Correlate with clinical outcomes: putative imaging markers – Aid in diagnosis, as well as prognosis & treatment response





Lambin et al, Eur J Cancer, 2012 Gillies et al, Radiology, 2017

Segmentation Feat

### **Outline** of this Talk

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- · Current status and recent developments in radiomics
  - Claimer: Not a comprehensive review (Lambin, Nat Rev Clin Oncol 2017)
     Translational: how radiomic tools are applied to a clinical problem
  - Theme: better characterize intra-tumor heterogeneity
- Conventional radiomics: Gross tumor
- Histogram, morphology, texture
- · Emerging paradigms: Tumor subregions/habitats
- Volume; Texture
- Spatial interaction
- · Radiogenomics
- · Challenge, outlook

### Radiomic Signature for Lung Cancer Prognosis

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Clinical problem: Unable to predict which patients will develop distant metastasis after SABR; this makes decision regarding adjuvant therapy uncertain.
 Image signature = 2.1 x SUV<sub>peak\_2cc</sub> + 3.6 x Gaussian\_ClusterShade



Wu et al, Radiology, 2016







# Gross Tumor Features To Detect Recurrence

• A radiomic signature measured at post-treatment CT detected local recurrence after SABR for lung cancer, with an error rate of 24% in 45 patients, compared with physicians (average 35%).





#### 4 Gross Tumor Features To Evaluate Response

- 13.3% (149 out of 1119) of the radiomic features showed
- an R<sup>2</sup> above 0.85 between planning CT and CBCT. • CBCT preserved prognostic value: CI=0.66 (0.69 for CT).
- Potential of CBCT for monitoring radiation response?



### Beyond Gross Tumor: Multi-Region Analysis

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- · Current radiomics approach
  - Aggregate features from the bulk tumor
  - Assuming tumor is well mixed. Not spatially explicit.
- · Branched evolution causes regional differences within a tumor.











45 High-Risk Subvolume is Prognostic in Stage III NSCLC suv,,,, CI = 0.63 \* wank p = 0.34 CI = 0.61 Logrank p = 0.79 Freedom from nh-risk sub-.... MTVs CI = 0.76 Logrank p = 0.002 CI = 0.63 Logrank p = 0.88 Green: < median Red: >= median Freedom ree e se se de la característic è 8 A 4 a second and second services and the service of percent. 14 Wu et al. IJROBP, 2016







# High-Risk Subvolume in HN Tumors · Large poorly perfused subvolumes of HNC at DCE MRI

persisting during the early course of chemo-RT associated with local or regional failure in 14 patients.



#### 45 Combine Radiomics with Multi-Region Analysis

- · Intra-tumor partitioning using multiparametric MRI of GBM
- Extract radiomic features for each subregion and gross tumor.



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Cui et al, Radiology, 2016

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### Prognostic Imaging Signature in GBM





#### 45 Early Response Prediction in Breast Cancer

Neoadjuvant chemotherapy is used to shrink tumor for breast-conserving •

surgery. 30% of patients have poor response, and yet suffer from the toxicity. Early response prediction would enable the use of alternative therapies. • .





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The change in texture of tumor subregion associated with rapid wash-out pattern on DCE-MRI predicted pathological complete response.

∆Tumor Volume: AUC=0.53 ∆Tumor Texture: AUC=0.65 ΔSubregion Volume: AUC=0.57 ΔSubregion Texture: AUC=0.79

> 20 Wu et al. JMRI 2016







Wu et al Radiology 2018



Subregions correspond to the highly, moderately, poorly perfused portion of the tumor.



Wu et al Radiology 2018



to stratify patients 24



\*Independent predictor of RFS adjusting for age, ER, PR, HER2, tumor volume. 25



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### Wu et al Radiology 2018



### Radiogenomics of Lung Cancer



A Quantitative CT Imaging Signature Predicts Survival and Complements Established Prognosticators in Stage I Non-Small Cell Lung Cancer

Johnen Lee, PHS." Bellang LJ, PSD," Yi Cut, PhD." Xiaoli San, MB. Jia Wu, PSD, "Hoi Zhu, ND, PSD, Jonning Yu, ND, PHD, Michael K, Ganathiener, MD, "Bibly M, Lou, X, MB, PSD," Hardrellian Divin, ND, PbB,"" and Rulflang LL, PMD."

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### Prognostic Value of Tumor Immunity ARMA Decology ( Original Investigation

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 Correlate gene expression with CT imaging signature. · Top enriched pathways are

related to immune response,

such as lymphocyte activation and chemotaxis.

Development and Validation of an Individualized Immune Prognostic Signature in Early-Stage Nonsquamous Non-Small Cell Lung Cancer

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- 19 cohorts of over 2400 patients
  Immune gene signature strongly
- prognostic, independent of stage, grade. Higher accuracy than existing gene signatures.

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### Challenges & Potential Solutions

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- Standardization NIH QIN, RSNA QIBA
  - Technical: Image acquisition: scanner, protocol, techniques
  - Analytical: Image reconstruction, feature definition, calculation
- Small sample size may lead to spurious findings
- Collaboration & data sharing TCIA; both imaging & clinical information
   Reproducibility
  - Need rigorous, truly independent validation.
- Biological interpretation
  - Correlating imaging with genomics may help beware of caveat

### Conclusion & Outlook

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- · Radiomics is a useful tool to discover new imaging markers.
  - Beyond gross tumor: intratumoral subregions/habitats
  - Mostly for initial characterization; less for response
  - Delta radiomics may not be optimal/sensitive to evaluate response
     Need better tools specific for detecting changes at serial imaging
- Prospective validation in RCT is required to establish clinical utility.

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