

# The Cancer imaging Phenomics Toolkit (CaPTk)

Christos Davatzikos, on behalf of the team



Center for Biomedical Image Computing and Analytics

Computational Breast Imaging Group

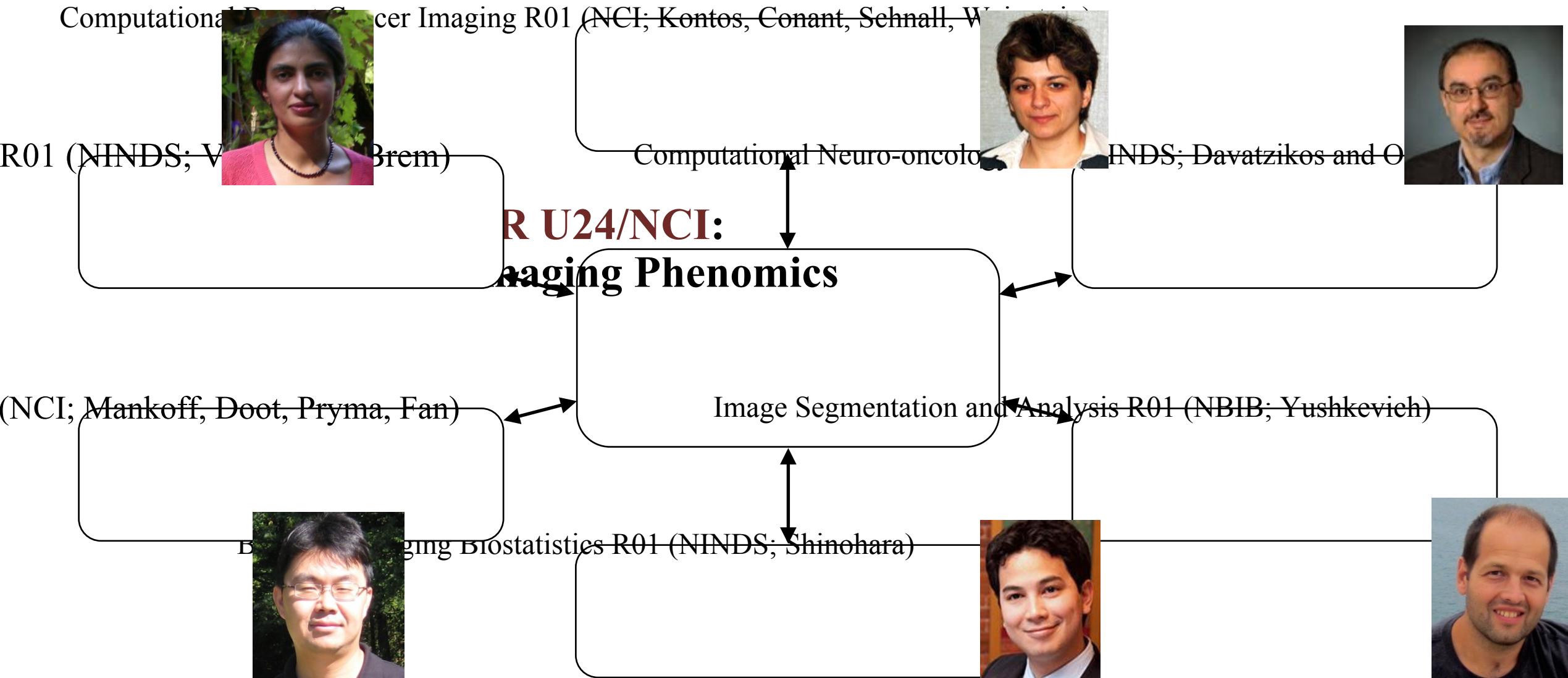
Penn Image Computing and Science Lab

Penn Statistical Imaging and Visualization Endeavor

Section for Biomedical Image Analysis



# Participating PIs



# Two Major Goals

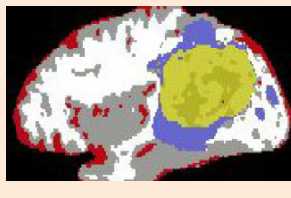
- To leverage a rich family of advanced image computing algorithms
- To leverage extensive and long-standing collaborations with clinical teams who have provided input in the development of the algorithms, as well as data for training and validation of models

## First Level

### Image Analysis Algorithms

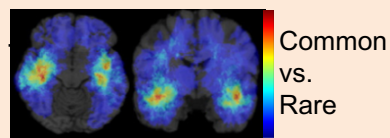
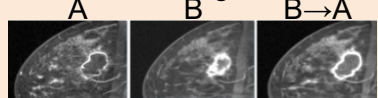
#### Segmentation:

Regions of Interest (ROIs)



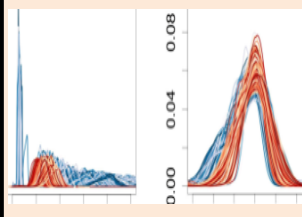
#### Registration:

- Measure change with time:  
A B B→A



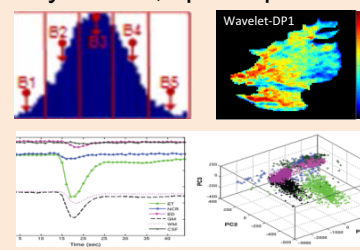
#### Image Harmonization:

Before After



#### Feature Extraction:

Texture, histogram, dynamics, spatial pattern



#### Image Operations:

- DICOM access
- Format conversion
- Intensity normalization
- Co-registration
- Noise Reduction
- ROI annotation
- Seed-point initialization

ITK

### CaPTk Radiomic Panel

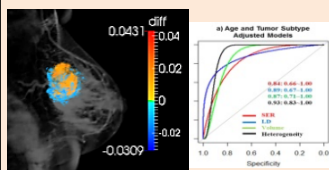
### Feature Synthesis and Integration via Machine Learning

Open-CV

## Second Level

### Output Modules and Outcomes

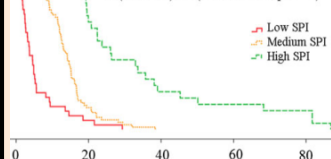
#### Precision Diagnosis, Risk Estimation:



Breast Density Factor

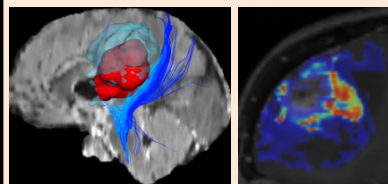
#### Predictive models:

HR(low&high):10.64(95% CI 5.9-19.3, p<0.001)  
HR(med&high):3.88(95% CI 2.3-6.6, p<0.001)  
HR(low&med):2.77(95% CI 1.8-4.2, p<0.001)



Kaplan-Meier Estimator

#### Personalized Treatment:

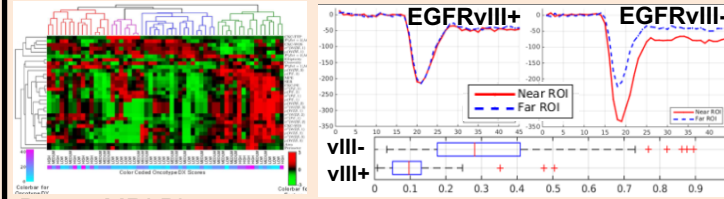


Connectomics

Infiltration

#### Radiogenomics

#### Imaging signatures of molecular characteristics:



Breast MRI Phenotypes  
vs. Oncotype DX

Imaging Signatures of  
GBM mutations



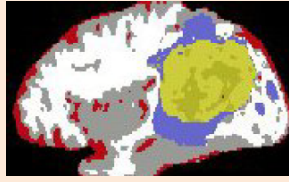
# Input Images

## First Level

### Image Analysis Algorithms

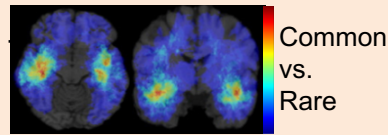
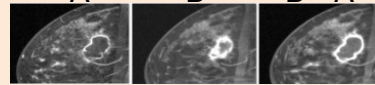
#### Segmentation:

Regions of Interest (ROIs)



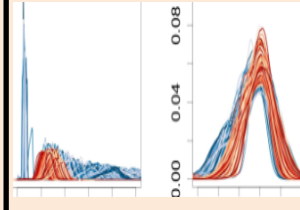
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- Measure change with time:  
A B B→A



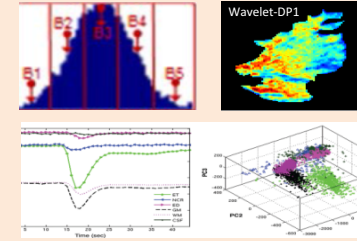
#### Image Harmonization:

Before After



#### Feature Extraction:

Texture, histogram, dynamics, spatial pattern



#### Image Operations:

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ITK

CaPTk Radiomic Panel

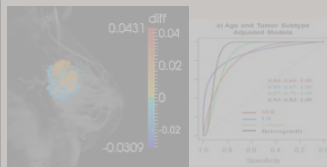
Feature Synthesis and Integration  
via Machine Learning

Open-CV

## Second Level

### Output Modules and Outcomes

#### Precision Diagnosis, Risk Estimation:



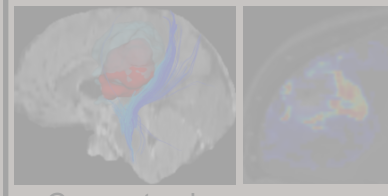
Breast Density Factor

#### Predictive models:



Kaplan-Meier Estimator

#### Personalized Treatment:

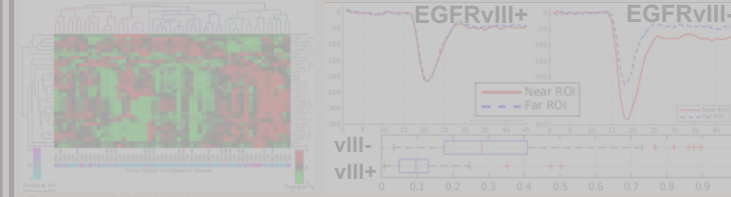


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#### Radiogenomics

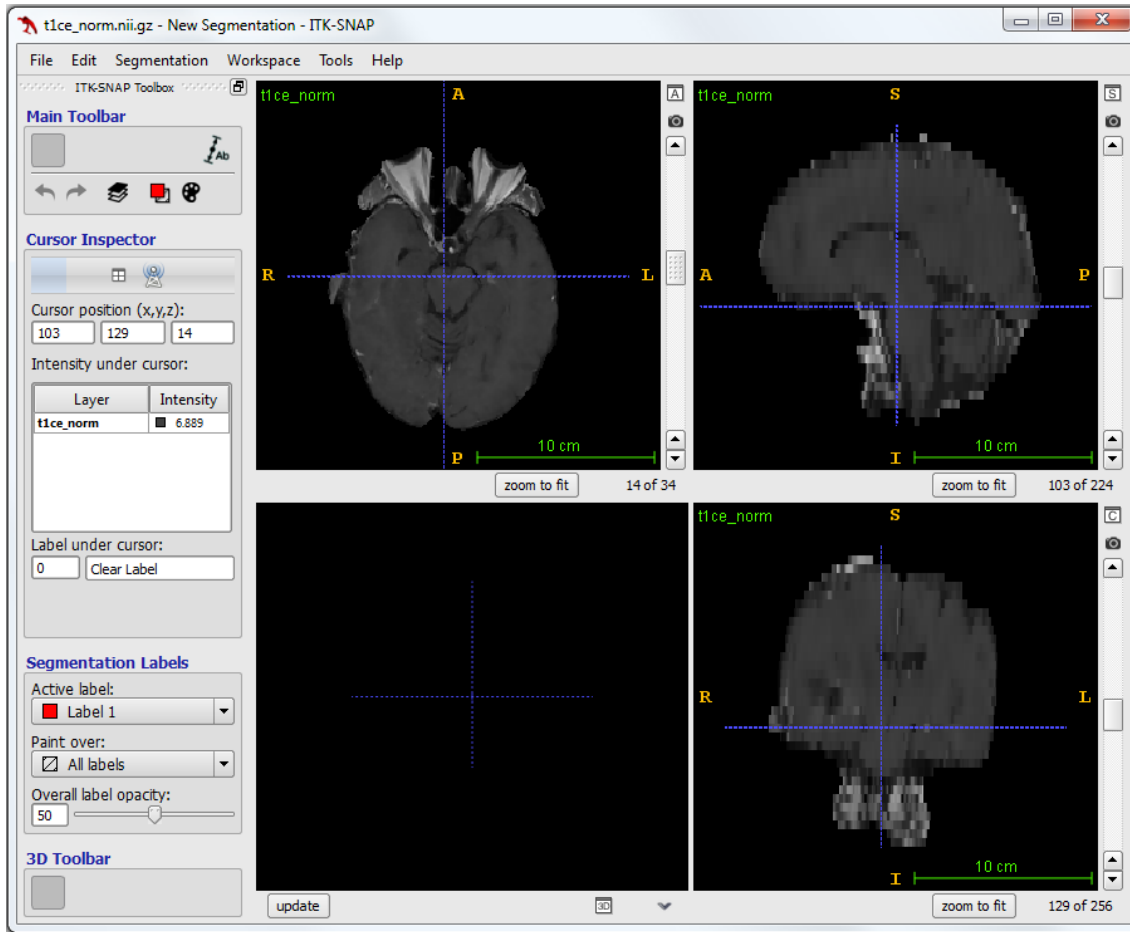
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Breast MRI Phenotypes  
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GBM mutations

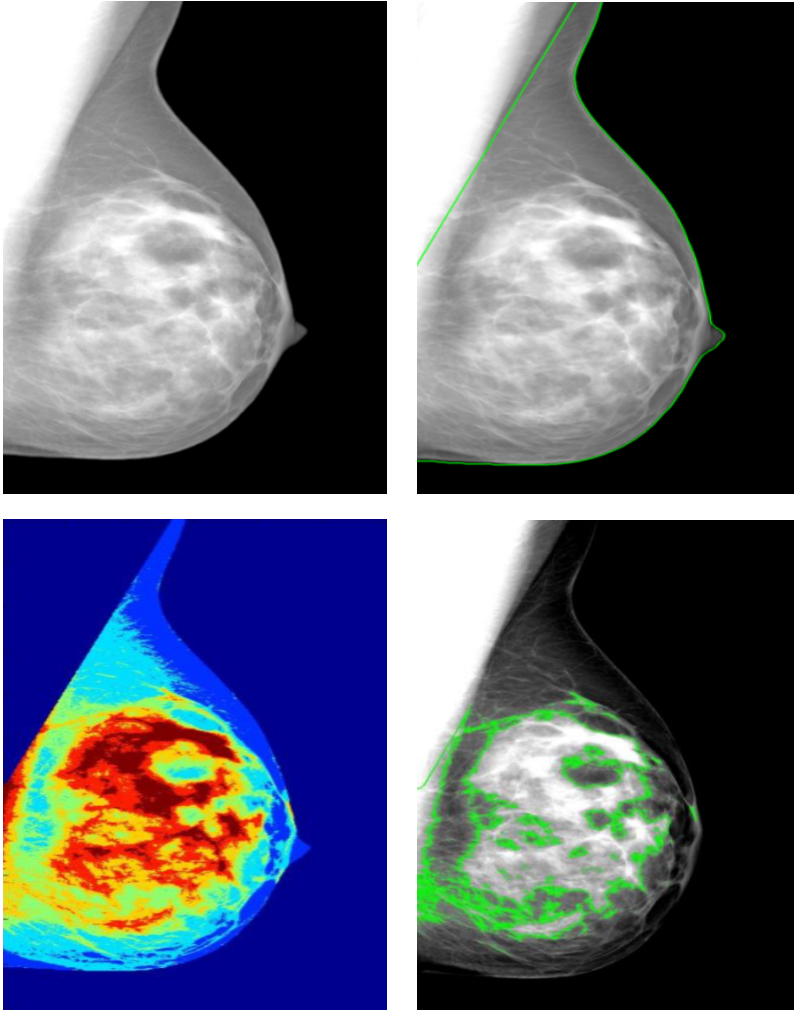
# ITK-SNAP



Screenshot of ITK-SNAP's interface

- Support for color, multi-channel, and time-variant images
- Segmentation done using Random Forest and Level Sets algorithms
- Transfer of data between ITK-SNAP and CaPTk is seamless – giving users the option to use the former's segmentation and user interaction functionality with latter's computation capabilities

# Breast Segmentation

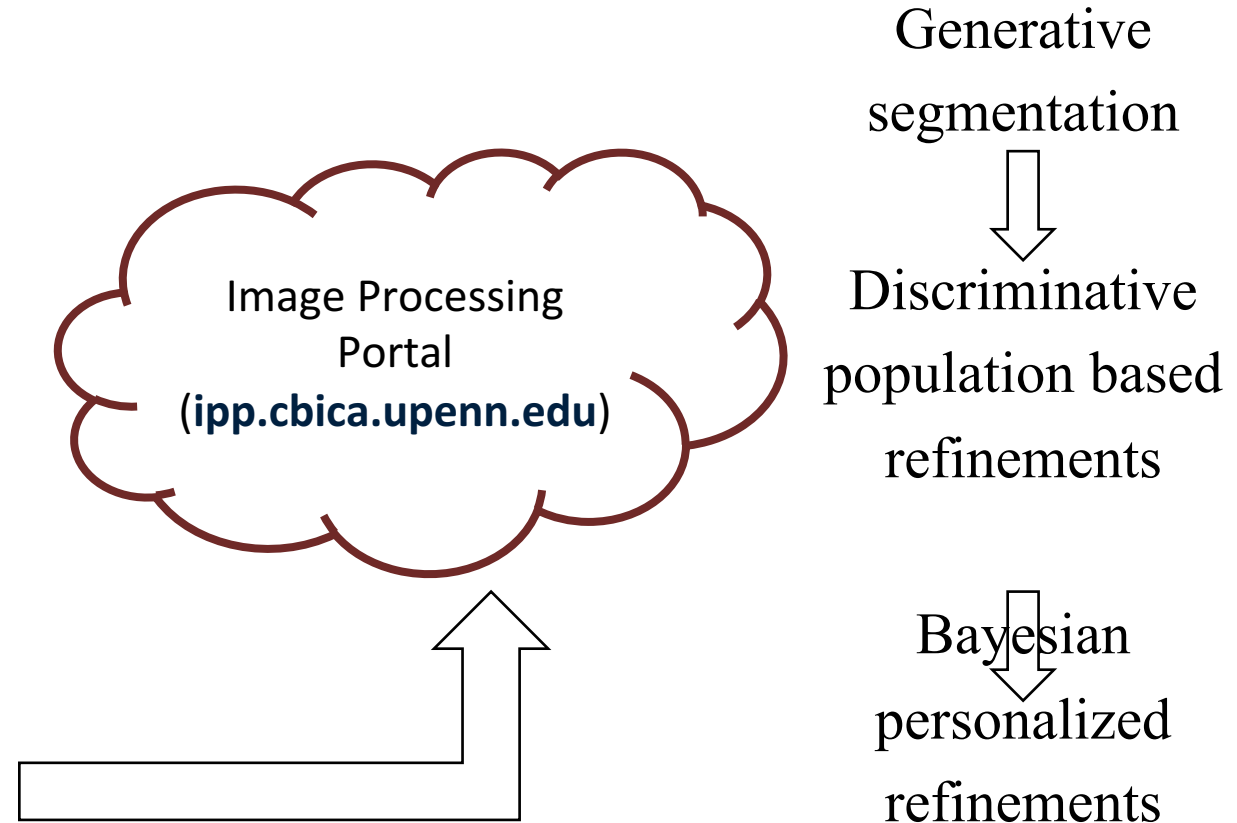


- Fully-automated segmentation of the breast area and the dense breast tissue, estimation of **Percent Density** (PD%)
- Adaptive fuzzy-c-mean (FCM) clustering based on intensity histogram and acquisition parameters
- Well calibrated versus radiologists estimates <sup>[9]</sup> demonstrated associations to breast cancer risk <sup>[10]</sup> for raw and processed FFDM.

# GLISTRboost Segmentation



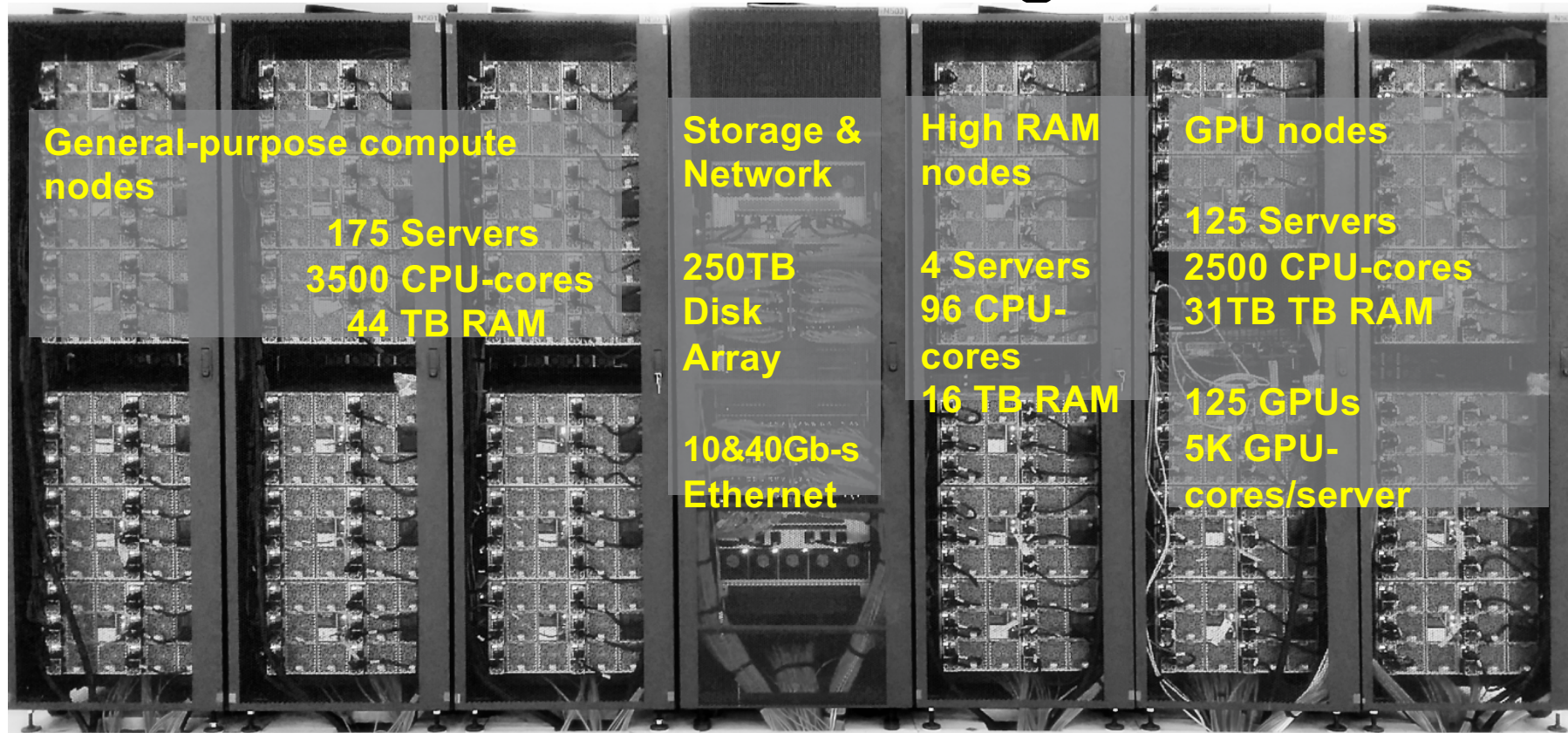
Initialize seed points for the MICCAI BRATS 2015 award-winning GLISTRboost method using CaPTk Console



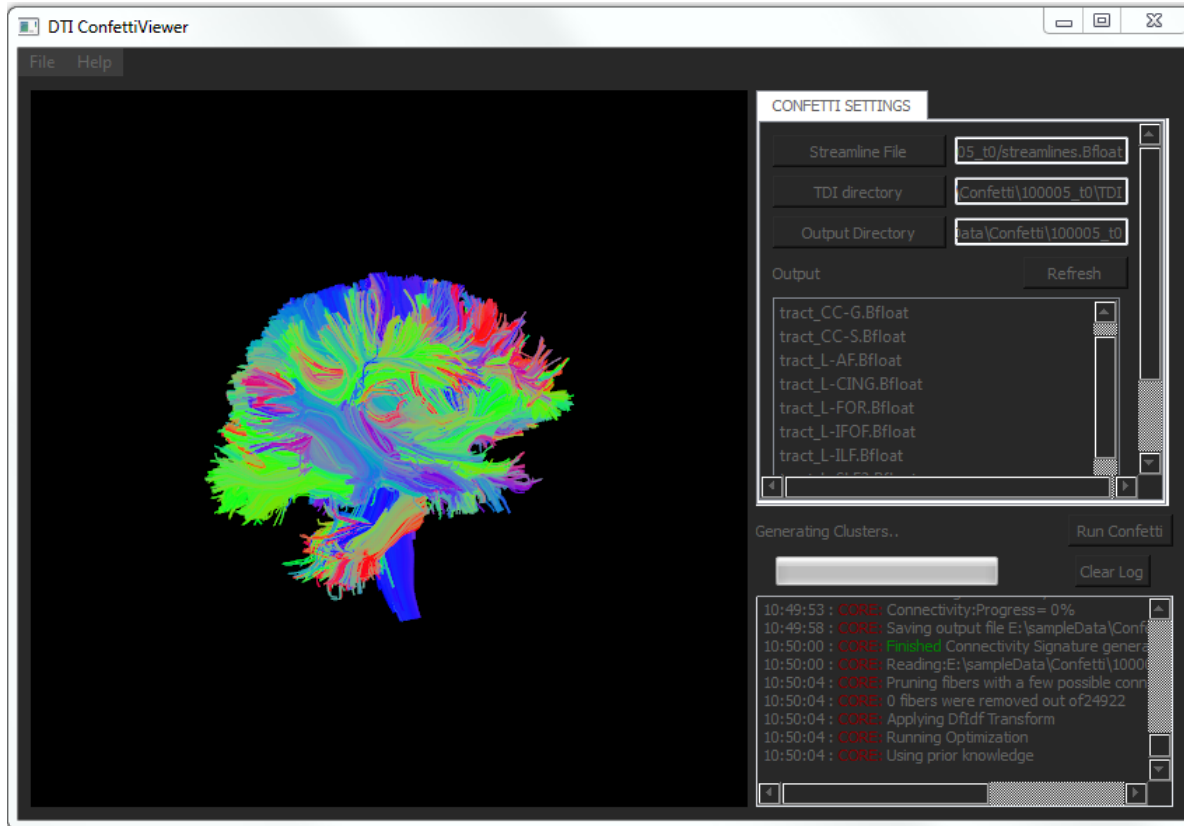


# Web Portal for Compute-heavy CaPTk

Functions: CBICA cluster via high-end NIH instrumentation grant:



# Confetti – Visualizing Fiber Tracts



Confetti interface is dedicated for fiber tract visualization and it is integrated with CaPTk Console

- Advances in tractography enhance neurosurgical planning, but are limited by edema, mass effect, and tract infiltration
- Confetti facilitates automated identification of all tracts (including eloquent tracts) without manual drawing of ROIs, making the tracts robust and replicable
- Confetti enhances the primary objective of neurosurgery: maximal safe resection in the presence of tumor edema and infiltration

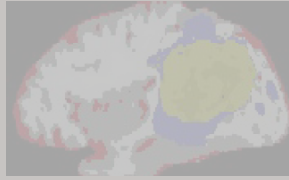
## Input Images

### First Level

#### Image Analysis Algorithms

##### Segmentation:

Regions of Interest (ROIs)

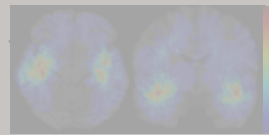


##### Registration:

- Measure change with time:  
A B B→A

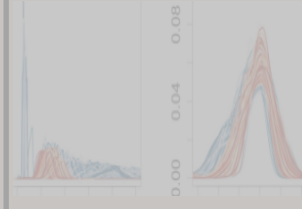


Common  
vs.  
Rare

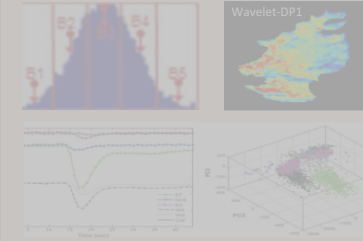


##### Image Harmonization:

Before After



##### Feature Extraction: Texture, histogram, dynamics, spatial pattern



##### Image Operations:

- DICOM access
- Format conversion
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ITK

## CaPTk Radiomic Panel

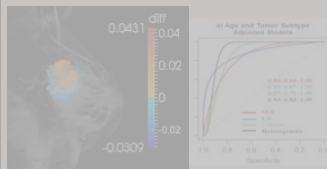
Feature Synthesis and Integration  
via Machine Learning

Open-CV

### Second Level

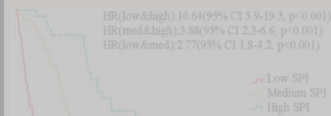
#### Output Modules and Outcomes

##### Precision Diagnosis, Risk Estimation:



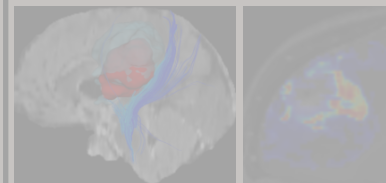
Breast Density Factor

##### Predictive models:



Kaplan-Meier Estimator

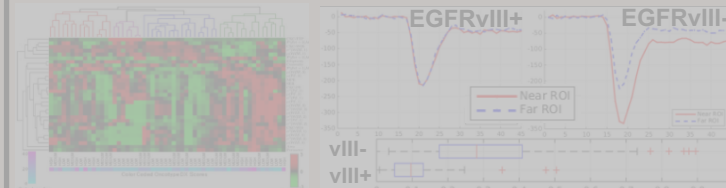
##### Personalized Treatment:



Connectomics

Infiltration

##### Radiogenomics Imaging signatures of molecular characteristics:



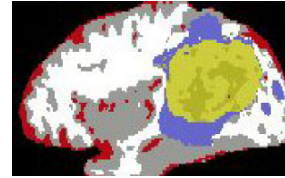
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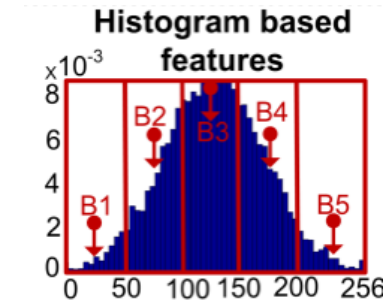


# CaPTk Radiomic Panel

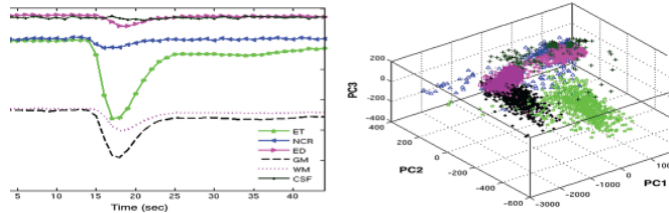
- Segmentations: volumes and signals within ROIs



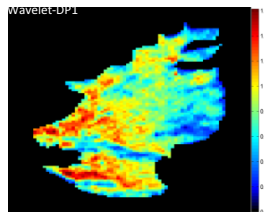
- Normalized histograms of different protocols; optimized binning



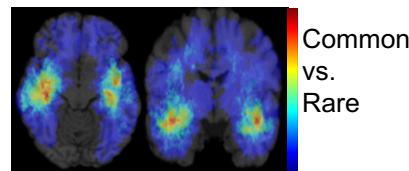
- Kinetics



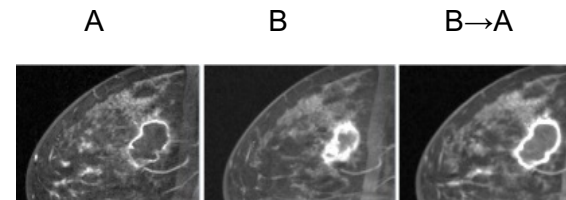
- Textures



- Spatial patterns/distributions



- Parametric maps from longitudinal scans



## Input Images

### First Level

#### Image Analysis Algorithms

##### Segmentation:

Regions of Interest (ROIs)



##### Registration:

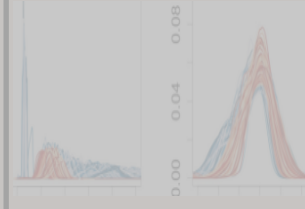
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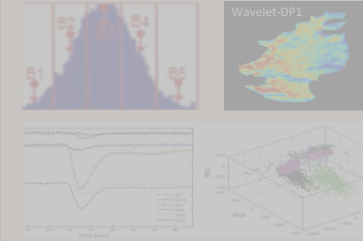
Common  
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ITK

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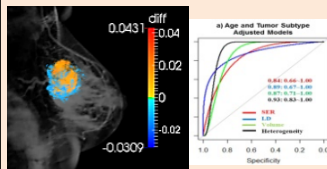
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Open-CV

### Second Level

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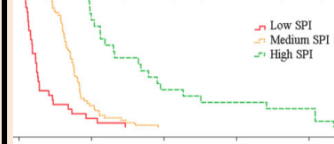
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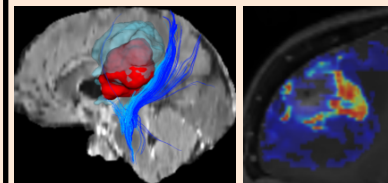
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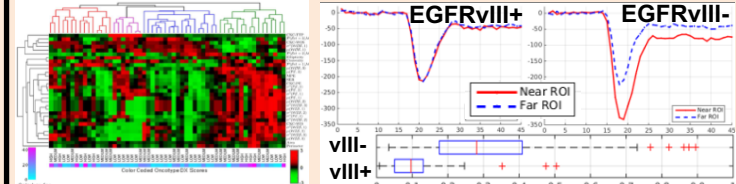
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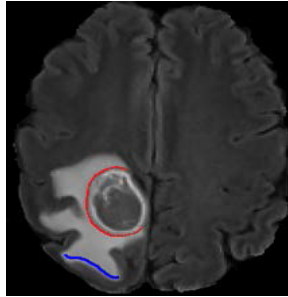


Breast MRI Phenotypes  
vs. Oncotype DX

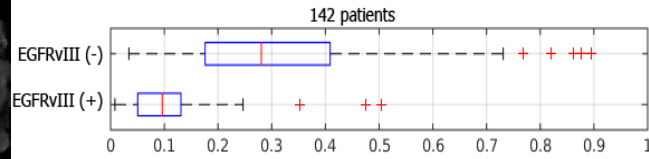
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# Computational Neuro-Oncology

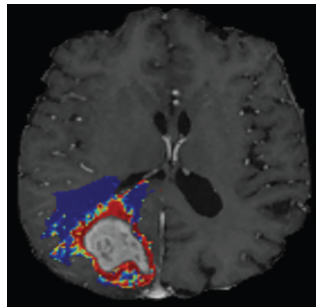
## Imaging Signatures of Molecular Characteristics



Bakas et al, Clinical Cancer Research 2017



## Predicting Infiltration and Recurrence

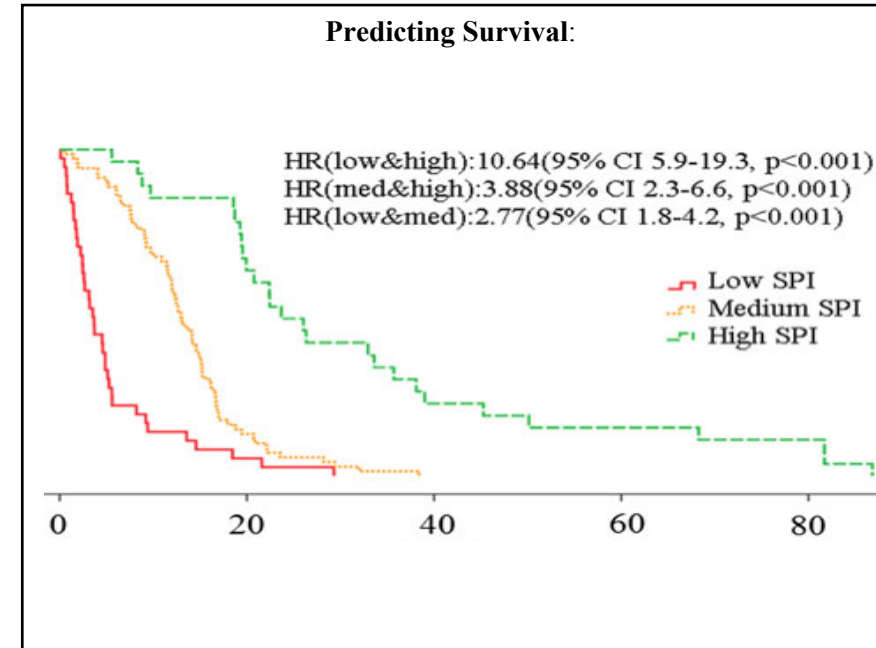


Akbari et al., Radiology, 2014  
Akbari et al., Neurosurgery, 2016

## Predictors of clinical outcome

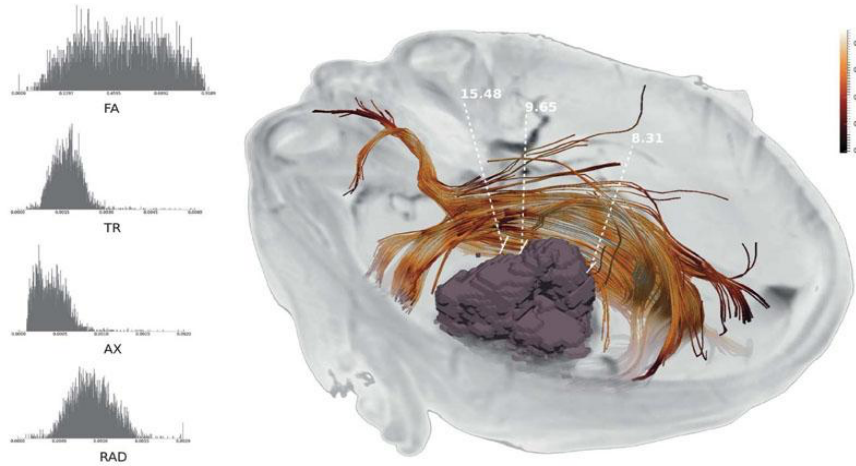


### Predicting Survival:



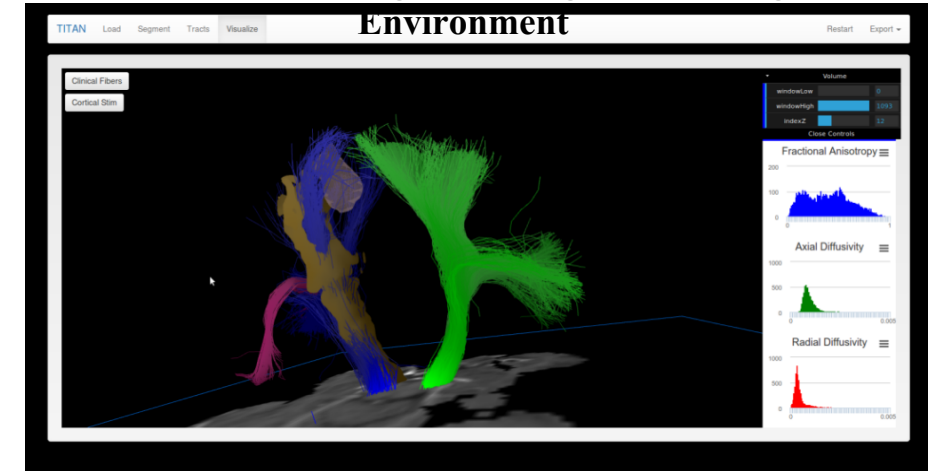
# Brain Connectomics

## Peri-lesional Effects of GBMs

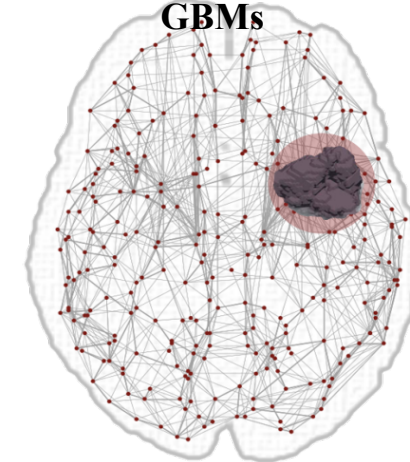


- Automated atlas-based tract extraction (using connectivity signatures instead of shape help address mass effect)
- Edema invariant tractography (using multicompartment models fitted to multishell imaging)
- Tumor connectome (effect of tumor on distant regions, regional vulnerability and functional rerouting)

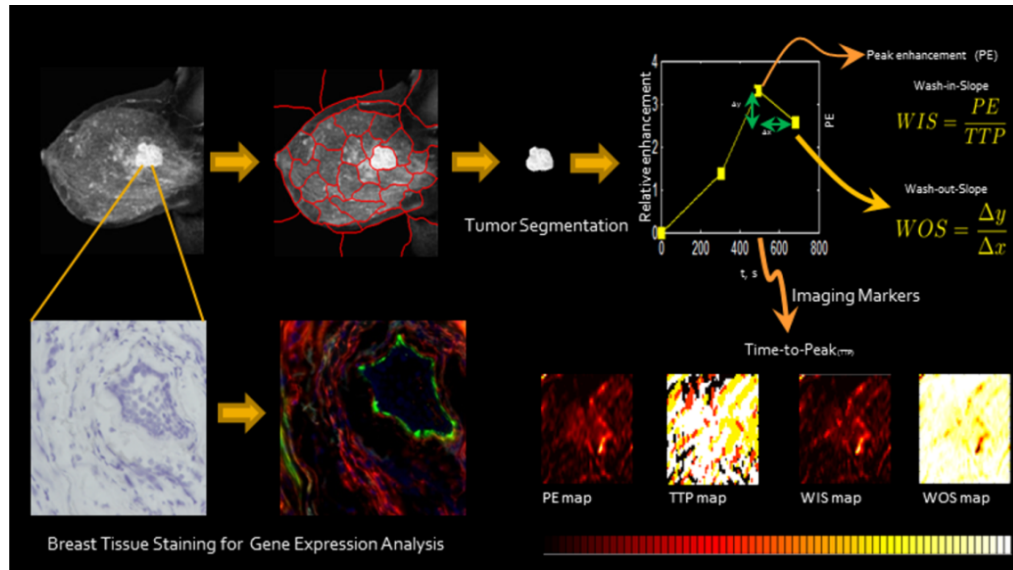
## Web-based Integrated Surgical Planning



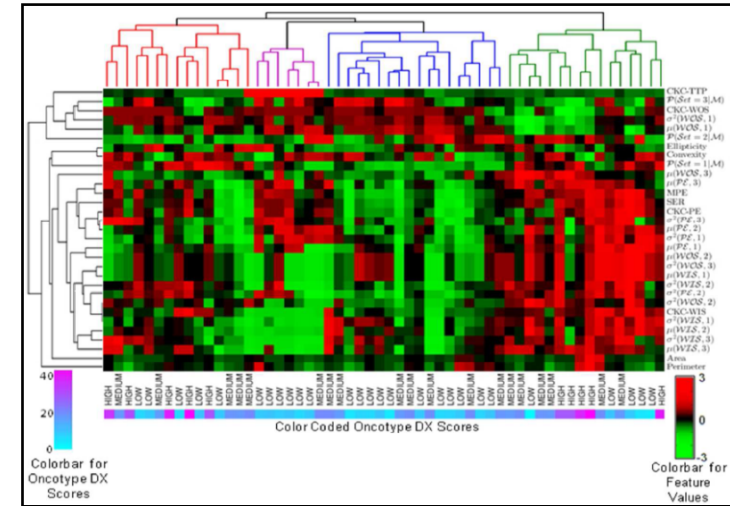
## Global Effects of GBMs



# Radiomic Breast Cancer Phenotypes



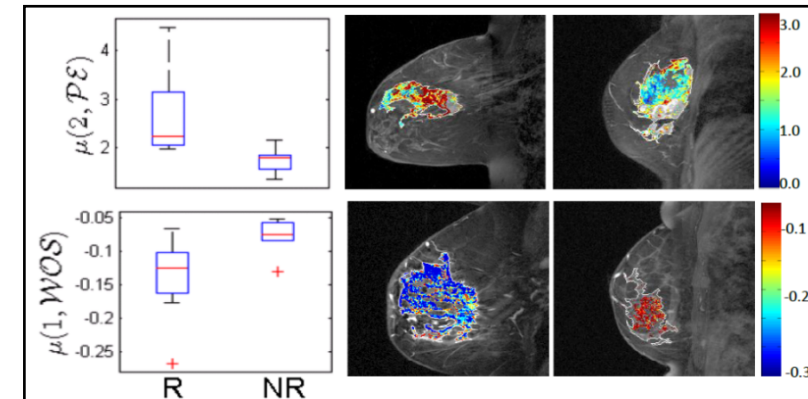
Ashraf et al., IEEE TMI 2013; Mahrooghy et al. IEEE TBME 2015



## Intrinsic Imaging Phenotypes for Breast Cancer Prognostic and Predictive Value

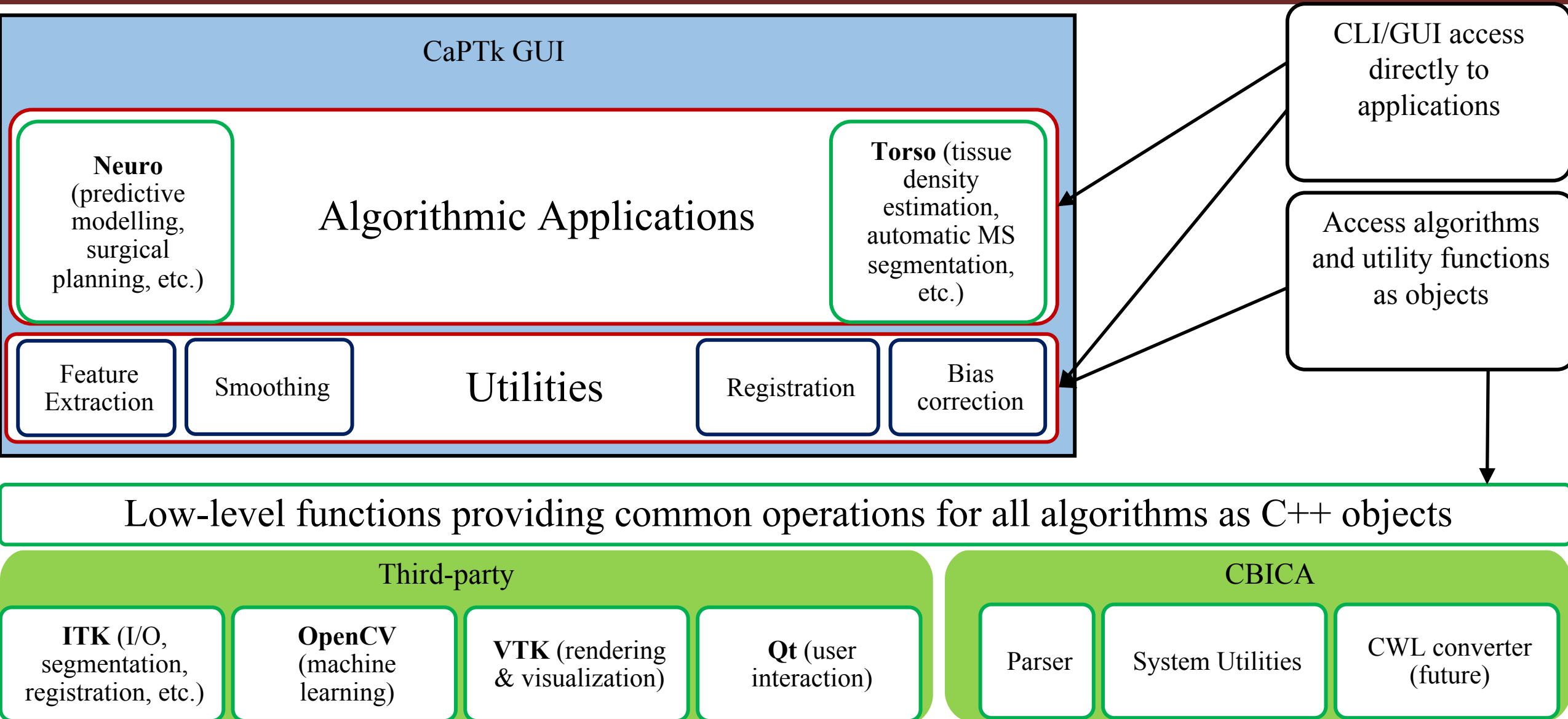
## Breast Cancer Phenotyping via Imaging:

- Segmentation and multi-parametric feature extraction
- Identification of intrinsic phenotype patterns
- Prognostication and treatment response prediction





# Software Architecture Overview



# Installation

- Only 64-bit machines supported due to processing requirement
- Installers are designed to work as double-click and install
- Windows and Linux installers are available and macOS to be ready soon
- The entire CaPTk (console and all applications) will also be available as docker images soon

Operating System	Version	Installer System
Windows	7+	MSI
macOS	Snow Leopard+	DMG
Ubuntu	14.04+	DEB
centOS	6+	RPM



# Dependencies

- Core Application (written in C++ for efficiency):

Library	Utility	Reference
Qt	Graphical Framework of application	<a href="http://www.qt.io">www.qt.io</a>
Visualization Toolkit (VTK)	Visualization of images, charts, etc.	<a href="http://www.vtk.org">www.vtk.org</a>
Insight Toolkit (ITK)	I/O, image processing, etc.	<a href="http://www.itk.org">www.itk.org</a>
OpenCV	Machine Learning	<a href="http://www.opencv.org">www.opencv.org</a>

- Different Components (can be C++ or Python):

Library	Utility	Reference
numpy	Higher level math operations	<a href="http://numpy.org">numpy.org</a>
pyqt	User interface of some applications	<a href="http://wiki.python.org/moin/PyQt">wiki.python.org/moin/PyQt</a>
dipy	Diffusion Imaging	<a href="http://nipy.org/dipy">nipy.org/dipy</a>

# Video

