Code, Chat and Collab 13:00, Feb 17, 2023





A brief tutorial on exploring clinical associations in cancer samples using SEAS

Analysis in Systems Biology using genomic and clinical features of samples

Genomic Insights such as: Enriched Pathways, Genesets, PAGS, etc.

Co-membership and regulatory gene AND/OR protein networks.

Metabolic models with Gene perturbation.

Genomic **Analysis**

7897620 11.58961496 11.51082781 11.67322239 11.34210418 11.5495066

Samples

Gene Expression Profiling of patient samples

Gene Expression Matrix

Clinical Features i.e., Clinotypes

81.57

34.85

GSM1830170 49 Control

GSM1830171 70 COPD

Correlation, Enrichment, etc. Analysis

Clinical Analysis

age disease fev1 fvc fev1 predicted sex smoking status statin user GSM1830157 57 COPD 43.13 Former smoker N GSM1830158 72 COPD 48.21 GSM1830159 70 COPD 59.93 Former smoker N GSM1830160 57 COPD 40.2 Former smoker GSM1830161 62 Control 76.93 GSM1830162 67 COPD 43.07 Former smoker N GSM1830163 60 COPD 28.97 Former smoker Y GSM1830164 66 COPD 43.52 GSM1830165 74 COPD Former smoker | N 42.04 GSM1830166 61 COPD Former smoker N GSM1830167 70 COPD GSM1830168 68 COPD 45.52 Former smoker | N GSM1830169 71 COPD 57.02

UMAP, PCA, **tSNE**

Genotypically AND/OR Phenotypically resolved patient embedding.

Former smoker Y

Former smoker N

sample_id	X1	X2
GSM1830157	2.998425653	-1.391358321
GSM1830158	-1.401068876	1.81903131
GSM1830159	-0.396056858	-2.896121068
GSM1830160	1.55616542	-1.620665901
GSM1830161	0.160953406	2.76096781
GSM1830162	0.030356896	-2.519666312
GSM1830163	-2.740846083	-0.013063805
GSM1830164	0.939161057	-1.217158655
GSM1830165	-1.645087191	-0.856133376
GSM1830166	1.132666571	-0.6928069
GSM1830167	-1.612448305	-0.466482063
GSM1830168	1.175554088	-0.42713845
GSM1830169	2.117619142	0.180591641
GSM1830170	2.284750197	1.673612929
GSM1830171	2.215471982	1.133465544

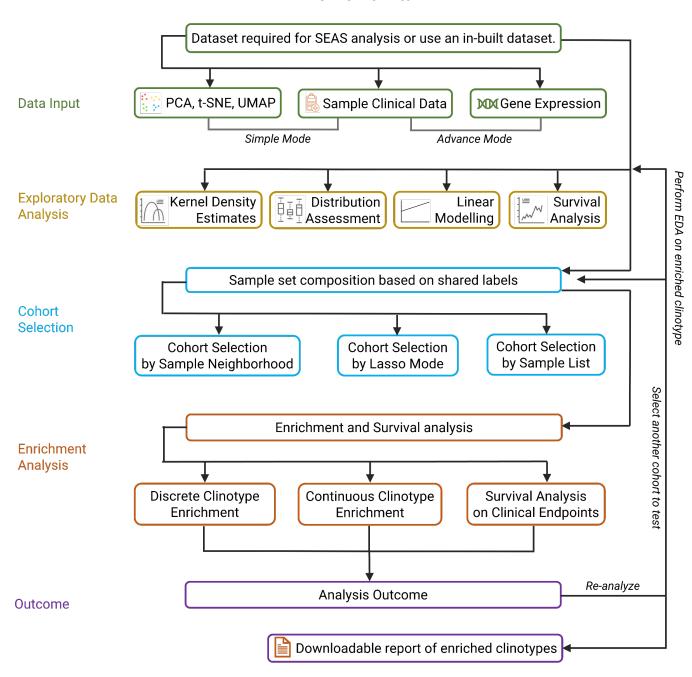
A snapshot of Clinical Features associated to 389 TCGA GBM patients

A B	C	D	E	F	G	Н	1	J	K	L	M	N	0	P	Q	R	S	T	U	V	W	Х	Υ	Z	AA	AB
	Cluster		Discrete_CDE_chemo																							
TCGA-02-C TCGA	Cluster 1		<=100	<=50	>300	>300	<=250	>250	<=200	>80<=100		FALSE	FALSE	FALSE	FALSE	0		FALSE		FALSE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA		>50<=65	<=100	<=50	<=300	<=300	<=250	>250	<=200			FALSE	FALSE	FALSE	FALSE	0	FALSE	FALSE		FALSE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUI
TCGA-02-C TCGA		>50<=65	>100	>50	>300	>300	>250	>250	>200	>80<=100	59.18		TRUE	TRUE	TRUE	110	TRUE	TRUE	110	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUI
TCGA-02-C TCGA	Cluster 1		>100	>50	>300	>300	>250	>250	>200	>80<=100	40.53		TRUE	TRUE	TRUE	306	TRUE	TRUE	306	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA		>50<=65	<=100	<=50	>300	>300	>250	>250	>200	>80<=100		FALSE	FALSE	FALSE	FALSE	0	FALSE	FALSE		FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUI
TCGA-02-C TCGA	Cluster 1	<=50	<=100	>50	>300	>300	>250	>250	>200	>80<=100	20.4		TRUE	TRUE	TRUE	61	TRUE	TRUE	61	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUI
TCGA-02-C TCGA	Cluster 1	<=50	>100	>50	>300	>300	<=250	<=250	>200	>80<=100	18.96	TRUE	TRUE	TRUE	TRUE	125	TRUE	TRUE	125	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUI
TCGA-02-C TCGA	Cluster 2	<=50	<=100	<=50	>300	>300	>250	>250	>200		25.65		FALSE	FALSE	TRUE	0	FALSE	TRUE	0	FALSE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUI
TCGA-02-C TCGA	Cluster 2	>50<=65	<=100	<=50	>300	>300	>250	>250	>200	>80<=100	50.39		TRUE	FALSE	TRUE	0	FALSE	FALSE		FALSE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUI
TCGA-02-C TCGA	Cluster 1	<=50	>100	>50	>300	>300	>250	>250	>200	>80<=100	43.9	FALSE	TRUE	TRUE	TRUE	119	TRUE	TRUE	119	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRU
TCGA-02-C TCGA	Cluster 2	<=50	<=100	<=50	>300	>300	>250	>250	>200	>80<=100	38.34	TRUE	TRUE	TRUE	TRUE	44	TRUE	TRUE	44	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRU
TCGA-02-C TCGA	Cluster 1	<=50	>100	>50	>300	>300	>250	>250	>200		35.91	FALSE	TRUE	TRUE	TRUE	551	TRUE	TRUE	551	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUI
TCGA-02-CTCGA	Cluster 1	<=50	>100	>50	>300	>300	>250	>250	>200		47.64	FALSE	TRUE	TRUE	TRUE	539	TRUE	TRUE	539	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUI
TCGA-02-CTCGA	Cluster 2	<=50	>100	>50	>300	>300	>250	>250	>200	>80<=100	27.44	FALSE	TRUE	TRUE	TRUE	327	TRUE	TRUE	327	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 1	<=50	>100	>50	>300	>300	>250	>250	>200		33.86	TRUE	TRUE	TRUE	TRUE	230	TRUE	TRUE	230	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 1	<=50	>100	>50	>300	>300	>250	>250	>200		39.16	FALSE	TRUE	TRUE	TRUE	366	TRUE	TRUE	366	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 1	>50<=65	<=100	<=50	<=300	<=300	<=250	<=250	>200		54.95	FALSE	FALSE	FALSE	FALSE	0	FALSE	FALSE	0	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE
TCGA-02-C TCGA	Cluster 2	>50<=65	<=100	>50	>300	>300	>250	>250	>200	>80<=100	60.69	FALSE	TRUE	TRUE	TRUE	71	TRUE	TRUE	71	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 1	<=50	<=100	<=50	>300	>300	<=250	<=250	>200	>80<=100	48.59	FALSE	FALSE	FALSE	FALSE	0	FALSE	FALSE	0	FALSE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUI
TCGA-02-C TCGA	Cluster 2	>50<=65	<=100	<=50	>300	>300	<=250	<=250	>200	>60<=80	54.93	FALSE	FALSE	FALSE	FALSE	0	FALSE	FALSE	0	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUI
TCGA-02-CTCGA	Cluster 1	>50<=65	<=100	<=50	>300	>300	>250	>250	>200	>80<=100	54.43	FALSE	TRUE	FALSE	TRUE	63	TRUE	FALSE	0	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 1	>50<=65	<=100	<=50	<=300	<=300	<=250	<=250	>200	>60<=80	61.37	FALSE	FALSE	FALSE	FALSE	0	FALSE	FALSE	0	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE
TCGA-02-CTCGA	Cluster 1	>65	<=100	<=50	>300	>300	<=250	<=250	>200	>80<=100	78.74	FALSE	FALSE	FALSE	FALSE	0	FALSE	FALSE	0	FALSE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE
TCGA-02-CTCGA	Cluster 2	>65	<=100	<=50	<=300	<=300	<=250	<=250	>200		80.22	FALSE	FALSE	FALSE	FALSE	0	FALSE	FALSE	0	FALSE						
TCGA-02-C TCGA	Cluster 2	<=50	<=100	>50	>300	>300	<=250	<=250	>200		43.76	TRUE	TRUE	TRUE	TRUE	69	TRUE	TRUE	69	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE
TCGA-02-C TCGA	Cluster 1	<=50	>100	>50	>300	>300	<=250	<=250	>200	>80<=100	49.45	TRUE	TRUE	TRUE	TRUE	171	TRUE	TRUE	171	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 1	<=50	<=100	<=50	<=300	<=300	<=250	<=250	>200	>80<=100	44.42	TRUE	TRUE	TRUE	TRUE	46	TRUE	TRUE	46	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 1	>65	>100	>50	>300	>300	>250	>250	>200	>80<=100	66.09	FALSE	TRUE	FALSE	TRUE	331	TRUE	TRUE	331	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA	Cluster 1	<=50	>100	>50	<=300	<=300	<=250	<=250	>200	>80<=100	28.79	TRUE	TRUE	TRUE	TRUE	123	TRUE	TRUE	123	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 2	>65	<=100	<=50	<=300	<=300	<=250	>250	<=200	>80<=100	68.71	TRUE	TRUE	TRUE	TRUE	43	TRUE	TRUE	43	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE
TCGA-02-CTCGA	Cluster 1		>100	>50	<=300	<=300	<=250	<=250	>200	>80<=100	66.12	TRUE	TRUE	TRUE	TRUE	168	TRUE	TRUE	168	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-C TCGA	Cluster 2	>50<=65	>100	>50	>300	>300	>250	>250	>200		50.05	TRUE	TRUE	TRUE	TRUE	336	TRUE	TRUE	336	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA		>50<=65	>100	>50	>300	>300	<=250	>250	<=200	>80<=100	57.93	TRUE	TRUE	TRUE	TRUE	166	TRUE	TRUE	166	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA		>50<=65	<=100	<=50	<=300	<=300	<=250	<=250	>200	>80<=100	53.18		FALSE	FALSE	FALSE	0	FALSE	FALSE	0	FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALS
TCGA-02-CTCGA	Cluster 2		>100	>50	>300	>300	<=250	<=250	>200	>80<=100	68.19		TRUE	TRUE	TRUE	233	TRUE	TRUE	233	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA		>50<=65	<=100	<=50	>300	>300	>250	>250	>200	>80<=100	63.53		TRUE	TRUE	TRUE	21	TRUE	TRUE	21	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA		>50<=65	<=100	<=50	>300	>300	>250	>250	>200	>80<=100	57.97		TRUE	TRUE	TRUE	42	TRUE	TRUE	42	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE
TCGA-02-CTCGA	Cluster 2		<=100	>50	>300	>300	>250	>250	>200	>80<=100	28.22		TRUE	TRUE	TRUE	92	TRUE	TRUE	92	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA		>50<=65	>100	>50	>300	>300	>250	>250	>200	>80<=100	59.21		TRUE	TRUE	TRUE	337	TRUE	TRUE	337	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA	Cluster 2		<=100	<=50	>300	>300	>250	>250	>200	>80<=100	36.31		FALSE	FALSE	FALSE	0	FALSE	FALSE		FALSE	FALSE	FALSE	FALSE	FALSE	FALSE	FALS
TCGA-02-CTCGA		>50<=65	<=100	<=50	>300	>300	>250	>250	>200	>80<=100	63.76		TRUE	TRUE	TRUE	46	TRUE	TRUE	46	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA	Cluster 2		<=100	>50	<=300	<=300	<=250	<=250	>200	×30/-100	45.89		TRUE	TRUE	TRUE	73	TRUE	TRUE	73	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA			>100	>50	>300	>300	>250	>250	>200		45.89 52.66		TRUE	TRUE	TRUE	216	TRUE	TRUE		TRUE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
		>50<=65								>00/-100									216							TRUE
TCGA-02-CTCGA	Cluster 2		>100	>50	<=300	<=300	>250	>250	>200	>80<=100	46.76		TRUE	TRUE	TRUE	137	TRUE	TRUE	137	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	
TCGA-02-CTCGA	Cluster 2		>100	<=50	>300	>300	>250	>250	>200	-004-405	42.87		TRUE	FALSE	TRUE	246	TRUE	FALSE	0	FALSE	FALSE	TRUE	TRUE	TRUE	TRUE	TRUE
TCGA-02-CTCGA	Cluster 2	K=50	>100	>50	>300	>300	>250	>250	>200	>80<=100	29.29	TRUE	TRUE	TRUE	TRUE	251	TRUE	TRUE	251	TRUE	FALSE	FALSE	TRUE	TRUE	TRUE	TRU

Statistical Enrichment Analysis of Sample Clinical Attributes Using SEAS

- Embedding techniques has gained popularity in visualizing the highdimensional gene expression profiles of patient samples yet the systematic extraction of sample set composition based on shared labels instead of shared embedding neighbourhood remains a major challenge.
- SEAS can be used to perform exploratory analysis of embedded sample data by focusing on the "clinotypes" of selected sample sets.
- Clinotypes: Clinotypes are referred as the clinical/phenotypical features
 of a sample. For SEAS analysis clinotypes are classified in two i.e.,
 discrete and continuous clinotypes.
- Discrete Clinotypes are the clinotypes which take specific value in quantitative or qualitative data. For examples, age groups, cancer subtypes, treatment method, etc.
- Continuous Clinotypes are the clinotypes which take continuous quantitative values. For example, age, survival days, treatment days, dose levels, etc.
- CFEA: Clinical Feature Enrichment Analysis is a method defined in SEAS to identify clinotypes which are over-represent in a selected cohort from population.
- We used Hypergeometric Test, KS-test, and Kaplan-Meier Method to perform discrete clinotype enrichment, continuous clinotype enrichment and survival analysis, respectively.

SEAS Workflow



https://github.com/informaticsclub/ccc presentations

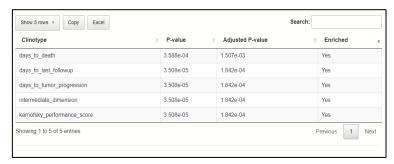
Functional Enrichment Analysis of GBM patients uncovers clinical/phenotypic difference in additional chemotherapy lacking cohort.

We acquired and preprocessed TCGA-GBM dataset, which consists of 389 patients, according to the pipeline in Jia et al. (2018). The dataset had both the genetic and the clinical sections. We also used 45 GBM tumor-samples hosted in patient-derived xenograft (PDX) models (Willey et al., 2020). We performed SEAS analysis to test enrichment in the patient samples where no additional chemotherapy was given

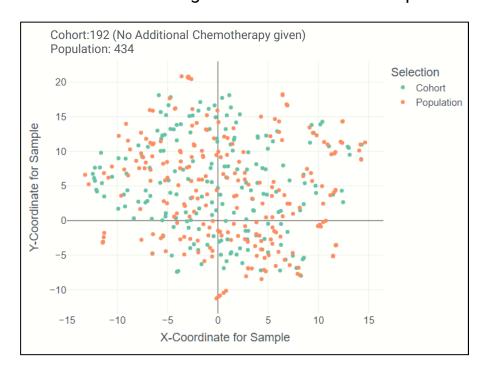
5 of 50 Enriched Discrete Clinotypes

Show 5 rows v Copy Excel		Searc	h:					
Clinotype	♦ Variable	P-Value	Adjusted P-value	Enriched -	N \$	n \$	K (k
Discrete_CDE_DxAge	>65	1.826e-06	3.579e-05	Yes	389	192	128	85
Discrete_CDE_chemo_alk_days	<=100	2.571e-06	4.725e-05	Yes	389	192	245	143
Discrete_CDE_chemo_tmz_days	<=50	1.073e-03	9.279e-03	Yes	389	192	245	136
Discrete_CDE_survival_time	<=300	4.200e-15	2.470e-13	Yes	389	192	157	115
Discrete_days_to_death	<=300	1.667e-15	1.226e-13	Yes	389	192	156	115
Showing 1 to 5 of 50 entries	'		Previous 1	2 3 4	5		10	Next

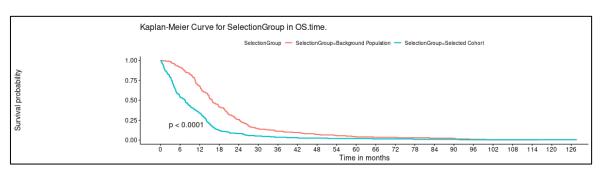
5 Enriched Continuous Clinotypes



UMAP Embedding of 434 GBM cancer samples



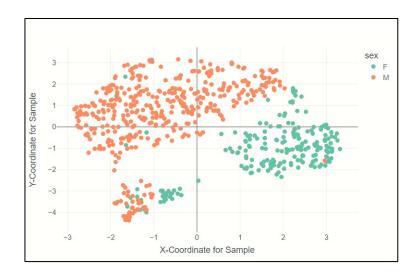
Survival Difference Between Cohort and Population.



- We found 50 discrete and 5 continuous enriched clinotypes in the selected cohort. Performing survival analysis we found a significant p-value of 0.0001 suggesting there's a significant effect of additional chemotherapy on survival of GBM cancer patients.
- GBM cancer patients who didn't receive additional chemotherapy died earlier.

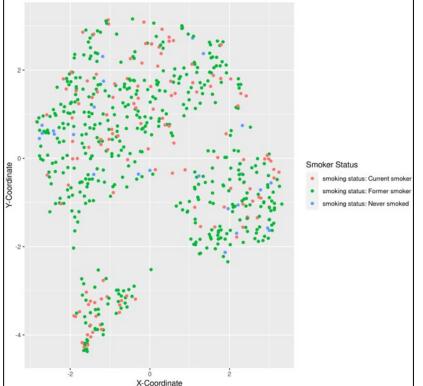
Exploratory Analysis of COPD Patient Profiles using SEAS reveal understandings in clinotype-explained genomic variation.

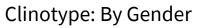
We acquired and preprocessed clinical and genetic data of 617 COPD patient samples publicly accessible at GEO: GSE71220. We perform a basic exploratory analysis of data using SEAS to demonstrate how researchers can identify clinotype showing high association with sample embedding.

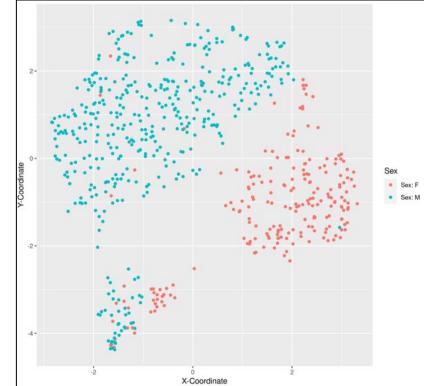


UMAP embedding of 617 COPD patient samples

Clinotype: By Smoking Status







- We showed a simple exploratory analysis using SEAS on COPD patient samples revealing sample embedding highly associated by gender difference than originally studied smoking status clinotype.
- COPD patient transcriptomes were highly influenced by gender difference than smoking status.