



# Tools to Analyze Morphology and Spatially Mapped Molecular Data ITCR U24CA180924 Joel Saltz Department of Biomedical Informatics

Stony Brook University PI Meeting June 13-14 2016

Video Overview/Demonstration –

https://www.youtube.com/watch?v=kShwls0O5o4



### **Stony Brook University**

- Joel Saltz
- Tahsin Kurc
- Yi Gao
- Allen Tannenbaum
- Erich Bremer
- Jonas Almeida
- Alina Jasniewski
- Fusheng Wang
- Tammy DiPrima
- Andrew White
- Le Hou
- Furqan Baig
- M Reda El-MKhantar
- Tara Maloney

## **Emory University**

- Ashish Sharma
- Adam Marcus
- Dan Brat
- Roberd Bostick
- Yue Hou
- Ganesh Iyer

## **Harvard University**

- Richard Cummings
   UTK
- Jeremy Logan
- Scott Klasky
- Dave Pugmire

## Yale University

Michael Krauthammer

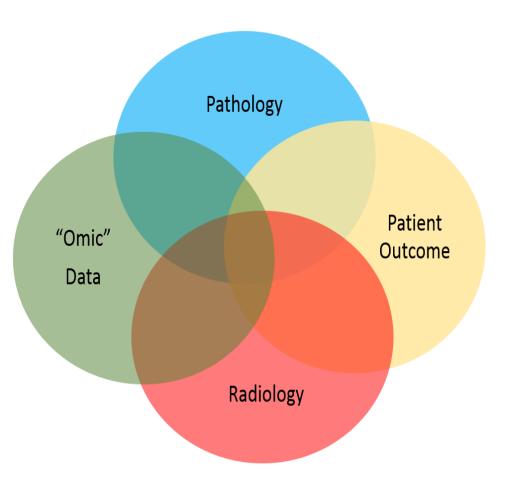
## **3DSlicer Pathology and FeatureScape for Linked Radiology/Pathology Image Analysis Competitions Collaboration between SBU, BWH and MGH Groups**



- Jonas Almeida, Stony Brook
- Nicole Aucoin, Brigham and Women's Hospital
- Erich Bremer, Stony Brook
- Tammy DiPrima, Stony Brook
- Andrey Fedorov, Brigham and Women's Hospital
- Yi Gao, Stony Brook
- Jayashree Kalpathy-Cramer, Massachusetts General Hospital
- Ron Kikinis, Brigham and Women's Hospital
- Tahsin Kurc, Stony Brook
- Arten Mamonov, Massachusetts General Hospital
- Steve Pieper, Isomics, Inc.
- Jean-Christophe Fillion-Robin, Kitware Inc.
- Bruce Rosen Massachusetts General Hospital
- Allen Tannenbaum, Stony Brook

## Multi-scale Integrative Analysis in Biomedical Informatics

- Predict treatment outcome, select, monitor treatments
- Reduce inter-observer variability in diagnosis
- Computer assisted exploration of new classification schemes
- Multi-scale cancer simulations





# Pathomics, Radiomics

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

Extract features

Queries against ensembles of features

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



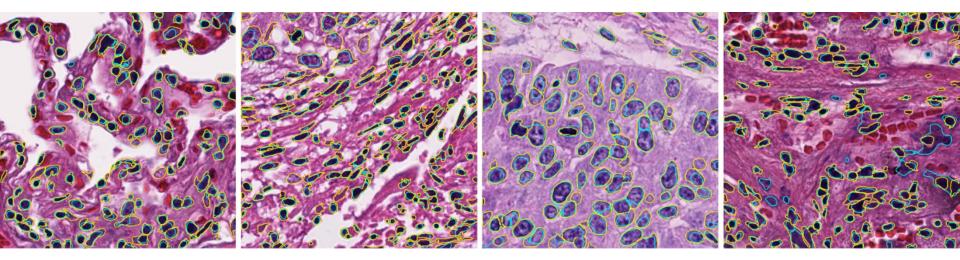
## Tools to Analyze Morphology and Spatially Mapped Molecular Data - U24 CA180924

- **Specific Aim 1** Analysis **pipelines** for multi- scale, integrative image analysis.
- Specific Aim 2: Database infrastructure to manage and query Pathomics features.
- Specific Aim 3: HPC software that targets clusters, cloud computing, and leadership scale systems.
- **Specific Aim 4:** Develop **visualization** middleware to relate Pathomics feature and image data and to integrate Pathomics image and "omic" data.



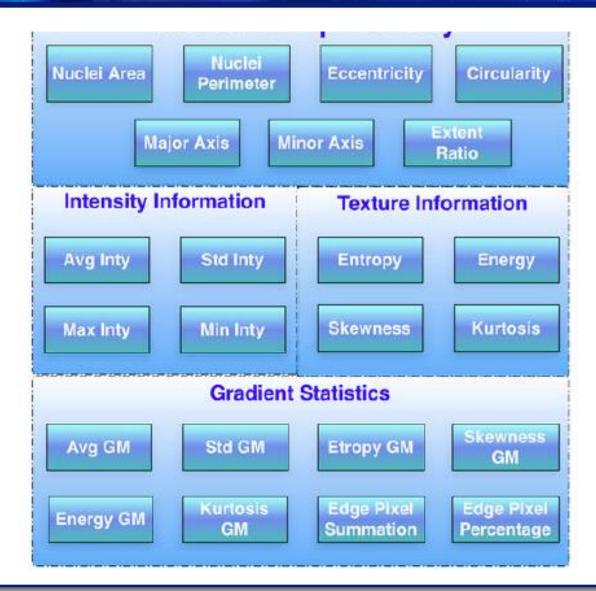
## **Robust Nuclear Segmentation**

- Robust ensemble algorithm to segment nuclei across tissue types
- Parameter exploration to optimize quality
- Systematic Quality Control pipeline encompassing tissue image quality, human generated ground truth, convolutional neural network critique
- Yi Gao, Allen Tannenbaum, Dimitris Samaras, Le Hou, Tahsin Kurc



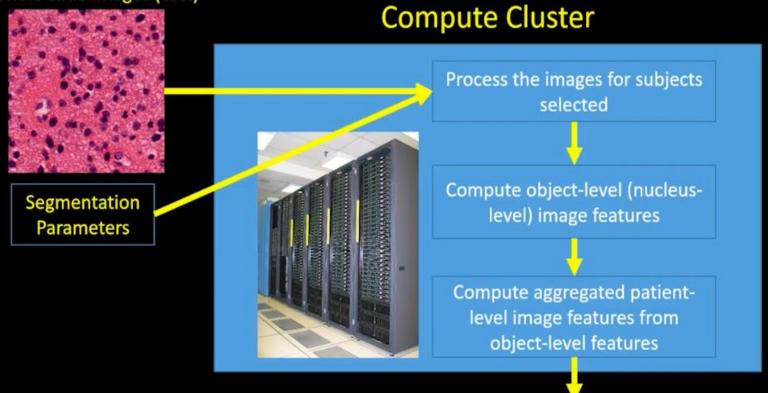


## **Cell Morphometry Features**



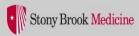


#### Whole Slide Images (WSI)



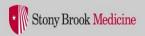
## **FeatureDB**

Load object-level imaging features and segmentation results
 Load patient-level imaging features along with a selected subset of clinical and genomic data (e.g. gene mutations, days to death, vital status)

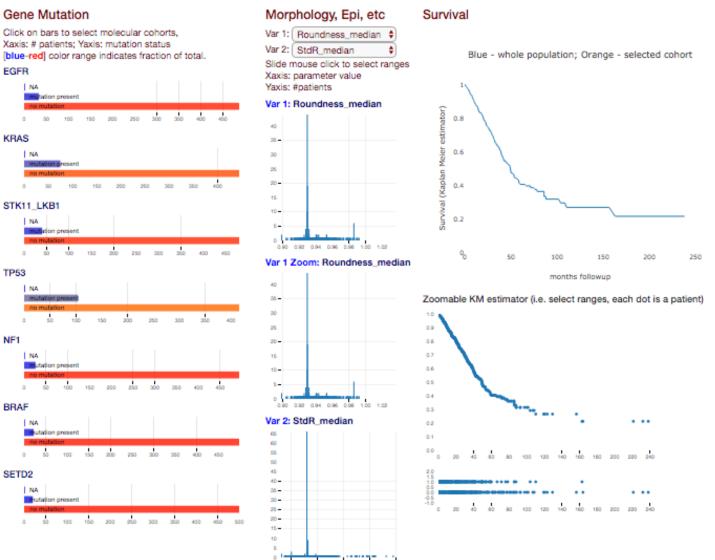


# FeatureExplorer

Explore relationships between patient-level imaging features and gene mutations and survival patterns.



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

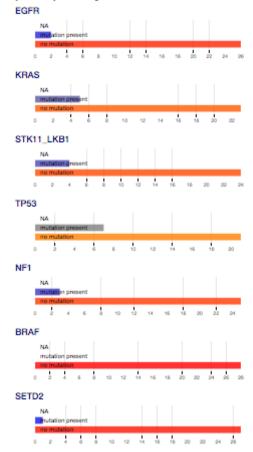


100 120 140 160 180 200 220 240

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

#### Gene Mutation

Click on bars to select molecular cohorts, Xaxis: # patients; Yaxis: mutation status [blue-red] color range indicates fraction of total.



0.0

100 120 140

#### Morphology, Epi, etc Survival Var 1: Roundness median \$ 0 B Q 🕂 B B X # 🖛 🚍 🥼 Var 2: StdR\_median ۵ Blue - whole population; Orange - selected cohort Slide mouse click to select ranges Xaxis: parameter value Yaxis: #patients 1 0.9638548 trace 0 Var 1: Roundness median estimator) 2.0 0.8 1.0 1.6 Meler 14 0.6 12 (Kaplan 0.4 o all 10 0.6 0.4 0.2 0.2 0.0 0.92 0.94 0.96 100 150 200 250 3.975359 Var 1 Zoom: Roundness\_median months followup 0.9 0.8 Zoomable KM estimator (i.e. select ranges, each dot is a patient) 6.7 1.0 0.6 0.9 • 0.5 0.6 0.4 0.7 0.5 0.6 0.5 0.1 0.4 0.0 0.90 0.92 0.94 0.96 0.96 1.00 1.02 Var 2: StdR median 6.2 2.0 0.5 1.8 0.0 43 43 40 100 120 140 160 180 200 220 240 1.6 14 1.2 .... 1.0 100 120 140 160 180 200 220 240 0.8 -0.6 0.4 0.2

# Collaboration with MGH – Feature Explorer – Radiology Brain MR/Pathology Features

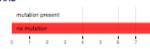


### Collaboration with SBU Radiology – TCGA NSCLC Adeno Carcinoma Integrative Radiology, Pathology, "omics", outcome

#### Gene Mutation

Click on bars to select molecular cohorts, Xaxis: # patients; Yaxis: mutation status [blue-red] color range indicates fraction of tota EGFR





#### PTEN

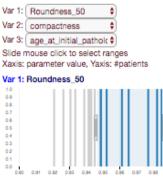
ne mutation

no mutation

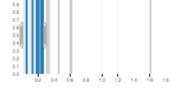
#### TP53

mutation present no mutation 00 05 1.0 1.5 2.0 2.5 3.0 3.5 4.0

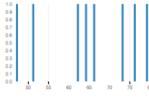
#### Rad, Path, Epi, etc



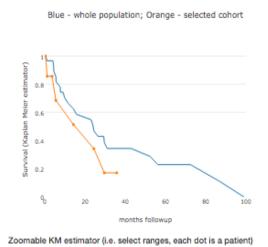


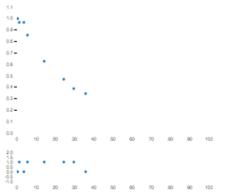


Var 3: age\_at\_initial\_pathologic\_diagnosis



#### Survival

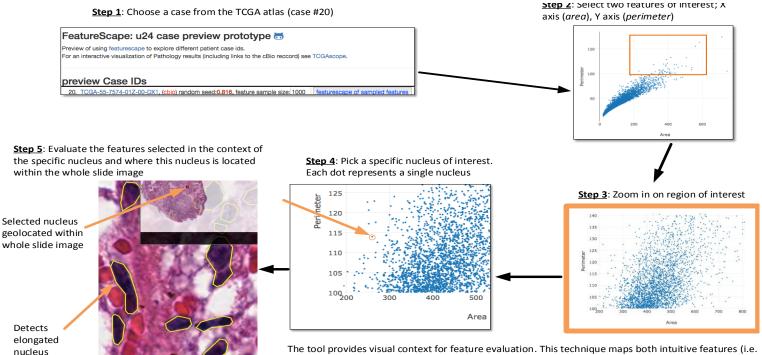




Mary Saltz, Mark Schweitzer SBU Radiology

## Feature Scape

## **Relationship Between Image and Features**



The tool provides visual context for feature evaluation. This technique maps both intuitive features (i.e. size, shape, color) and non-intuitive features (i.e. wavelets, texture) to the ground truth of source images through an interactive web-based user interface.



#### FeatureScape: u24 case preview prototype 🐱

Preview of using featurescape to explore different patient case ids. For an interactive visualization of Pathology results (including links to the cBio reccord) see TCGAscope.

#### preview Case IDs

 TCGA-34-2605-01Z-00-DX1, (cbio) random seed: 0.559, feature sample size: 1000 2. TCGA-38-4625-01Z-00-DX1, (cbio) random seed:0.628, feature sample size: 1000 TCGA-38-4626-01Z-00-DX1, (cbio) random seed:0.700, feature sample size: 1000 TCGA-38-4628-01Z-00-DX1, (cbio) random seed:0.016, feature sample size: 1000 TCGA-38-4629-01Z-00-DX1, (cbio) random seed:0.185, feature sample size: 1000 TCGA-38-6178-01Z-00-DX1, (cbio) random seed:0.317, feature sample size: 1000 TCGA-38-A44F-01Z-00-DX1, (cbio) random seed:0.906, feature sample size: 1000 TCGA-50-5044-01Z-00-DX1, (cbio) random seed:0.055, feature sample size: 1000 TCGA-50-5045-01Z-00-DX1, (cbio) random seed: 0.946, feature sample size: 1000 TCGA-50-5045-01Z-00-DX2, (cbio) random seed: 0.551, feature sample size: 1000 11. TCGA-50-5055-01Z-00-DX1, (cbio) random seed:0.127, feature sample size: 1000 TCGA-34-5232-01Z-00-DX1, (cbio) random seed: 0.208, feature sample size: 1000 TCGA-50-5055-01Z-00-DX2, (cbio) random seed:0.321, feature sample size: 1000 TCGA-50-5066-01Z-00-DX1, (cbio) random seed: 0.711, feature sample size: 1000 TCGA-50-5066-02Z-00-DX1, (cbio) random seed:0.008, feature sample size: 1000 TCGA-50-5942-01Z-00-DX1, (cbio) random seed:0.031, feature sample size: 1000 17. TCGA-50-5946-01Z-00-DX1, (cbio) random seed: 0.768, feature sample size: 1000 TCGA-50-6590-01Z-00-DX1, (cbio) random seed: 0.668, feature sample size: 1000 TCGA-50-6591-01Z-00-DX1, (cbio) random seed:0.498, feature sample size: 1000

featurescape of sampled features featurescape of sampled features



### FeatureScape 5 + Preliminary demo of integrative use of multiple FeatureScape tools

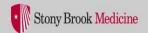
1000 entries sampled from https://tahsin175.informatics.stonybrook.edu:4500/?limit=1000&find={%22randval%22: {%22\$gte%22:0.149},%22provenance.analysis\_execution\_id%22:%22lung-features-v4%22,%22image.caseid%22:%22TCGA-38-4628-01Z-00-DX1%22}

+ Load Data

#### Cross-tabulated feature correlations

FeretDiameter	000000000000000000000000000000000000000
MeanR	000000000000000000000000000000000000000
MeanG	000000000000000000000000000000000000000
MeanB	000000000000000000000000000000000000000
StdR	000000000000000000000000000000000000000
StdG	000000000000000000000000000000000000000
StdB	000000000000000000000000000000000000000
EquivalentSphericalRadius	000000000000000000000000000000000000000
EquivalentEllipsoidDiameter0	000000000000000000000000000000000000000
EquivalentEllipsoidDiameter1	000000000000000000000000000000000000000
Perimeter	000000000000000000000000000000000000000
EquivalentSphericalPerimete	r0000000000000000000000000000000000000
PrincipalMoments0	000000000000000000000000000000000000000
PhysicalSize	000000000000000000000000000000000000000
PrincipalMoments1	000000000000000000000000000000000000000
Area	000000000000000000000000000000000000000
NumberOfPixels	000000000000000000000000000000000000000
NumberOfPixelsOnBorder	000000000000000000000000000000000000000
Roundness	000000000000000000000000000000000000000
Elongation	000000000000000000000000000000000000000
Flatness	000000000000000000000000000000000000000

#### (click on symbols for densities)



#### FeatureScape 5 + Preliminary demo of integrative use of multiple FeatureScape tools

1000 entries sampled from https://tahsin175.informatics.stonybrook.edu:4500/?limit=1000&find={%22randval%22: {%22\$gte%22:0.149},%22provenance.analysis\_execution\_id%22:%22lung-features-v4%22,%22image.caseid%22:%22TCGA-38-4628-01Z-00-DX1%22}

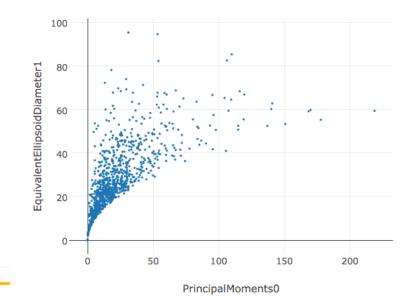
+ Load Data

#### Cross-tabulated feature correlations

FeretDiameter	000000000000000000000000000000000000000
MeanR	0 000 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
MeanG	0 <b>000</b> 00000000000000000000000000000000
MeanB	0 <b>000</b> 00000000000000000000000000000000
StdR	000000000000000000000000000000000000000
StdG	000000000000000000000000000000000000000
StdB	000000000000000000000000000000000000000
EquivalentSphericalRadius	000000000000000000000000000000000000000
EquivalentEllipsoidDiameter	)0000000 <mark>00</mark> 000000000000000000000000000
EquivalentEllipsoidDiameter	00000000000000000000000000000000000000
Perimeter	000000000000000000000000000000000000000
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PrincipalMoments0	000000000000000000000000000000000000000
PhysicalSize	000000000000000000000000000000000000000
PrincipalMoments1	000000000000000000000000000000000000000
Area	00000000000000000000000
NumberOfPixels	000000000000000000000000000000000000000
NumberOfPixelsOnBorder	000000000000000000000000000000000000000
Roundness	000000000000000000000000000000000000000
Elongation	000000000000000000000000000000000000000
Flatness	000000000000000000000000000000000000000

#### Pearson correlation between

- PrincipalMoments0
- EquivalentEllipsoidDiameter1



Resample from selected region (under development)



#### FeatureScape 5 + Preliminary demo of integrative use of multiple FeatureScape tools

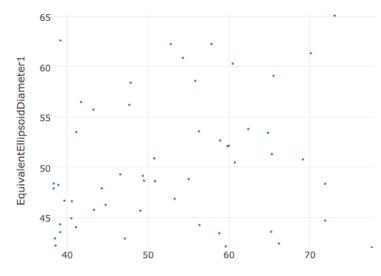
1000 entries sampled from https://tahsin175.informatics.stonybrook.edu:4500/?limit=1000&find={%22randval%22: {%22\$gte%22:0.149},%22provenance.analysis\_execution\_id%22:%22lung-features-v4%22,%22image.caseid%22:%22TCGA-38-4628-01Z-00-DX1%22}

+ Load Data

#### Cross-tabulated feature correlations

FeretDiameter	000000000000000000000000000000000000000
MeanR	000000000000000000000000000000000000000
MeanG	000000000000000000000000000000000000000
MeanB	000000000000000000000000000000000000000
StdR	000000000000000000000000000000000000000
StdG	000000000000000000000000000000000000000
StdB	000000000000000000000000000000000000000
EquivalentSphericalRadius	000000000000000000000000000000000000000
EquivalentEllipsoidDiameter0	000000000000000000000000000000000000000
EquivalentEllipsoidDiameter1	00000000000000000000000000000000000000
Perimeter	000000000000000000000000000000000000000
EquivalentSphericalPerimeter	rooooooo00000000000000
PrincipalMoments0	000000000000000000000000000000000000000
PhysicalSize	000000000000000000000000000000000000000
PrincipalMoments1	000000000000000000000000000000000000000
Area	000000000000000000000000000000000000000
NumberOfPixels	000000000000000000000000000000000000000
NumberOfPixelsOnBorder	000000000000000000000000000000000000000
Roundness	000000000000000000000000000000000000000
Elongation	000000000000000000000000000000000000000
Flatness	000000000000000000000000000000000000000

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PrincipalMoments0

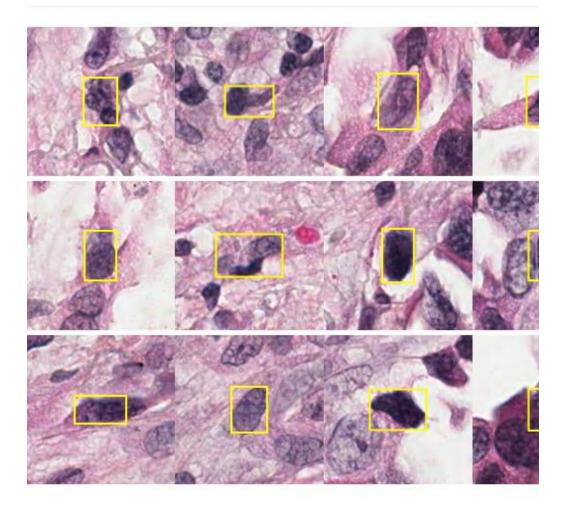
Resample from selected region (under development)

#### Pearson correlation between

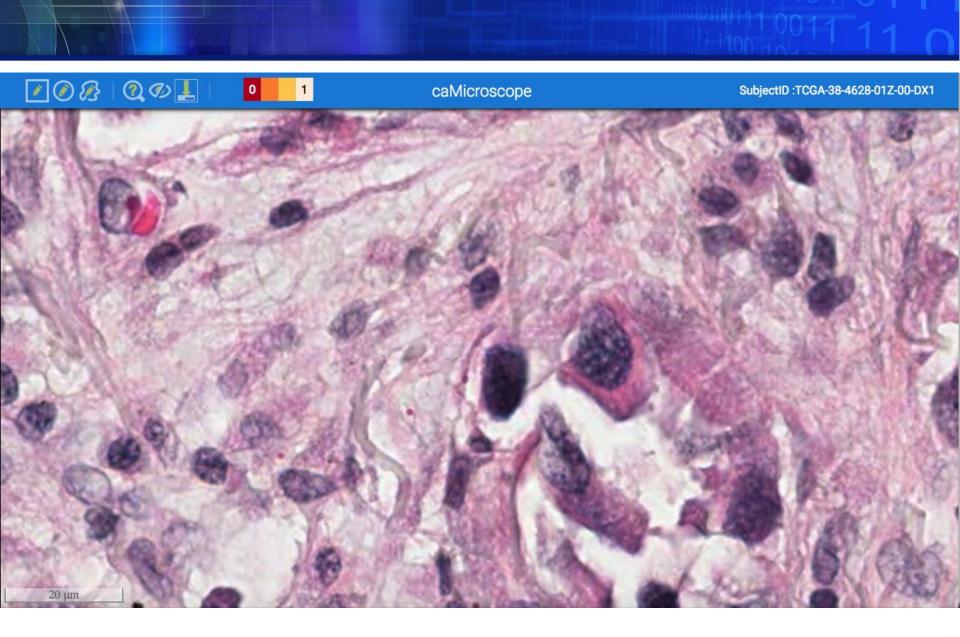
- PrincipalMoments0EquivalentEllipsoidDiameter1

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Hover over the smaller patch containing a nucleus to view it larger. The larger patch underneath shows the nucleus in context. Click on either patch to go to the location in caMicroscope.





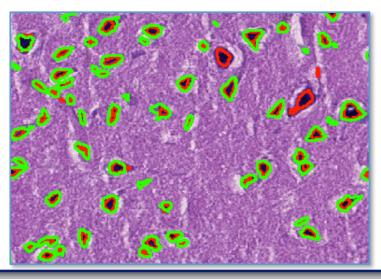




## Algorithm Comparison, Validation, Uncertainty Quantification

- High quality image analysis algorithms are essential to support biomedical research and diagnosis
  - Validate algorithms with human annotations
  - Compare and consolidate different algorithm results

e.g.: what are the distances and overlap ratios between markup boundaries from two algorithms?



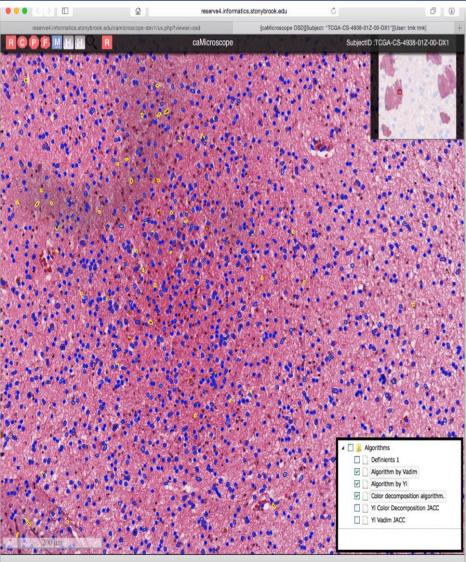
Cross matching of two spatial data sets

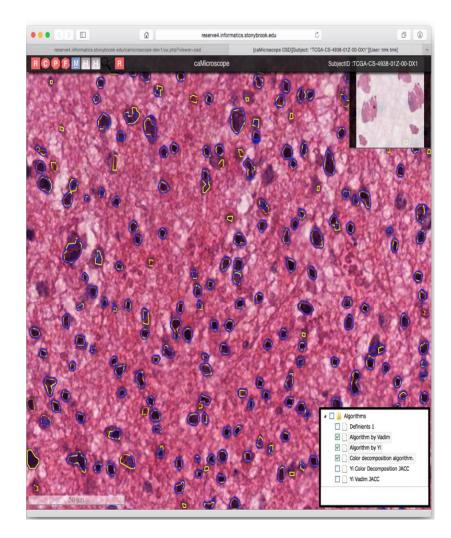
Green: algorithm 1 Red: algorithm 2





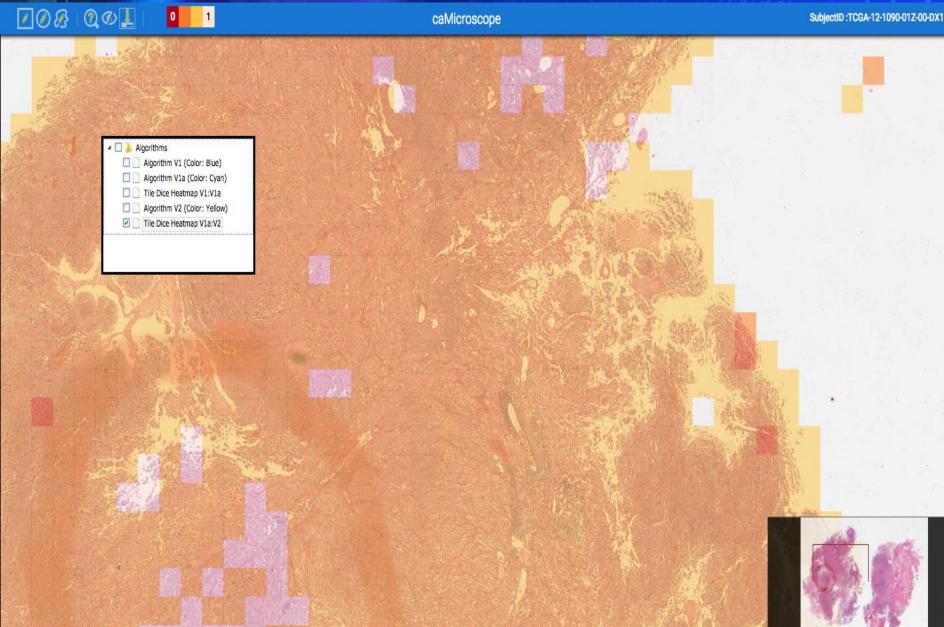
## caMicroscope/MongoDB - Multiple Algorithm Comparison; Generate and Curate Pathomics Feature set







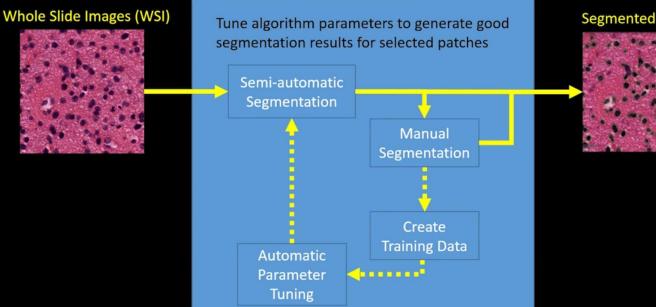
## Heatmap – Depicts Agreement Between Algorithms

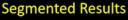


## **3D Slicer Pathology – Generate High Quality Ground Truth**

ITCR - Tools to Analyze Morphology and Spatially Mapped Molecular Data

### **3D Slicer Pathology**



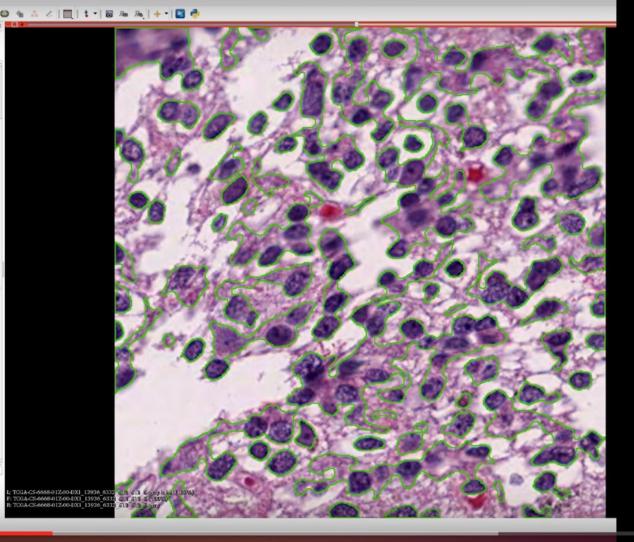




## **Apply Segmentation Algorithm**

## ITCR - Tools to Analyze Morphology and Spatially Mapped Molecular Data

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Create and Selec	t Label Maps				
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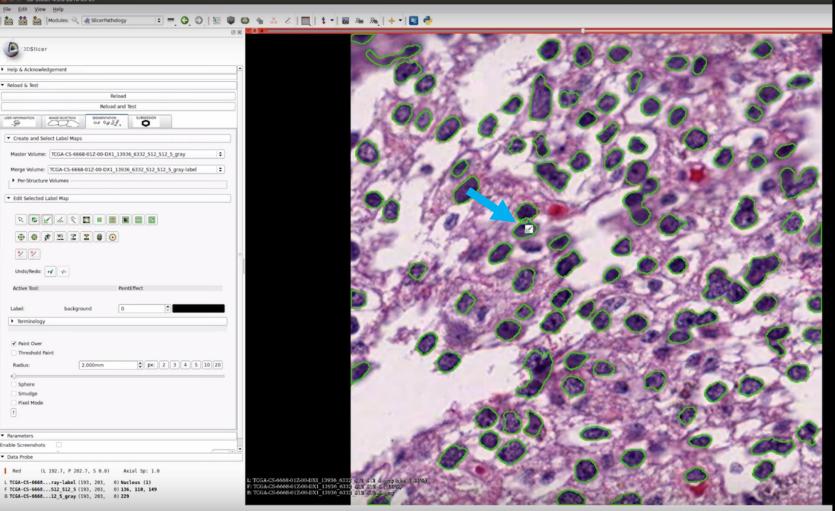


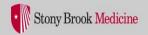
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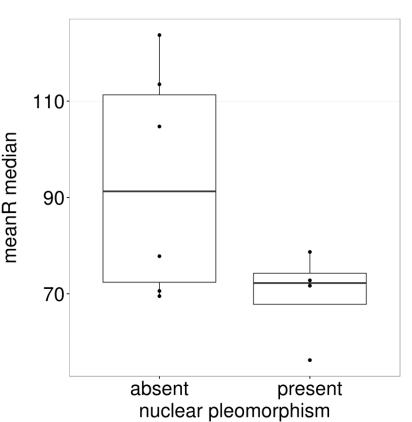
## Adjust algorithm parameters, manual fine tuning

### ITCR - Tools to Analyze Morphology and Spatially Mapped Molecular Data





## Integrated feature, text extraction



#### GBM – Text/Image Features

## **Collaboration with TIES group**

- 1. Docker based Segmentation/FeatureDB
- 2. TIES plus text feature extractor
- 3. SBU ITCR + Balasubramanian, Lynn

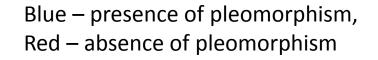
## Relate text features to image features

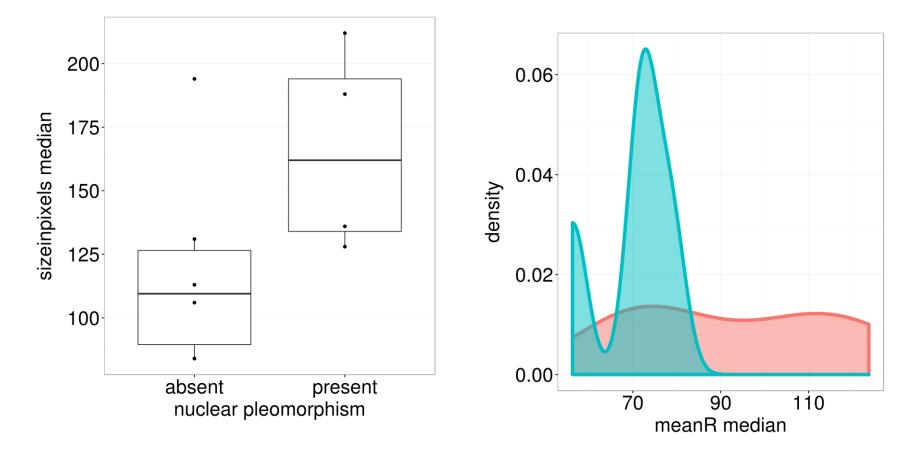
- 1. Demonstrate relationships
- 2. Create integrated human/algorithm feature metadata
- 3. Relate to "omics", Radiology, outcome
- 4. In silico classifications

Dockers Run at U Pittsburgh; extracted data analyzed at Stony Brook



## Relationship between Text Nuclear Pleomorphism Feature and Imaging Features

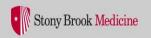






## Dissemination

- http://quip1.bmi.stonybrook.edu
- Containers
- Cloud
- 3D Slicer Pathology
- TCIA Integration between Radiology, Pathology images and features
- TCIA plugins will fetch GDC data for particular TCIA subject
- HPC via NSF and DOE
- TCGA PanCanAtlas Joint with Cooper and Gutman -Lymphocyte characterization
- Integrated Features/NLP joint with TIES



# Funding – Thanks!

- This work was supported in part by U24CA180924-01, NCIP/Leidos 14X138 and HHSN261200800001E from the NCI; R01LM011119-01 and R01LM009239 from the NLM
- This research used resources provided by the National Science Foundation XSEDE Science Gateways program under grant TG-ASC130023 and the Keeneland Computing Facility at the Georgia Institute of Technology, which is supported by the NSF under Contract OCI-0910735.



# Thanks!

Movie!

