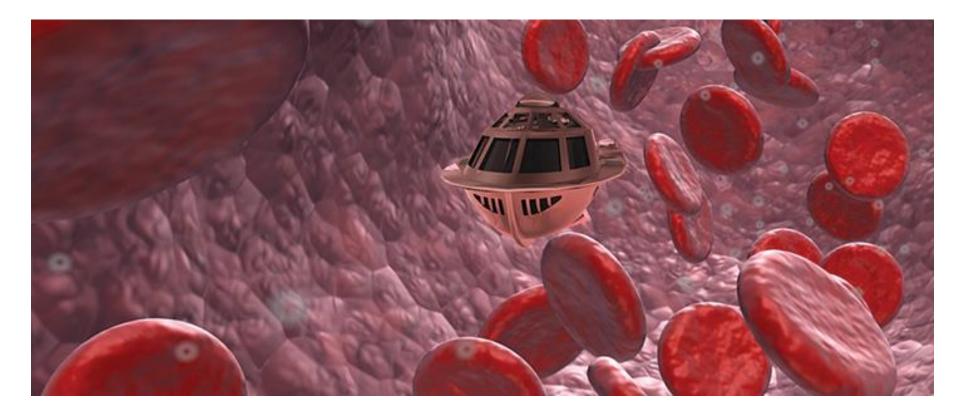
### **Spatio-temporal Sensor Integration, Analysis, Classification or**





## Can Exascale Cure Cancer?

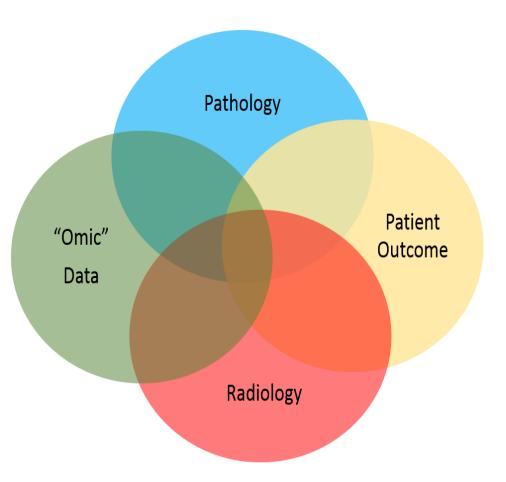
## Clusters, Clouds, and Data for Scientific Computing CCDSC 2014

#### September 2<sup>nd</sup> – 5<sup>th</sup>, 2014



## Multi-scale Integrative Analysis Story Brook

- Predict treatment outcome, select, monitor treatments
- Computer assisted exploration of new classification schemes
- Tumor heterogeneity, Immune response
- Reduce inter-observer variability in diagnosis



## **Pathomics, Radiomics**



Identify and segment trillions of objects – nuclei, glands, ducts, nodules, tumor niches ... from Pathology, Radiology imaging datasets

Extract features from objects and spatio-temporal regions

Support queries against ensembles of features extracted from multiple datasets

Statistical analyses and machine learning to link Radiology/Pathology features to "omics" and outcome biological phenomena

Principle based analyses to bridge spatio-temporal scales – linked Pathology, Radiology studies

FDA News Release

### FDA allows marketing of first whole slide imaging system for digital pathology

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For Immediate	April 12, 2017
Release	

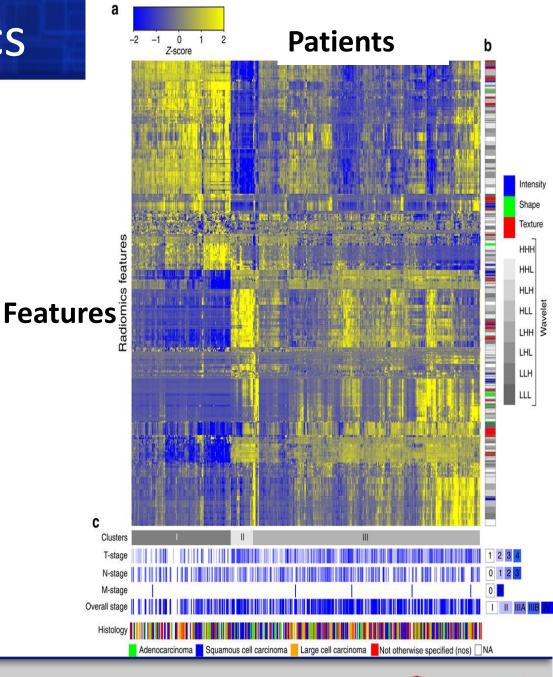
ReleaseThe U.S. Food and Drug Administration today permitted marketing of the Philips<br/>IntelliSite Pathology Solution (PIPS), the first whole slide imaging (WSI) system that<br/>allows for review and interpretation of digital surgical pathology slides prepared<br/>from biopsied tissue. This is the first time the FDA has permitted the marketing of a<br/>WSI system for these purposes.

## Radiomics

Decoding tumour phenotype by noninvasive imaging using a quantitative radiomics approach

#### Hugo J. W. L. Aerts et. Al.

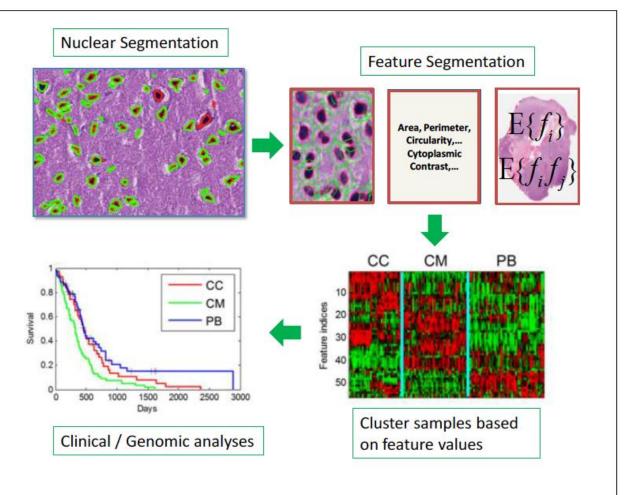
*Nature Communications* **5**, Article number: 4006 doi:10.1038/ncomms5006



\* Stony Brook Medicine

## Pathomics Integrative Morphology/"omics"

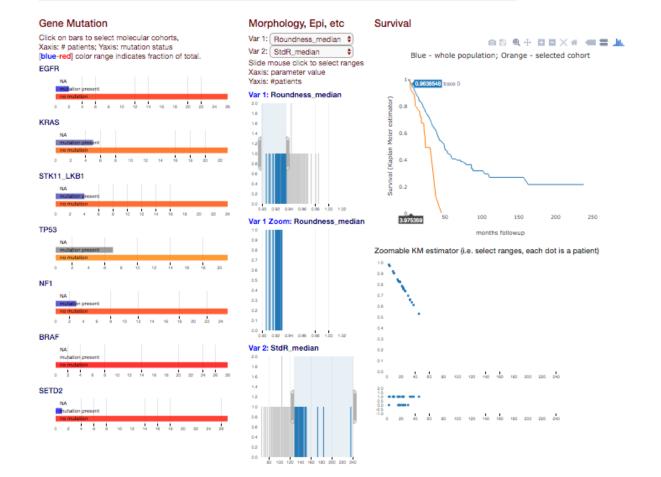
- Quantitative Feature Analysis in Pathology: Emory In Silico Center for Brain Tumor Research (PI = Dan Brat, PD= Joel Saltz)
- NLM/NCI: Integrative Analysis/Digital Pathology R01LM011119, R01LM009239 (Dual PIs Joel Saltz, David Foran)
- J Am Med Inform Assoc. 2012 Integrated morphologic analysis for the identification and characterization of disease subtypes.





#### Feature Explorer - Integrated Pathomics Features, Outcomes and "omics" – TCGA NSCLC Adeno Carcinoma University Patients

\_ . . . . . . . . . . . .



## SEER Virtual Tissue Repository

- Lynne Penberthy MD, MPH NCI SEER
- Ed Helton PhD NCI CBIIT Clinical Imaging Program
- Ulrike Wagner CBIIT Clinical Imaging Program
- Radim Moravec NCI PhD, NCI SEER
- Ashish Sharma PhD Biomedical Informatics Emory
- Joel Saltz MD, PhD Biomedical Informatics Stony Brook
- Tahsin Kurc PhD Biomedical Informatics Stony Brook
- Georgia Tourassi, Oak Ridge National Laboratory

Vision – Enable population/epidemiological cancer research that leverages rich cancer phenotype information available from Pathology tissue studies

NCIP/Leidos 14X138 and HHSN261200800001E - NCI



## SEER Virtual Tissue Repository

Pilot project that captures the core functionalities and processes needed in a Virtual Tissue Repository for SEER

- Supplement registry abstracts with images AND quantitative features
  - Curate Images; Extract and Curate Features; and Web-based Visualization
- Data → Breast and Pancreas Cancer Five Registries Currently 122 images and 450M objects
- High quality feature curation, using QuIP algorithms and tools, on a representative subset of data
- All images and features accessible via caMicroscope

Built with QuIP (ITCR U24 PI: Saltz) and caMicroscope (CBIIT CTIIP PI: Sharma)



### SEER VIRTUAL TISSUE REPOSITORY



- Create linked collection of de-identified clinical data and whole slide images
- Extract features from a sample set of images (pancreas and breast cancer).
- Enable search, analysis, epidemiological characterization
- Pilot focus on extreme outcome Breast Cancer, Pancreatic Cancer cases
- Display images and analyzed features

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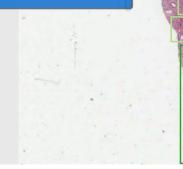
caMic Segment Curation App

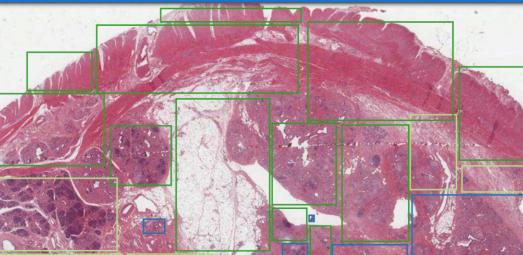
#### Case ID: PC\_055

#### Select Algorithm

wsi:r0.8:w0.8:13:u200:k20:j1
 wsi:r0.9:w0.8:13:u200:k20:j1
 wsi:r1.1:w0.8:13:u200:k20;j1
 wsi:r1:w0.8:13:u200:k20
 omar.awan\_composite\_input
 siwen.statistics\_composite\_input
 tigerfan7495\_composite\_input

Hide





#### **Curation of Segmentation Results**

#### **QuIP Segmentation Curation Web Application**

- 1. User interactively marks up regions and selects results from best analysis run for each region.
- 2. Selections are refined through review processes supervised by an expert Pathologist



### **TIL quantitation and distribution**



- The most common diagnostic tool in pathology is the H&E tissue image
- FDA just approved use of whole slide images in primary Pathology diagnosis
- TCGA dataset, which comprises 33 tumor types, contains over 30,000 tissue slide images.
- Link pattern of tumor infiltrating distribution to outcome, "omics", treatment
- Deep Learning TIL method requires modest training and curation – suitable for high throughput analyses

Importance of Immune System in Cancer Treatment and Prognet Stress Brook

- Tumor spatial context and cellular heterogeneity are important in cancer prognosis
- Spatial TIL densities in different tumor regions have been shown to have high prognostic value – they may be superior to the standard TNM classification
- Immune related assays used to determine Checkpoint Inhibitor immune therapy in several cancer types
- Strong relationships with molecular measures of tumor immune response – results to soon appear in TCGA Pan Cancer Immune group publications
- TIL maps being computed for SEER Pathology studies and will be routinely computed for data contributed to TCIA archive
- Ongoing study to relate TIL patterns with immune gene expression groups and patient response

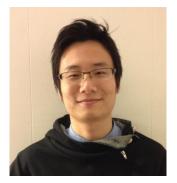
- Stony Brook, Institute for Systems Biology, MD Anderson, Emory Stony Brook group
- TCGA Pan Cancer Immune Group led by ISB researchers
- Deep dive into linked molecular and image based characterization of cancer related immune response

#### TCGA PAN CANCER ATLAS IMMUNE GROUP: INTEGRATIVE IMMUNE-IMAGE MOLECULAR ANALYSIS

- Joel Saltz BMI
- Tahsin Kurc BMI
- Dimitri Samaras CS
- Fusheng Wang BMI/CS
- Daifeng Wang BMI
- Le Hou Joint CS/BMI student
- Rebecca Batiste Pathology
- Vu Nyugen CS/BMI Student
- Tianhao Anne Zhao BMI/Pathology fellow

- Ashish Sharma Emory
- Alex Lazar MD Anderson
- Arvind Rao MD Anderson
- John Van Arnam MD Anderson
- Vesteinn Thorsson Institute for Systems Biology
- Ilya Shmulevich Institute for Systems Biology





Le Hou – Graduate Student Computer Science



Anne Zhao – Pathology Informatics Biomedical Informatics, Pathology (now Surg Path Fellow SBM)



Vu Nguyen– Graduate Student Computer Science

Deep Learning and Lymphocytes: Stony Brook Digital Pathology Trainee Team



Raj Gupta – Pathology Informatics Biomedical Informatics, Pathology

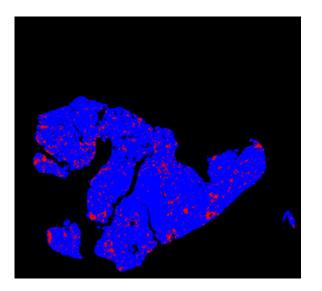




Ann Transl Med. 2016 Apr; 4(7): 142. doi: 10.21037/atm.2016.03.28 PMCID: PMC484239

#### Importance of tumor infiltrating lymphocytes in non-small cell lung cancer?

Roy M. Bremnes, 21,2 Tom Donnem, 1,2 and Lill-Tove Busund 3,4

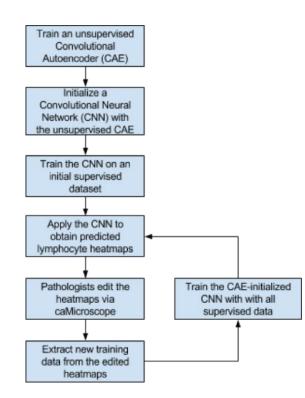


#### Percentage 5-year disease specific survival rates according to TNM and Immunoscore

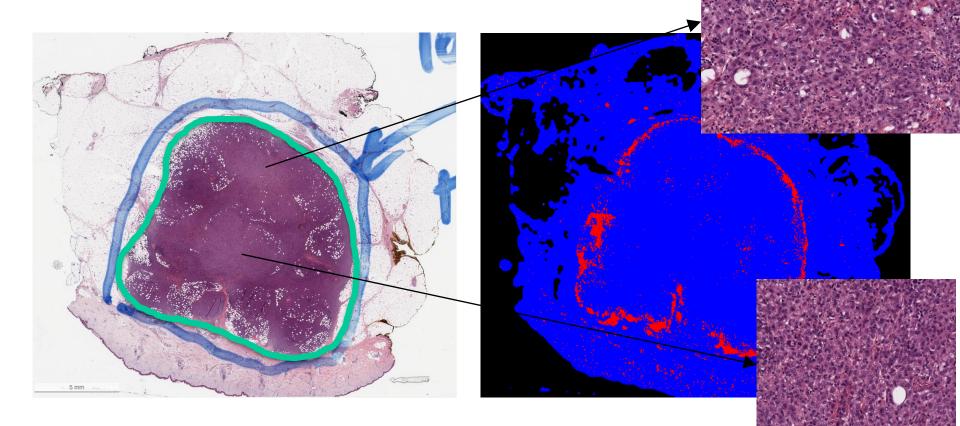
TNM-I	Immunoscore (stromal CD8)						
		Low	Intermediate	High			
	IA	71	75	87			
pStage	IB	59	73	83			
	IIA	51	58	61			
	IIB	25	55	61			
	IIIA	18	41	68			
5-year DSS rates							
≤ 40 %							
41-59 %							
60-80 %							
> 80 %							

### Deep Learning Training, Validation and Prediction

- Algorithm first trained on image patches
- Several cooperating deep learning algorithms generate heat maps
- Heat maps used to generate new predictions
- Companion molecular statistical data analysis pipelines

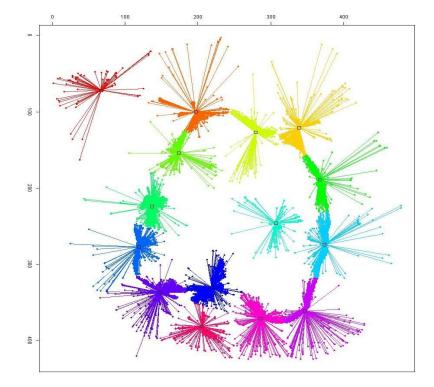


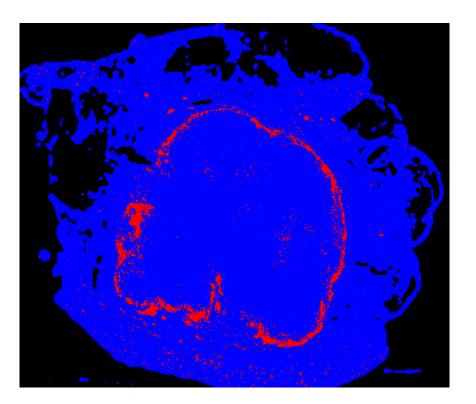
### Tumor Infiltrating Lymphocyte Maps and University Classification



### **Clustering and TIL Maps**



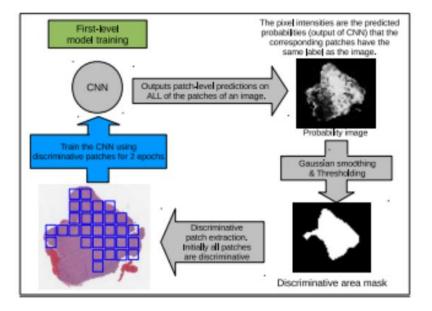


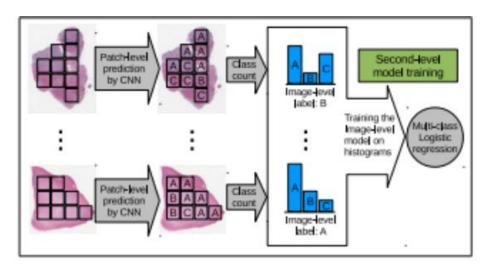


### Tumor Classification – Reduce Inter-observer variability

### Patch-Based Convolutional Neural Network for Whole Slide Tissue Image Classification

Le Hou, Dimitris Samaras, Tahsin M. Kurc, Yi Gao, James E. Davis, Joel H. Saltz; The IEEE Conference on Computer Vision and Pattern Recognition (CVPR), 2016, pp. 2424-2433







### Tumor Classification Results Stony Brook University

#### Methods Accuracy

- VGG16 features + BoW + SVM 0.667
  - 0.710 Patch-CNN + Voting
  - Patch-CNN + Max-pooling 0.710
    - Our method 0.771
- Pathologists' Agreement [M. Gupta 2015] 0.7-0.8 (on a similar dataset)

Confusion Matrix: OA is very hard even for pathologists		OD	OA	DA	AA	AO
Glioblastoma, Grade IV (GBM)	214		2		1	
Oligodendroglioma, Grade II (OD)	1	47	22	2		1
Oligoastrocytoma, Grade II & III (OA)	1	18	40	8	3	1
Diffuse Astrocytoma, Grade II (DA)	3	9	6	20		1
Anaplastic Astrocytoma, Grade III (AA)	3	2	3	3	4	
Anaplastic Oligodendroglioma, Grade III (AO)	2	2	3			1

Le Hou, Dimitris Samaras, Tahsin Kurc, Yi Gao, Liz Vanner, James Davis, Joel Saltz

#### Glioma is

- The most common brain cancer ٠
- The leading cause of cancer-related deaths in people under age 20

Information Technology for Cancer Research NCI Biomedical Informatics and Information Technology and NCI Cancer Imaging Program

 8+ major projects encompassing Pathology Informatics – will hear about several in the next three days

Stony Brook

- Broad recognition of importance of Radiology/Pathology/"omics" Integration
- Development of a Cancer Imaging Commons
- Peer institutions developing linked clinical/image sharing, feature extraction and trust fabric

# Thanks!

