

### TCGA computational histopathology pipeline reveals subtypes and their molecular signature

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#### Computational histopathology pipeline captures molecular basis for each morphometric subtype

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## Use case and target for analysis in the second seco

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- Glioblastoma multiforme (GBM)
  - Curated by removing tissue sections with artifacts (e.g., fold in tissue, pen mark, scanning anamoly)
  - Sample size
    - 380 tissue sections selected out of 447
      - 146 patients selected out of 152
- Challenges?
  - Technical and biological variations, very large datasets
- Approach
  - Development of robust and efficient image analysis algorithms
  - Computing morphometric features and meta-features
  - Subtyping based on selected features or reduced dimensionality (e.g., PCA, MDS)
  - Molecular association with morphometric subtypes

#### New algorithm enhances nuclear segmentation in the presence of technical variations



# Seed detection provides shape signature and local statistics



### **Cell-by-cell segmentation result**





### **Cell-by-cell segmentation result**





### Representation

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Normalization across all tissue sections

## What are subtypes based on cellularity and nuclear size at the patient level



What is the distribution of each subtype and how well each subtype predicts survival as a function of treatment?



# What are the molecular basis of each subtype?



- Gene selection
  - Univariate or multivariate methods
  - Pathway or subnetwork enrichment analysis

	Name	<b>Overlapping Entities</b>	-	p-value	
Subtype1	Focal Adhesion Regulation Actin Cytoskeleton	CAV1,MET,ERBB4,KIT,PDGFRA,RASA4 0.00		0.000208	
	Regulation	MET,ERBB4,KIT,PDGFRA,SGCE,RASA4,PDLIM3		0.000555	
	Gap Junction Regulation Adherens Junction	MET,ERBB4,KIT,NPY2R,PDGFRA,RASA4 0.0082		0.008248	
	Regulation	DAAM2,MET,ERBB4,KIT,PDGFRA,CDH6		0.011068	
	KIT -> STAT signaling	КІТ		0.017364	
	HGFR -> STAT signaling	MET 0.023089		0.023089	
	PDGFR -> STAT signaling	PDGFRA		0.025939	
	HGFR -> FOXO3A signaling	MET		0.054015	
Subtype3	Name	<b>Overlapping Entities</b>			p-value
	CCR1 -> STAT signaling	CCL4,CCL3			0.003127
	CCR5 -> TP53 signaling	CCL4,CCL3			0.004022
	Gap Junction Regulation	GNAO1,CCL4,HRH1,KIT,CCL3,CALCRL,ADCY2,FGF12,RASA4		0.008737	
	KIT -> STAT signaling	KIT			0.033533
Subtype4	Name	<b>Overlapping Entities</b>	p-value		
	IL11R -> STAT3 signaling	IL11RA	0.018322		
	ThromboxaneR -> CREB				
	signaling	RASGRP1,GNG4	0.026307		
	EphrinR -> actin signaling	EFNB3,SGCE,EPB41L2	0.02702		

# Can tumor composition be characterized?



 Since tumor is heterogeneous, can we query for subtypes at the block levels and learn about tumor composition?



#### What are the tumor histology subtypes?



## Does heterogeneity play a role in survival as a result of a more intense therapy ?





# Another view: Are cellularity and nuclear size correlated? And outcome?



High cellularity and low nuclear size are better predictive of a more aggressive therapy

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### Conclusion



- There are many ways to slice through the data and metadata
  - Cellularity, nuclear size
  - Heterogeneity
- Different indices lead to alternative subtypings
  - Alternative biological interpretation is possible
- Genomic association has the potential to reveal new insight
- Web site: tcga.lbl.gov
  - "Google map" like viewing of tissue sections with segmentation results overlaid