

Big Data in Biomedical Imaging: Pathomics Analysis of Cancer

Tahsin Kurc

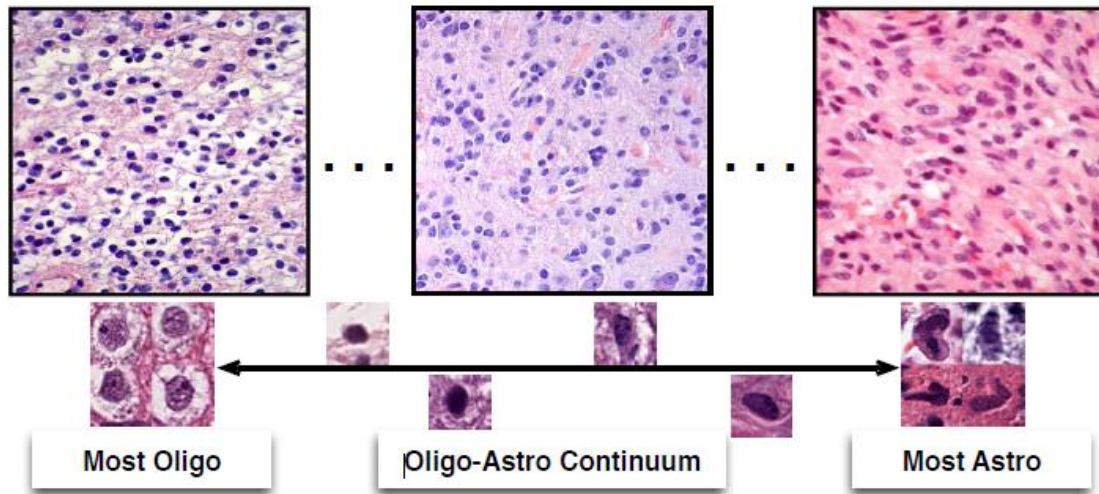
Biomedical Informatics Department
Stony Brook University

Team

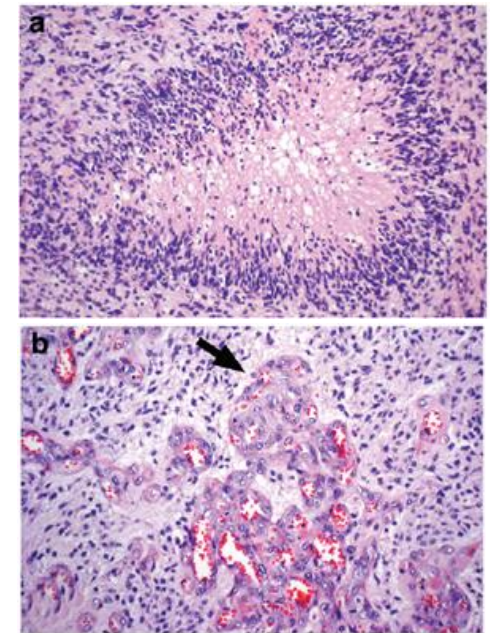
- Stony Brook:
 - Joel Saltz, Erich Bremer, Feiqiao Wang, Tammy Di Prima, Jonas Almeida, Mary Saltz, Le Hou, Vu Nyugen, Si Wen, Maozheng Zhao, Tianhao Zhao, Raj Gupta, Fusheng Wang, Furqan Baig, Yi Gao, Alina Jasniewski
- Emory University:
 - Ashish Sharma, Ganesh Iyer, Adam Marcus
- Oak Ridge National Laboratory:
 - Scott Klasky, Jeremy Logan, David Pugmire
- University of Brasilia, Brazil
 - George Teodoro, Luís F. R. Taveira, Alba C. M. A. Melo

Tissue Image Analysis

Different types of nuclei and cells



Regions of necrosis and angiogenesis

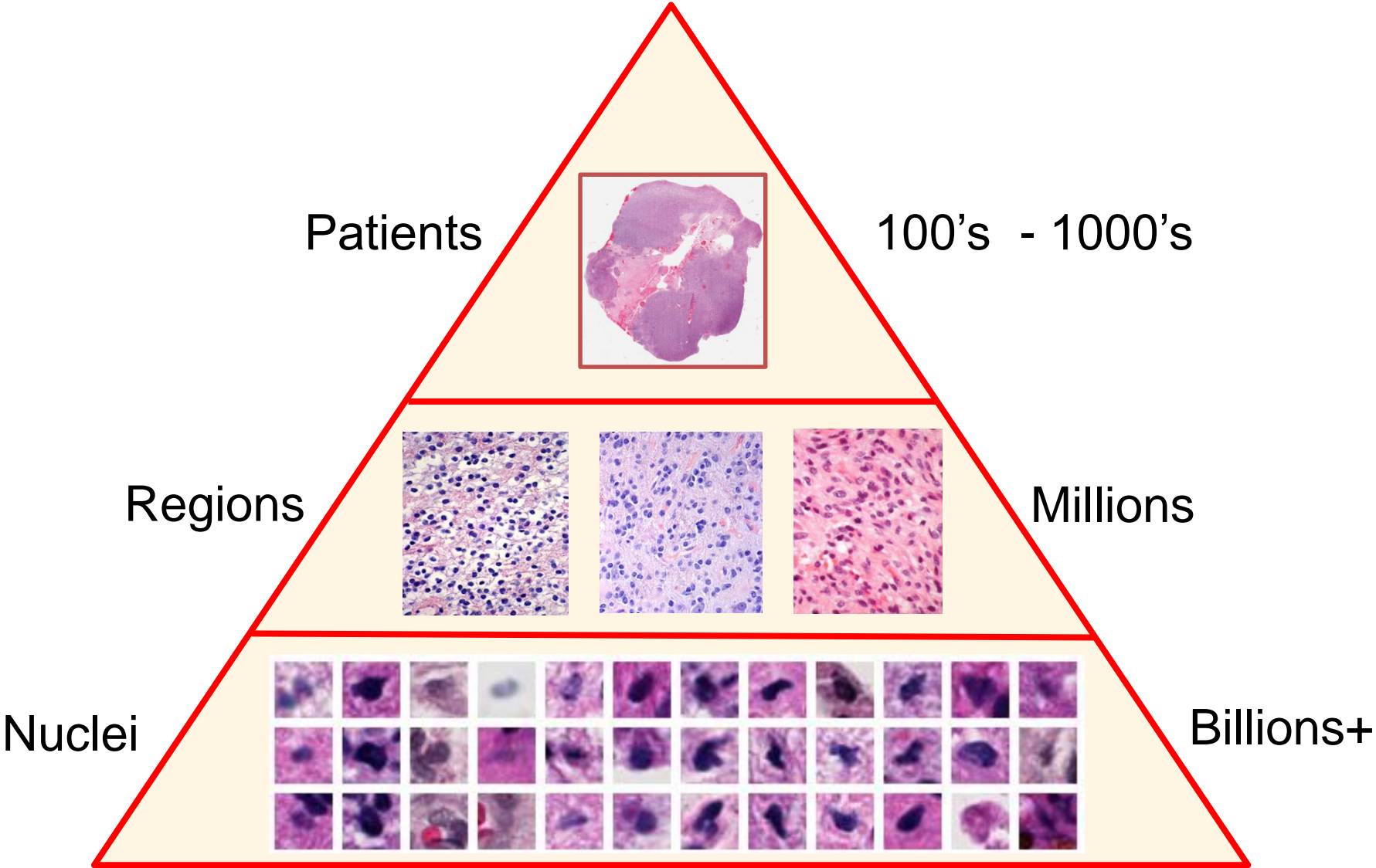


- Tumors are complex and heterogeneous
- Datasets of rich nuclear morphological information
 - Nuclear morphology
 - Maps of tumor infiltrating lymphocytes

Tissue Imaging

- Confluence of several technological advances is making slide scanning more practical
- Image scanning technology has progressed significantly in recent years
 - Sophisticated auto-focus mechanisms
 - Slide trays for batch scanning of slides
- Time required to scan a slide at high-resolution has reduced from multiple hours to several minutes
 - 200-300 slides / day
- Disk storage is getting cheaper

Big Data Challenges in Pathomics



TCGA Microscopy Image Data

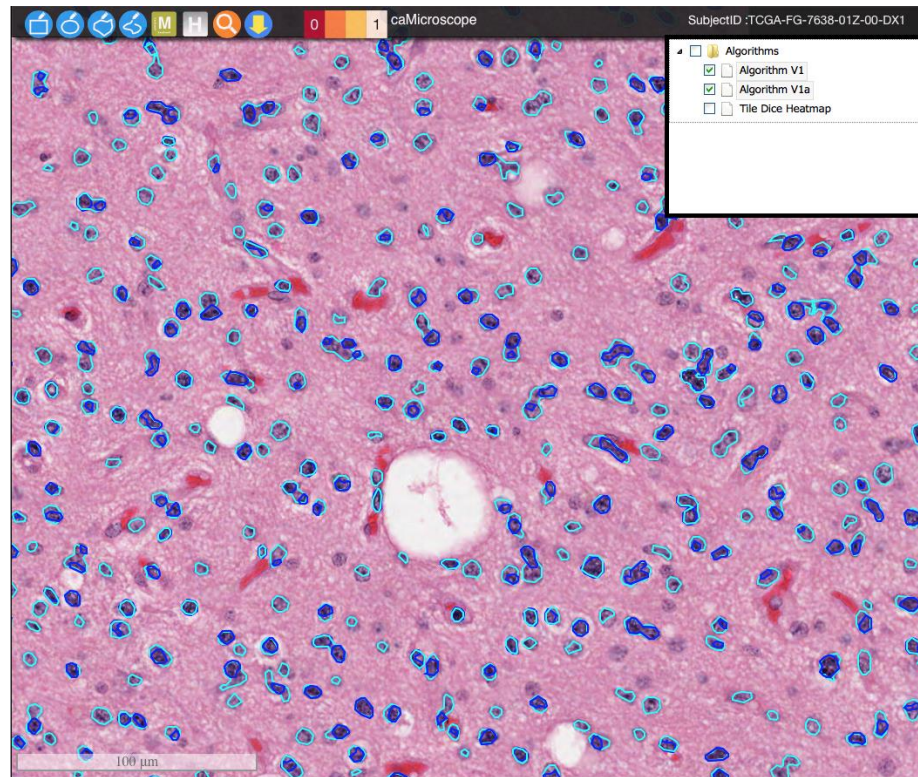
- Data from 11,000 subjects
- 30,000+ tissue slide images
- Image resolutions ranging from 4000x6000 pixels to 130,000x250,000 pixels
 - On average about **3.6 Billion** pixels per image

Big Data: Analyzing Whole Slide Tissue Images

- Needs large main memory
 - $130,000 \times 250,000 \times 3$ (RGB) = 90GB
- Takes from 30 minutes to 10-12 hours to process an image
 - A dataset with 100 images would require 8 days assuming 2 hours per image on average
- Hundreds of thousands to millions of objects in a whole slide tissue image

Analysis Sensitivity: Generating Robust Feature Sets

- Image analysis pipelines are sensitive to input parameters



Generating Robust Feature Sets

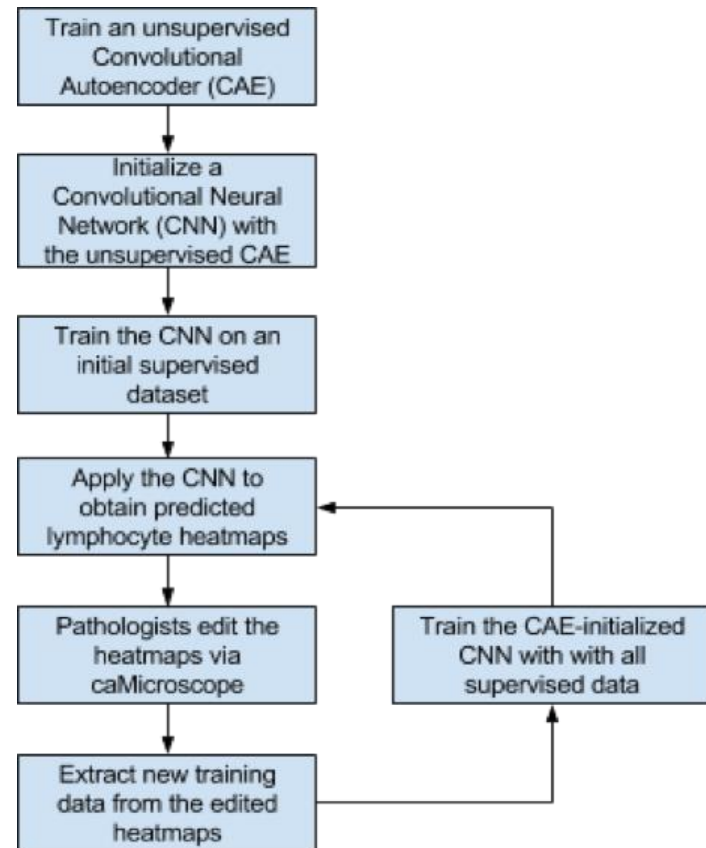
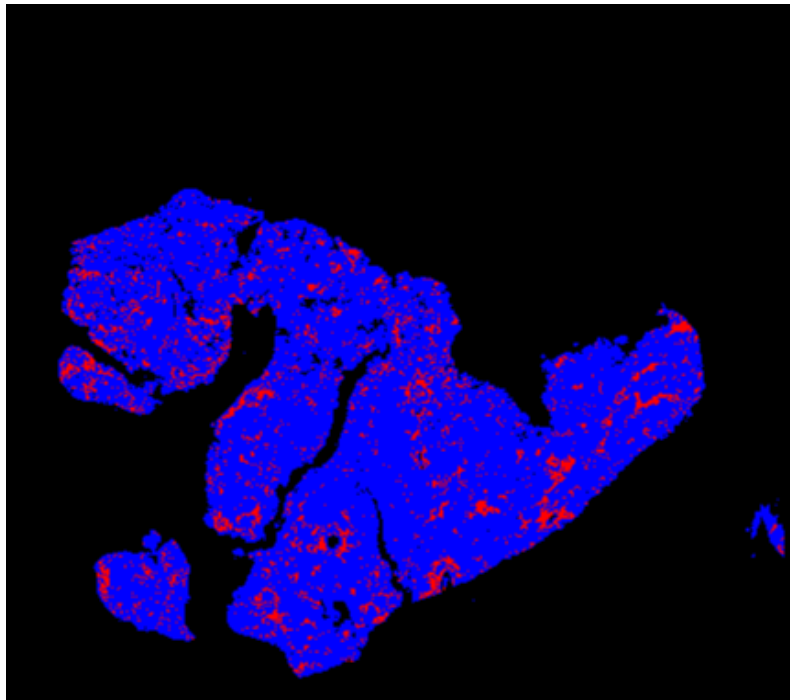
- Run multiple analyses
- Store, index, interact with results
- Computational comparison of results
 - Sensitivity analysis for algorithm evaluation and development
 - Parameter tuning
- Visual comparison of results
 - Curation

Feature Sets

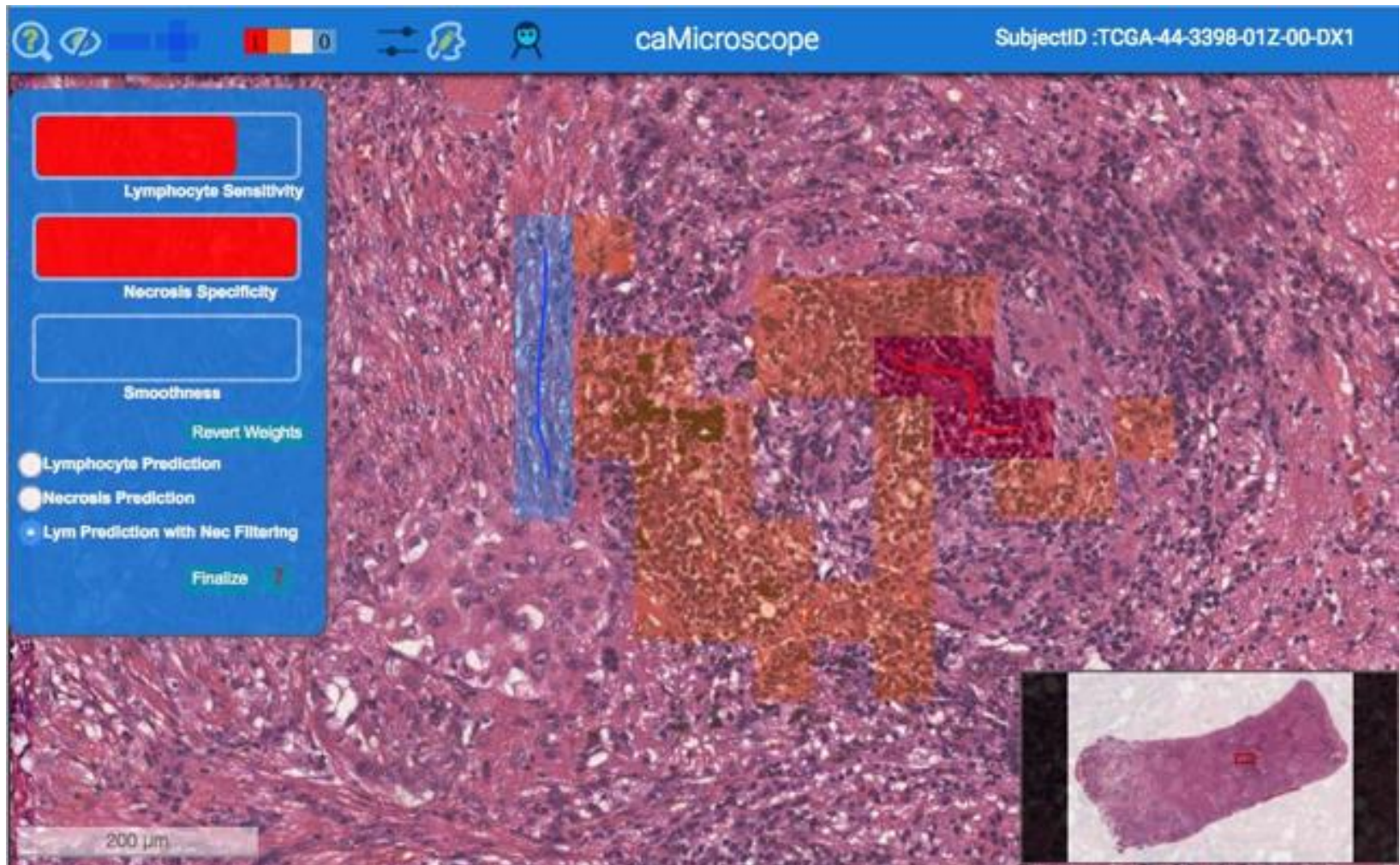
- 40-70 features per object (nucleus)
- Analysis of 4000 images
 - About 2 Billion segmented objects
- Multi-analysis of 300+ images
 - 6-10 analyses per image
 - About 2 Billion segmented objects

Deep Learning

- Tumor infiltrating lymphocytes

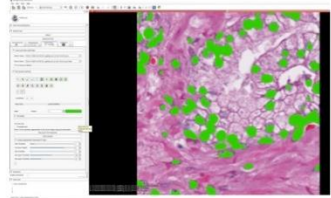


TIL Maps

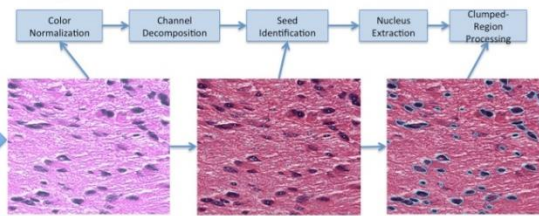


Pathomics Framework

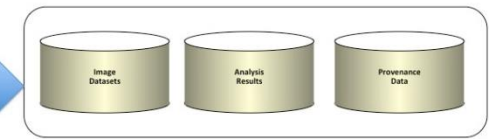
Parameter Adjustment Using 3D Slicer Pathology Extensions



Segmentation and Feature Computation Pipeline



Storage and Management of Pathology Imaging Features

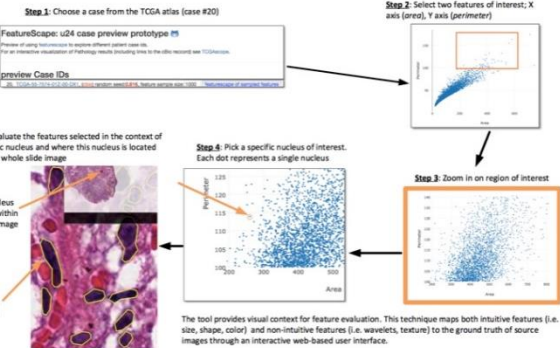


FeatureDB

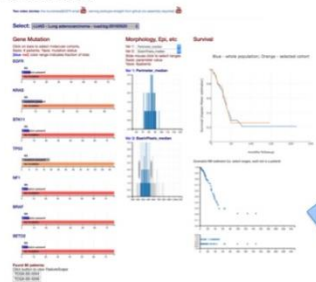
Visual Exploration of Features

Leveraging Visualization to Aid in Feature Management

- Going from the whole slide data set to selected features and back to the image
- Adding a visual perspective by using a live web-based interactive tool (<http://tsb-u-bmi.github.io/featurescape/u24/Preview.html>)



Feature Exploration in Multiple Dimensions – patient level aggregated imaging features with genomic data and survival information



Subset of Patients

Selected Patient or Image

Exploration of distribution of features for a subject or an image



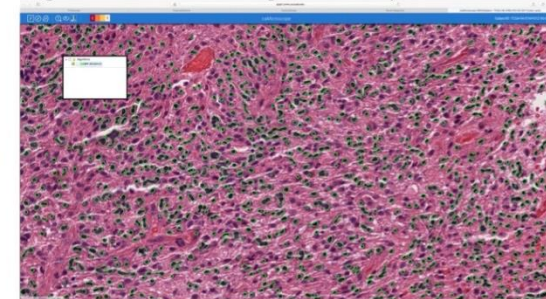
Input Image Dataset

Selection of subsets of Patients/Subjects based on clinical data

Visualization of sample set of segmented objects based on range queries on feature values



Visualization of whole slide tissue images and segmentation results using caMicroscope

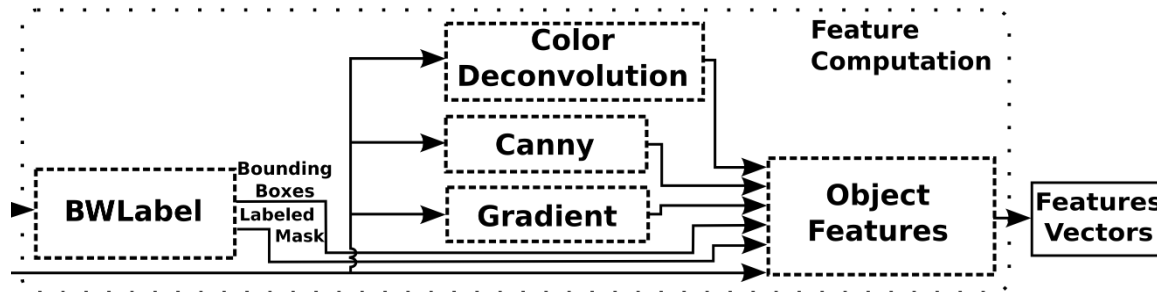
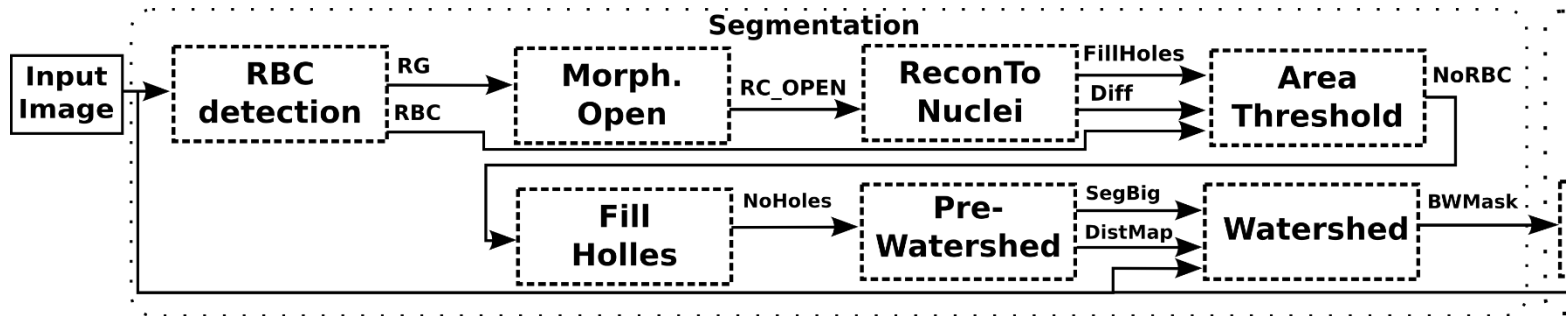
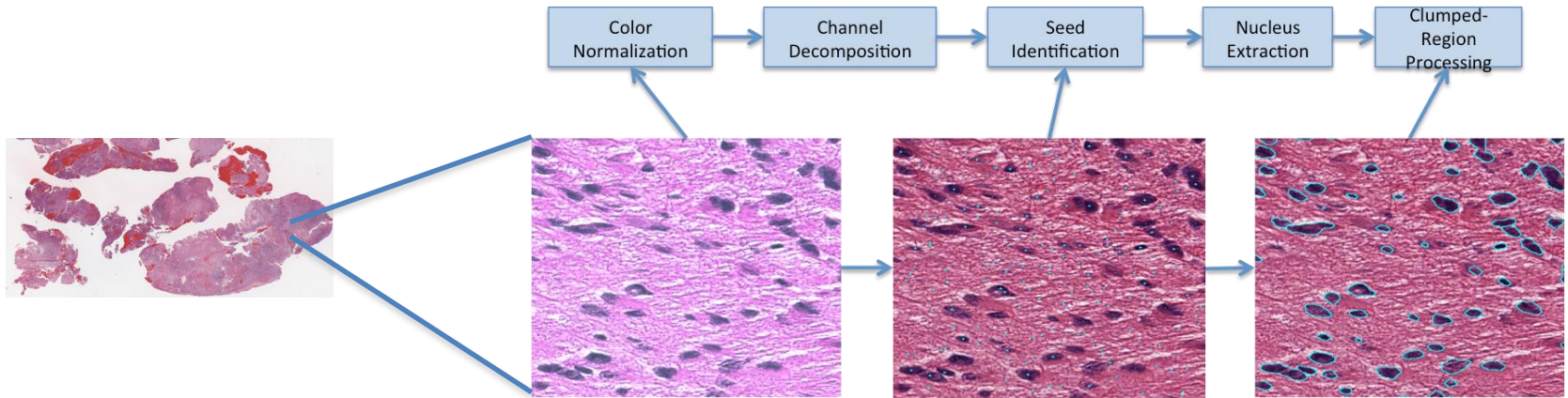


Leveraging Hybrid High Performance Computing Systems

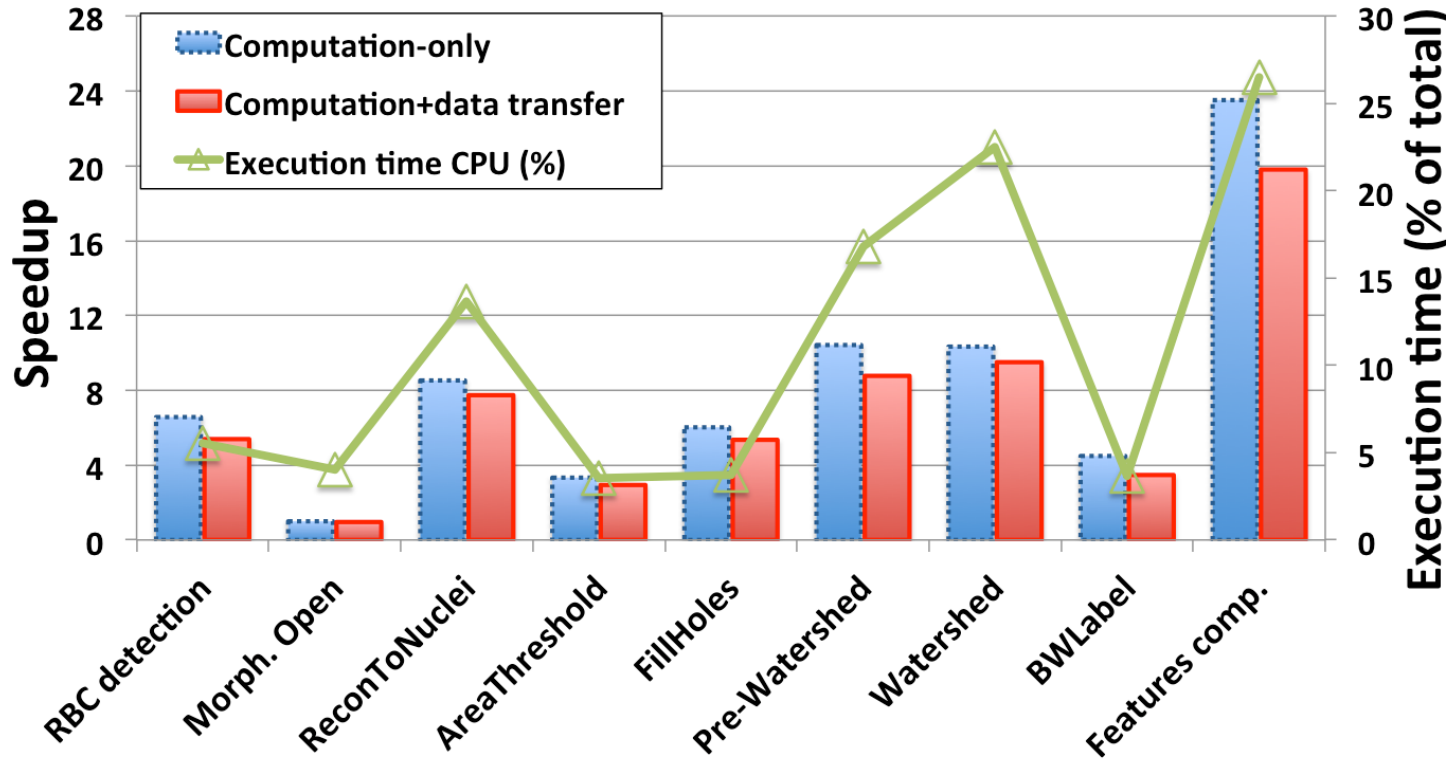
- Lots of nodes
- Multi-core CPUs, GPUs, and other hardware accelerators per node
- XSEDE, Supercomputing Centers



Nuclear Imaging Features

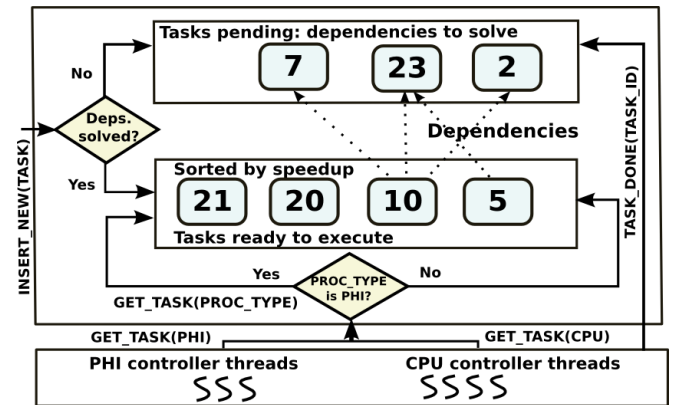
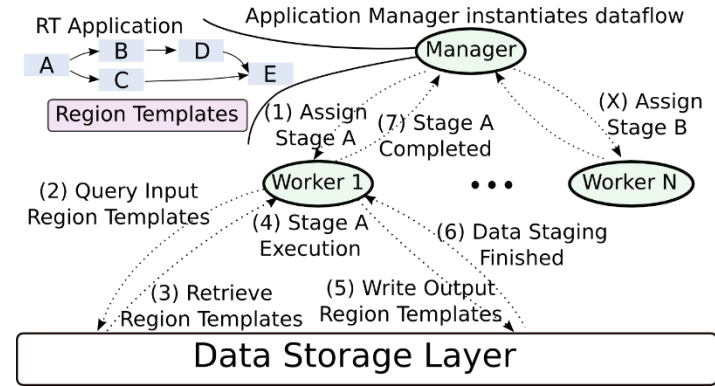
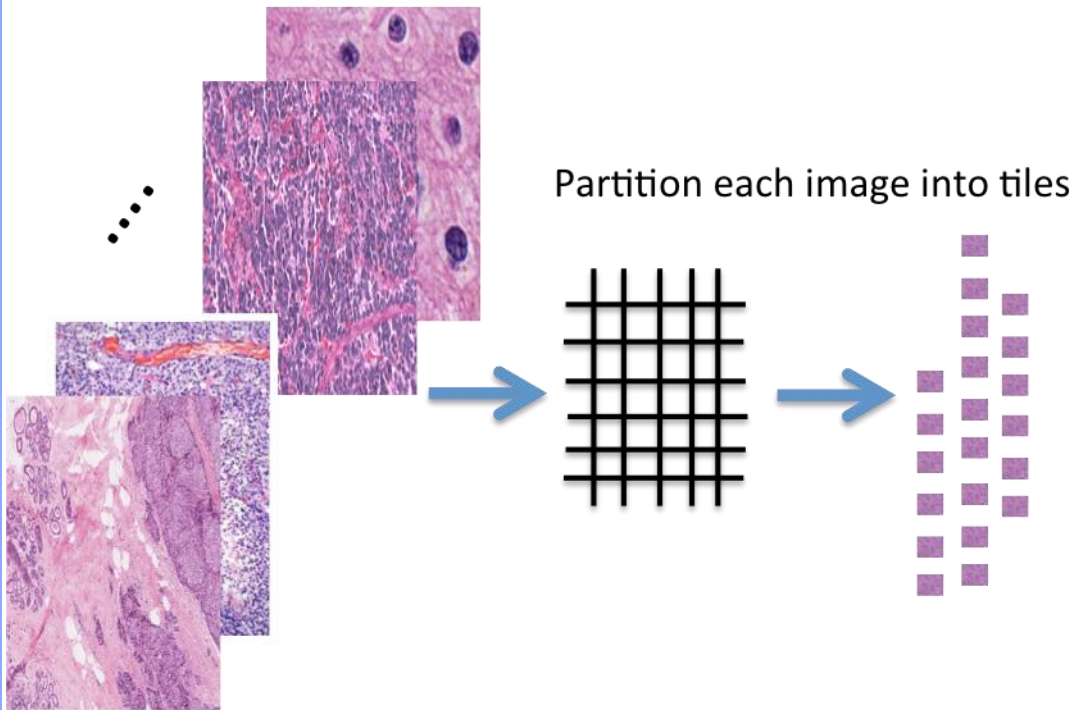


Inter-operation Performance Variations



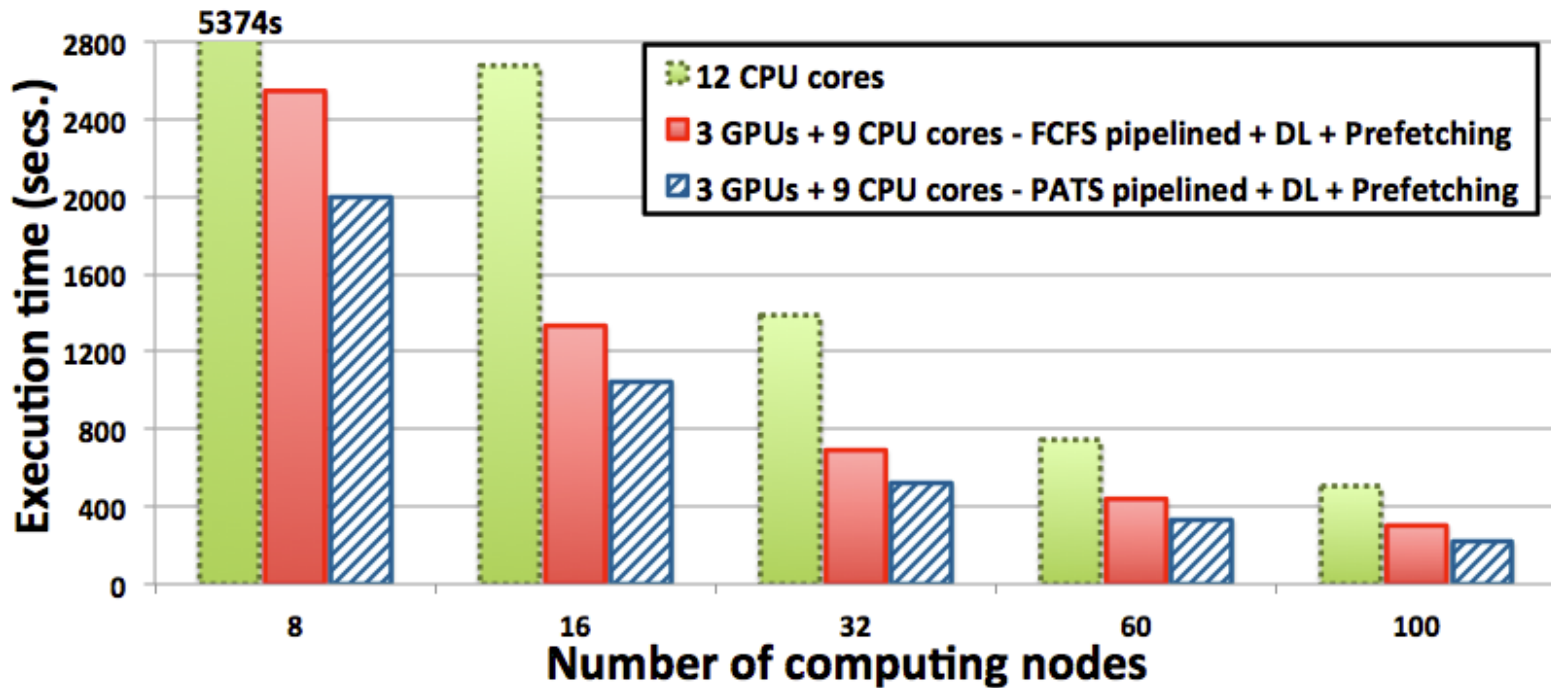
Experiments on Keeneland: equipped w/ M2090 GPUs

Analysis on HPC Platforms



- Thousands of CPU cores
- Hundreds of GPUs
- Large distributed memory

Coordinated Use of CPUs and GPUs

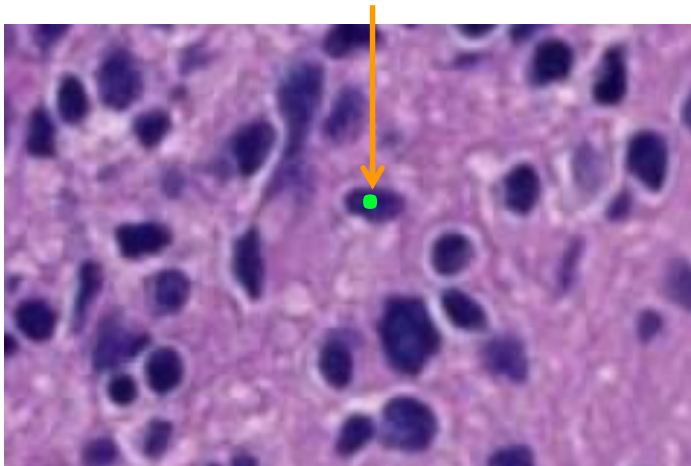


36,844 4Kx4K-pixel tiles from 340 whole slide tissue images.

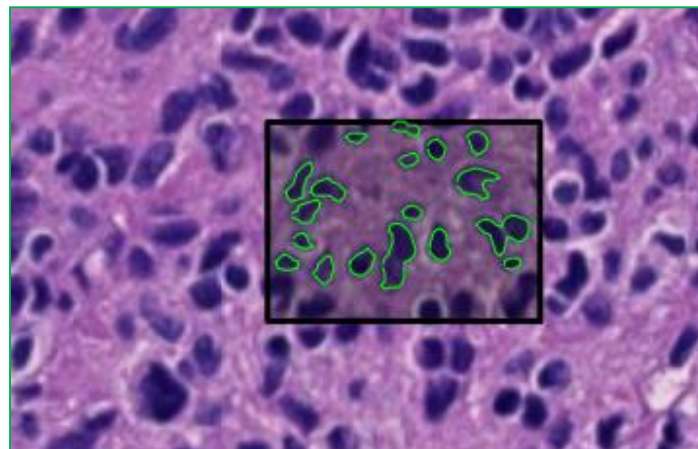
On 100 nodes, less than 4 minutes to process 36K tiles.

Support for Spatial Queries

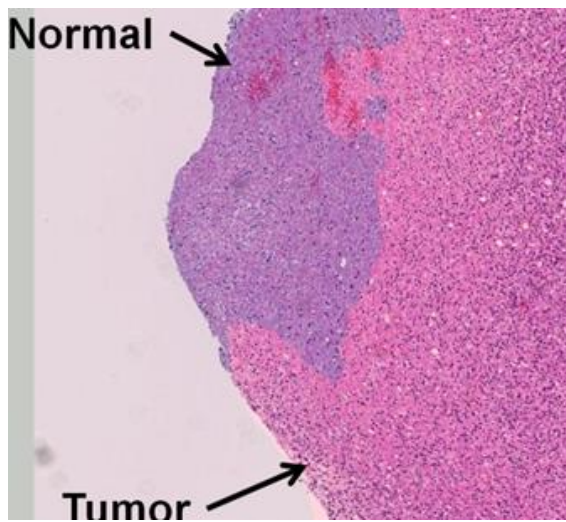
POINT query: human marked point inside a nucleus



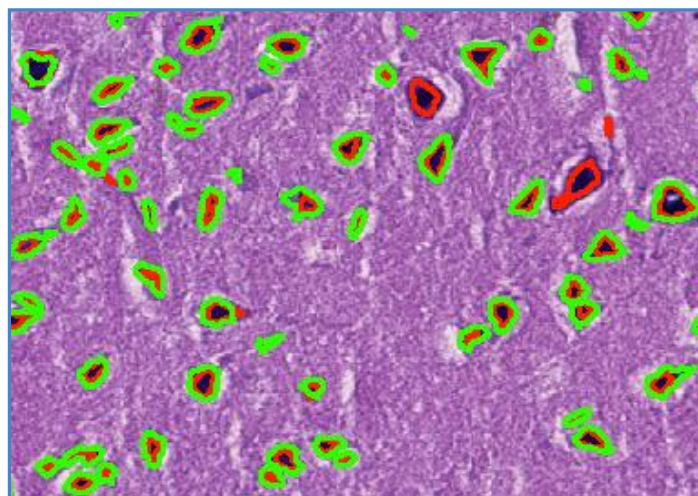
WINDOW query: return markups contained in a rectangle



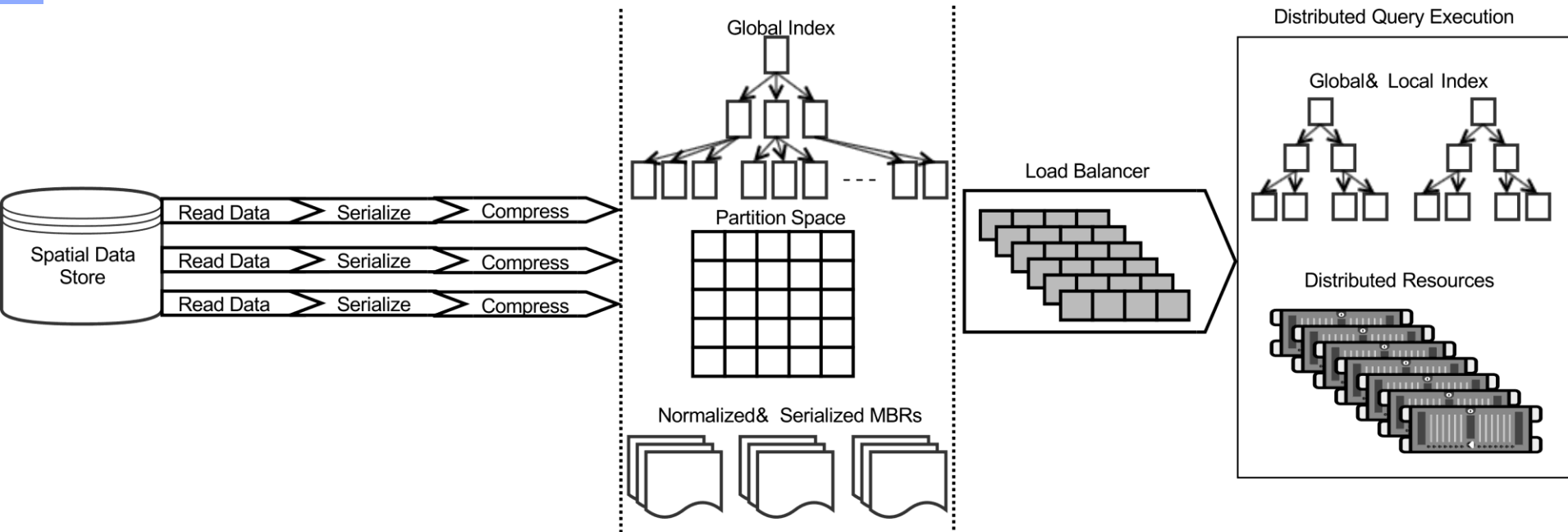
CONTAINMENT query: nuclear feature aggregation in tumor regions



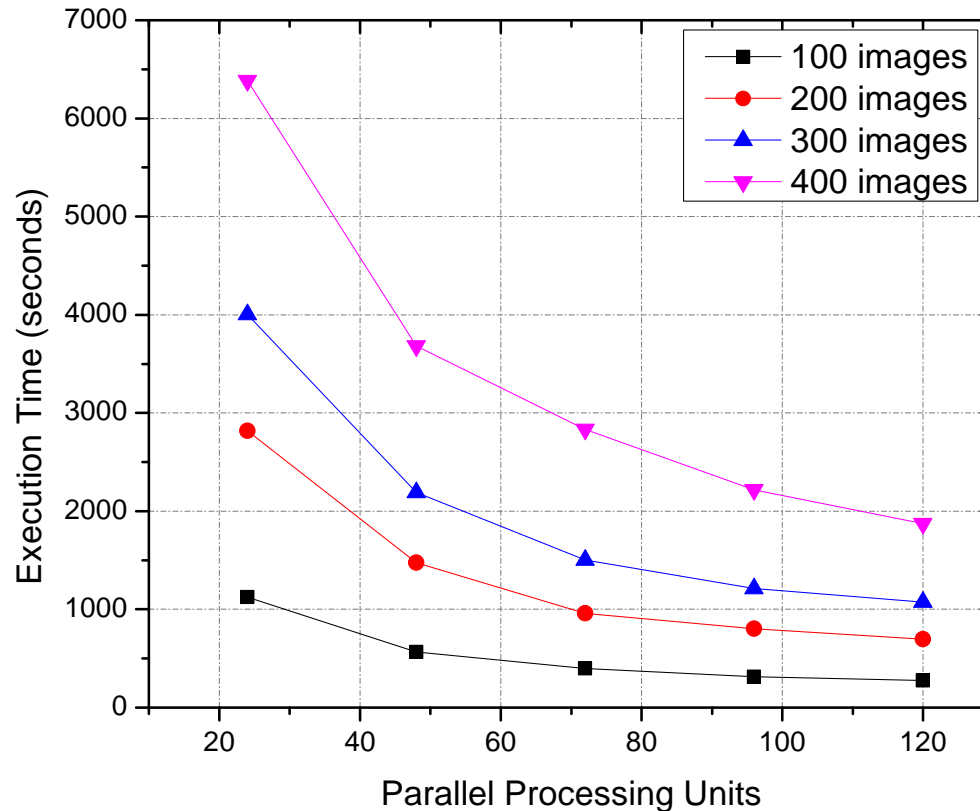
SPATIAL JOIN query: algorithm validation/comparison



SparkGIS



Heatmap Computations: Spatial Joins + Dice/Jaccard Metric

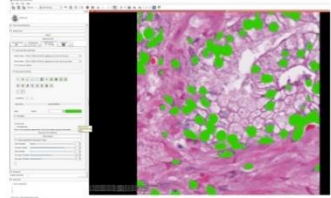


Going Forward

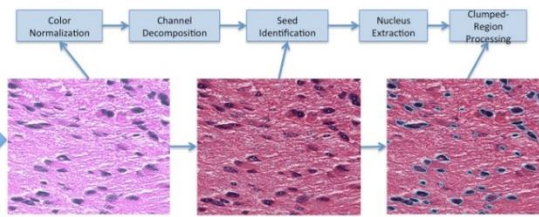
- Containers
 - Facilitates a modular design.
 - Self-contained, Isolated.
 - Software to support tissue image analysis needs to leverage a variety of existing libraries and tools.
 - Flexibility
 - Move computation to data
 - Move back and forth between Cloud platforms and local resources
- Scripting
 - JavaScript: Take advantage of web browsers
 - Python: Take advantage of large set of libraries

Pathomics Framework

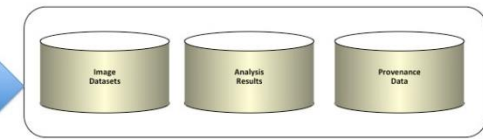
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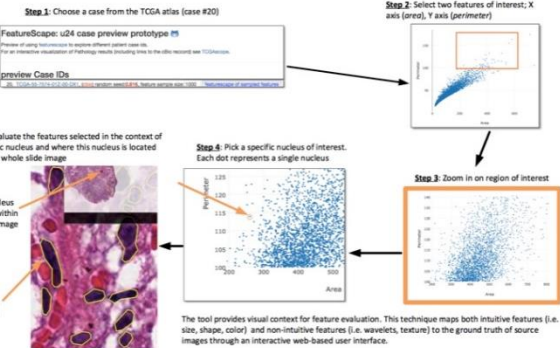


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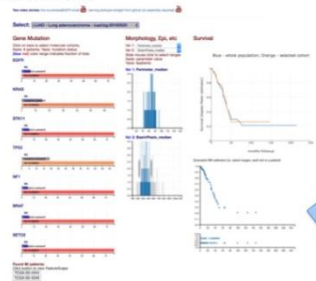
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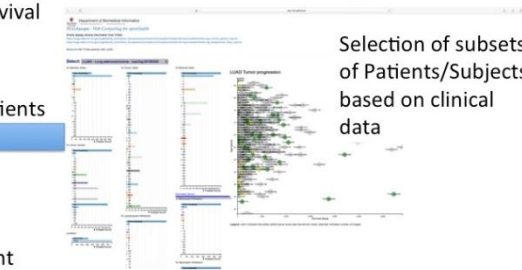
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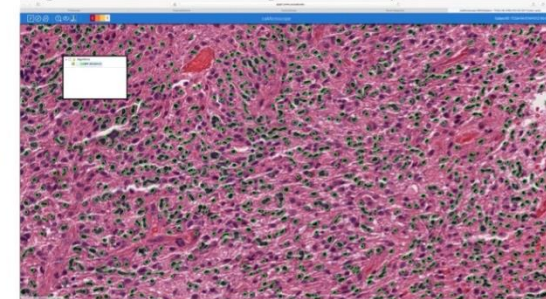


Selection of subsets of Patients/Subjects based on clinical data

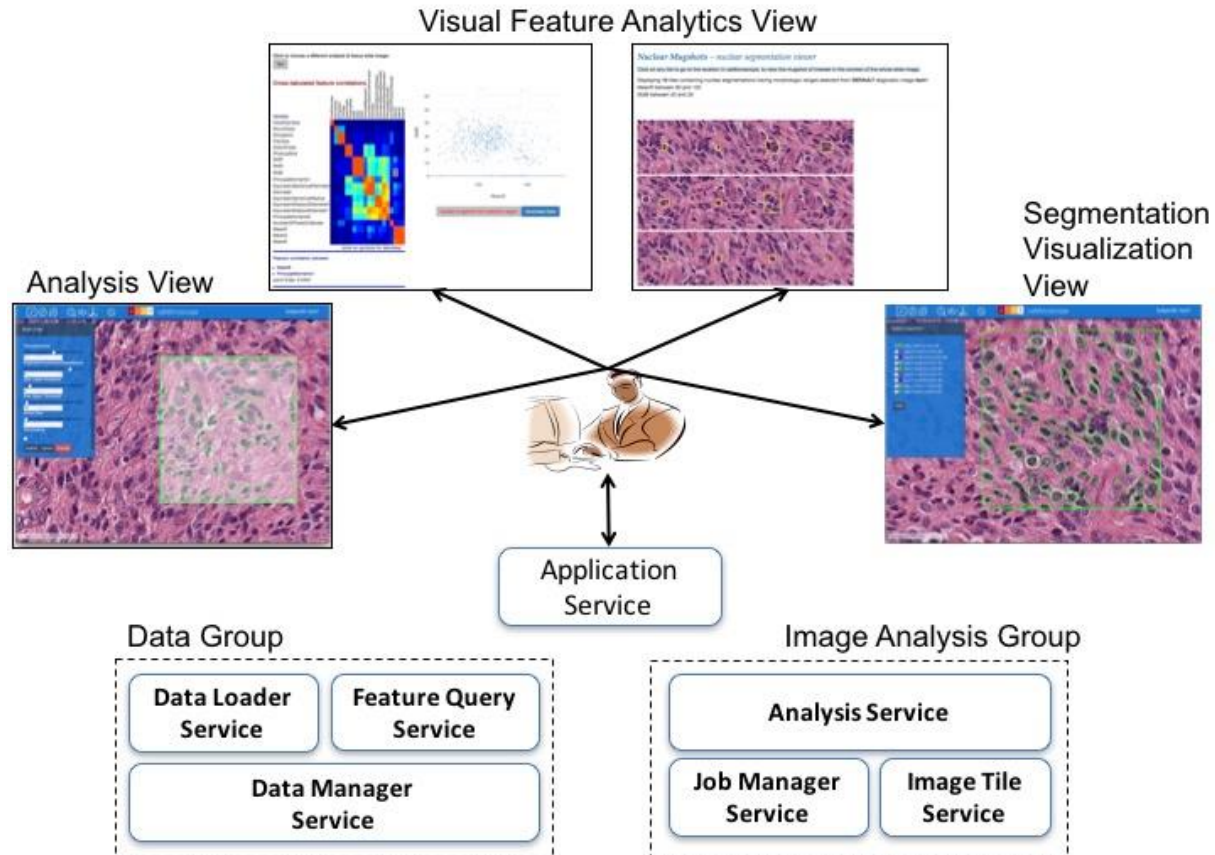
Visualization of sample set of segmented objects based on range queries on feature values



Visualization of whole slide tissue images and segmentation results using caMicroscope



Containerized Software for Tissue Image Analysis



Containerized Software for Tissue Image Analysis

- ***The application service group*** is a single container that hosts a suite of Web applications to view images and interact with analysis results.
- ***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

Software: https://github.com/SBU-BMI/quip_distro.git

Thank you.

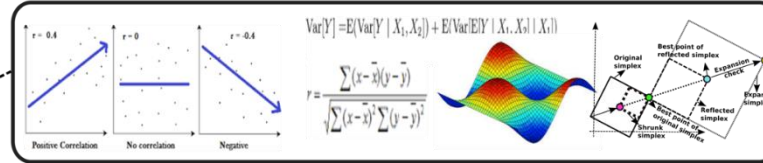
This work was supported in part by 1U24CA180924-01A1 from the NCI and by R01LM011119-01 and R01LM009239 from the NLM and NCIP/Leidos 14X138

Support for Sensitivity Analysis

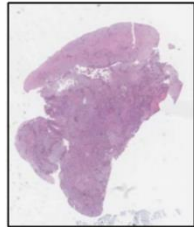
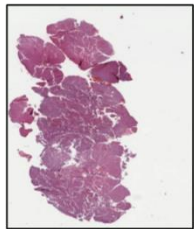
Sensitivity Analysis (SA) and Auto-tuning methods

Target Spatial metric:
Dice, Jaccard, Intersecton, etc.

One or More Application
Parameter Sets

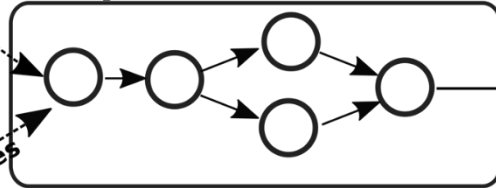


WSI
input
dataset



Input
Images

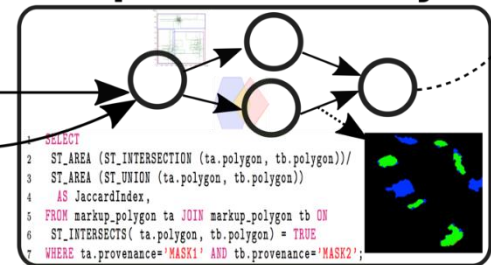
Application Workflow Composition/Instantiation



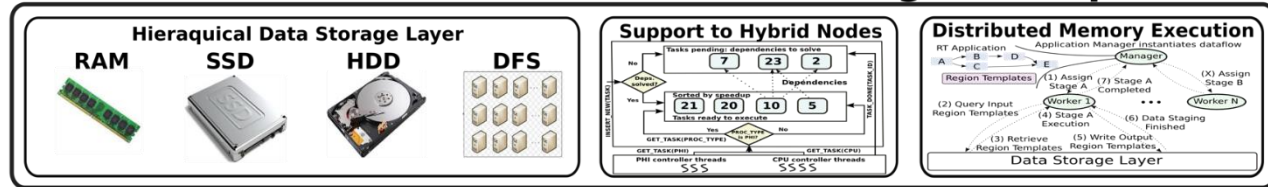
Segmentation
Computed

Reference
Segmentation

Spatial Query-based Comparative Analysis



Scalable and Efficient Execution with Region Templates



Supercomputer

