

# TOOLS TO ANALYZE MORPHOLOGY AND SPATIALLY MAPPED MOLECULAR DATA

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

- Aim 1: Analysis pipelines for multi-scale, integrative analysis of histopathology images
- Aim 2: Database infrastructure to manage and query Pathomics features
- Aim 3: Software that targets local machines as well as cloud computing, and high performance computing systems
- Aim 4: Develop visualization middleware to relate Pathomics feature and image data and to integrate Pathomics image and "omic" data.
- Combine and deploy tools and methods in an integrated software platform

## **Using Tissue Slides**

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## • Traditional Approach

- Tissue taken from patients is fixed on a glass slide and stained
- Pathologist examines the tissue slide under high-power microscope to classify the tissue and render a diagnosis
- Hard (or impossible) to generate (detailed) quantitative information
- Labor intensive, not feasible for studies using large numbers of slides

## • Quantitative Analysis/Assessment

- Whole slide tissue imaging has progressed significantly in recent years
- Time required to scan a slide at high-resolution has reduced from multiple hours to several minutes
- Storage space is getting cheaper

## Whole Slide Tissue Imaging





# Quantitative Assessment: Nuclear Morphology



Regions of necrosis and angiogenesis



- Datasets of rich nuclear morphological information
- Some Research Questions:
  - Are there natural clusters of tumor morphology?
  - Are there links to patient outcome and molecular/genomic characteristics?

#### **Quantitative Assessment: Nuclear Features**



- Big Data Problem
  - Thousands of images
  - Millions of regions
  - Billions of nuclei and features



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#### Quantitative Assessment: Classification and Spatial Patternstrong Brook

- Spatial characterization mapping of patterns
- Tumor Infiltrating Lymphocytes by CNN Deep Learning Method
- Classification of images and image regions





### **Quantitative Imaging in Pathology (QuIP) Software Platform**

Software: https://github.com/SBU-BMI/quip\_distro.git Description: https://sbu-bmi.github.io/quip\_distro/

#### **QuIP Software Platform**



#### Visual Feature Analytics View (FeatureScape)



#### **Containerized Software for QuIP Services**



- The **application service group** is a single container that hosts a suite of Web applications to view images and interact with analysis results.
  - Visualization caMicroscope, FeatureScape
  - Higher-level Apps TIL/Heatmap Editor, Segmentation Curation App
- The **image analysis group** is made up of three containers, which collectively execute image analysis requests.
  - Analysis service hosts analysis pipeline
  - Job Manager service tracks jobs
  - *Image tile service* services image tiles for analysis
- The **data management service group** is implemented as a set of three containers for data loading, data management, and query processing.
  - Data loader service -- load image metadata and analysis results
  - Data manager service manage and index image data, analysis results, features
  - Feature query service query feature data for visualization and exploration

## **Image Analysis Libraries**



- Deep learning workflow for analysis of TIL patterns
  - A set of programs and scripts to train, re-train, execute prediction
  - <u>https://github.com/SBU-BMI/u24\_lymphocyte</u>
  - Graphical user interfaces via QuIP for review and refinement of prediction results and iterative learning
  - Docker container for prediction and training steps
- Nuclear segmentation pipeline
  - A C++ library and programs to carry out segmentation of nuclear material in whole slide tissue images
  - <u>https://github.com/SBU-BMI/pathomics\_analysis.git</u>
  - Integrated in QuIP as a Docker container

#### **QuIP Web Interface**





#### Visualization: caMicroscope



- Web-based platform for visualizing digital pathology images with segmentation results and features that are overlaid on the images
- Smooth web-based viewing of high resolution whole slide tissue images
  - Panning, Zooming
  - Overlay of one or more segmentation results
  - Human markups
- Provides APIs that allow interactive back-and-forth between feature exploration and image visualization

#### caMicroscope







#### **Visualization: FeatureScape**



- Web-based apps for feature based visual analytics
- Visualization of inter-related imaging features
- Go from feature level to image level

#### **FeatureScape**





#### caMicroscope + FeatureScape



sbu-bmi.github.io/featurescape/u24Preview.html

sbu-bmi.github.io/featurescape/?https://falcon.bmi.s...

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#### Nuclei Mugshots 👼

Mugshots of the "bad guys"... the bad nuclei, that is. Wanted: for robbery (of good health!)

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.





#### **Database: FeatureDB**



- Based on NoSQL document database
- Flexible data model in JSON format
  - Borrows elements from AIM and PAIS models
- Organizes segmentation and feature data into GeoJSON compatible format
  - Segmentation results as polygons
  - Features as key-value pairs
    - Shape, intensity, texture and size features

#### FeatureDB – GeoJSON-based Data Model



```
"type" : "Feature",
"parent_id" : "self",
"randval": 0.43199294805526733,
"object_type" : "nucleus",
"x" : 0.017050,
"y" : 0.358914,
"normalized" : "true",
"bbox" : [
    0.016993561910465638.
    0.35885439819583437,
    0.017108055978792176,
    0.35897366395246716
」,
"geometry" : {
    "type" : "Polygon",
    "coordinates" : [
             [0.017042001709342003,0.3589736521244049],
             [0.01703319512307644,0.3589736521244049],
             . . .
"footprint" : 93,
```

#### **FeatureDB – GeoJSON-based Data Model**

```
"properties" : {
    "scalar_features" : [
              "ns" : "http://u24.bmi.stonybrook.edu/v1",
             "nv" : [
                  {"name" : "b_cytoGradientMean","value" : 10.4891996383667},
ſ,
"provenance" : {
    "image" : {
         "case_id" : "0001-215633",
         "subject_id" : "0001-212"
    },
    "analysis" : {
         "execution_id" : "Mean_Shift_1.0",
         "study_id" : "small",
         "source" : "computer",
         "computation" : "segmentation"
    },
了,
"submit_date" : ISODate("2017-05-24T19:48:37.776Z")
```



## **Higher-level Web Applications**

- TIL/Heatmap Editor
  - Used in study of Tumor Infiltrating Lymphocytes (TILs)
- Segmentation Curation Application
  - Used in SEER Virtual Repository Pilot/Project





#### **TIL/Heatmap Editor**

- Support for Iterative Learning/Refinement Workflows
- Motivated by and used in study of TIL patterns in 13 TCGA cancer types



## Spatial Organization and Molecular Correlation of Tumor-Infiltrating Lymphocytes Using Deep Learning on Pathology Images

Joel Saltz,<sup>1,\*</sup> Rajarsi Gupta,<sup>1,4</sup> Le Hou,<sup>2</sup> Tahsin Kurc,<sup>1</sup> Pankaj Singh,<sup>3</sup> Vu Nguyen,<sup>2</sup> Dimitris Samaras,<sup>2</sup> Kenneth R. Shroyer,<sup>4</sup> Tianhao Zhao,<sup>4</sup> Rebecca Batiste,<sup>4</sup> John Van Arnam,<sup>5</sup> The Cancer Genome Atlas Research Network, Ilya Shmulevich,<sup>6</sup> Arvind U.K. Rao,<sup>3,7</sup> Alexander J. Lazar,<sup>8</sup> Ashish Sharma,<sup>9</sup> and Vésteinn Thorsson<sup>6,10,\*</sup>

http://www.cell.com/cell-reports/pdf/S2211-1247(18)30447-9.pdf

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

#### Importance of Immune System in Cancer Treatment and Prognosis

Tumor spatial context and cellular heterogeneity are important in cancer prognosis

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- 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



 Deep learning based analysis of patterns of tumor infiltrating lymphocytes (TILs)

•TIL patterns generated from 4,759 TCGA subjects (5,202 H&E slides), 13 cancer types

•TIL patterns correlate with pathologist eye and molecular estimates

•TIL patterns linked to tumor and immune molecular features, cancer type, and outcome





# TIL Patterns – Clustering Indices





#### **Training, Model Creation**



- 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



## Training, threshold adjustment, quality control





#### **Heatmap View of TIL Probability Maps**





#### **TIL Map Editor**





#### **TIL Map Editor**







### **Curation Tool for Nuclear Segmentation Results**

- Motivated by sensitivity of analysis methods to input parameters
- Used in SEER Virtual Repository Pilot/Project

## Analysis Sensitivity: Generating Imaging Feature Sets Story Brook

• Image analysis pipelines are sensitive to input parameters



#### **Generating Robust Feature Sets**

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- Run multiple analyses
- Store, index, interact with results
- Curation: Visual comparison and selection of results

#### **Segmentation Results Curation Tool**





# Methods and tools for integrating pathomics data into cancer registries - UG3CA225021 Saltz, Sharma, Foran and Durban

- Enhance SEER registry data with machine learning based classifications and quantitative pathomics feature sets.
- The New Jersey State Cancer Registry, Georgia and Kentucky State Cancer Registries
- Prostate Cancer, Lymphoma and NSCLC
- Repository of high-quality digitized pathology images for subjects whose data is being collected by the registries.
- Extract computational features and establish deep linkages with registry data, thus enabling the creation of information-rich, population cohorts containing objective imaging and clinical attributes

#### Cancer Imaging Archive – Integration of Pathology and Radiology for Community Clinical Studies





http://www.cancerimagingarchive.net/

#### TCIA sustainment and scalability Platforms for quantitative imaging informatics in precision medicine Prior, Saltz, Sharma -- U24CA215109-01

- Identify quantitative imaging phenotypes across scale through the use of Radiomic/Pathomic analyses
- Well-curated data for algorithm testing and validation.
- Integrative Radiology/Pathology Image-Omics studies
- Extend TCIA to support its rapidly growing user community and continue to promote research reproducibility and data reuse in cancer precision medical research.

## **Quantitative Imaging in Pathology (QuIP) Platform**



- Software: <a href="https://github.com/SBU-BMI/quip\_distro.git">https://github.com/SBU-BMI/quip\_distro.git</a>
- Containerized, service oriented
- Several key services are available
- Actively developed
  - More services and applications are being developed

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