

SURVIVAL ANALYSIS

Poster: Integration Strategies for Multi-Omics Survival Analysis

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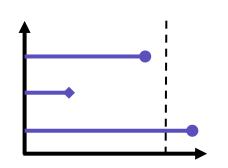




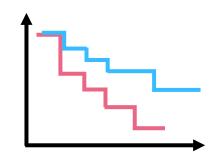


SURVIVAL ANALYSIS

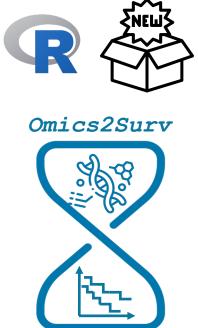
with Multi-Omics Data

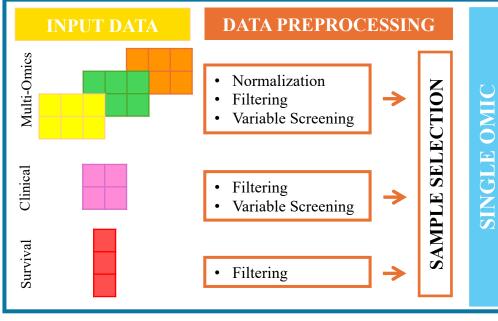




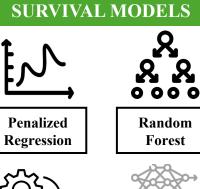






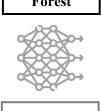


MULTI-OMICS
INTEGRATION
STRATEGIES



Cooperative

Learning



Stratification

Deep Learning Markers Discovery

OUTCOMES

Performance

Metrics

Individual Risk

Scores

Patients

PENALIZED REGRESSION

COOPERATIVE LEARNING

RANDOM FOREST

COOPERATIVE LEARNING

PENALIZED REGRESSION

Cox Proportional Hazard Model

Hazard function

$$h(t \mid X_i) = h_0(t) \exp(X_i \beta)$$

Partial log-likelihood

$$\ell(\beta) = \sum_{i: \delta_i = 1} \left(X_i^{\top} \beta - \log \sum_{j \in R_i} \exp(X_j^{\top} \beta) \right)$$

Penalized objective

$$\hat{\beta} = \arg \max_{\beta} \left\{ \ell(\beta) - \lambda P(\beta) \right\}$$

Penalizations $\begin{cases} P(\beta) = \|\beta\|_1 \text{ (LASSO)} \\ P(\beta) = \alpha \|\beta\|_1 + (1-\alpha)\|\beta\|_2^2 \text{ (Elastic Net)} \\ P(\beta) = \sum_j w_j |\beta_j| \text{ (Adaptive LASSO)} \\ P(\beta) = \alpha \|\beta\|_1 + (1-\alpha)\beta^\top L\beta \text{ (Network-based)} \end{cases}$



- \bigcirc Handles high-dimensional omics data $(p \gg n)$
- Produces interpretable linear risk scores
- Automatic variable selection



Note | This is also applicable to AFT models

COOPERATIVE LEARNING

Cooperative Cox Proportional Hazard Model¹

Hazard function

$$h(t \mid U_i, Z_i) = h_0(t) \exp(U_i \beta_U + Z_i \beta_Z)$$

Penalized $\max_{\beta_U,\beta_Z} \ell_{U,Z}(\beta_U,\beta_Z) - \rho \|U\beta_U - Z\beta_Z\|_2^2 - \lambda \|\beta_U\|_1/\hat{\beta}_U + \|\beta_Z\|_1/\hat{\beta}_Z$ objective

Encourages different omics to "agree" on predictive risk scores

Agreement parameter



Note | Support only 2 omics at the moment



Note | This is also applicable to AFT models²

RANDOM FOREST

RANDOM FOREST



The group of covariates corresponding to one specific data type is denoted as a 'block'

$$mtry = (\sqrt{p_1}, \sqrt{p_2}, ..., \sqrt{p_M})$$

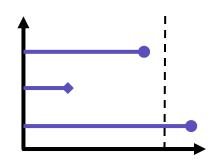
$$Log-rank(s) = \frac{(O_L - E_L)^2}{Var(O_L)}$$

- Nonlinear effects, interactions captured
- Robust to high-dimensional and noisy omics data
- No proportional hazards assumption

THE PACKAGE

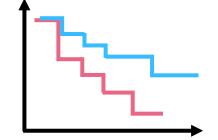
		Survival Model	R Package	Integration Strategy	Omics supported
Omics2Surv		Penalized Regression	glmnet	Early + Late	2+
	COX based	Random Forest	blockForest	Joint	2+
		Cooperative Learning	coop	Joint	2
WORK IN PROGRESS	COX based	Penalized Regression	NETSURVPROX	Early + Late	2+
	OON Dased	Deep Learning	AUTOSURV	Joint	<i>Σ</i> +
	AFT based	Penalized Regression	NETSURVPROX	Early + Late	2
		Cooperative Learning	AFTCoop	Joint	<i>L</i>

COMPARATIVE ANALYSIS

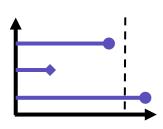












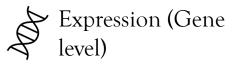
DATASETS

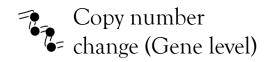




LinkedOmics

https://www.linkedomics.org





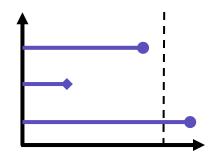
Dataset	Number of patiens (n)	RNA features Raw(p) Screened(m)		CNV features (p) (m)		Early Integration features
Lower Grade Glioma (LGG)	478	20086	2562	24776	3716	6278
Pan-Kidney Cohort (KIPAN)	851	20190	2499	24776	3716	6215
Head and Neck Squamous Cell Carcinoma(HNSC)	505	20164	2513	24776	3716	6229

SUMMARY

Analysis Modality			E	Late Integration							
Dataset	Train Test		Method	Alpha	Lambda min	(d)	C-index Train	C-index Test	C-index Train	C-index Test	
			LASSO	1	0,0716	20	0,8859	0,8530	0,5600	0,5954	
LGG 382	96	EN	0.2	0,4022	52	0,8768	0,8661	0,5450	0,5862		
		ADAPTIVE	1	0,2290	50	0,9221	0,8705	0,5289	0,5744		
				LASSO	1	0,0403	53	0,8015	0,7886	0,5164	0,3634
KIPAN 681	170	EN	0.8	0,1215	13	0,7589	0,7645	0,5201	0,4001		
		ADAPTIVE	1	0,2545	224	0,9580	0,8197	0,5225	0,4006		
HNSC 404	4 101	LASSO	1	0,0616	54	0,8048	0,6174	0,4902	0,4811		
		EN	0,9	0,0610	70	0,8205	0,6315	0,4893	0,4670		
			ADAPTIVE	1	0,1316	222	0,9889	0,6444	0,4951	0,4623	

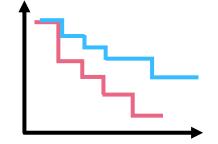
	Joint Integration			Joint Integration	
Method	C-index Train	C-index Test	Method	C-index Train	C-index Test
Random Forest	0,8172	0,8206	Cooperative	0,4055	0,4928
Random Forest	0,7262	0,7677	Cooperative	0,4742	0,5
Random Forest	0,5663	0,5901	Cooperative	0,4978	0,5

RESULTS









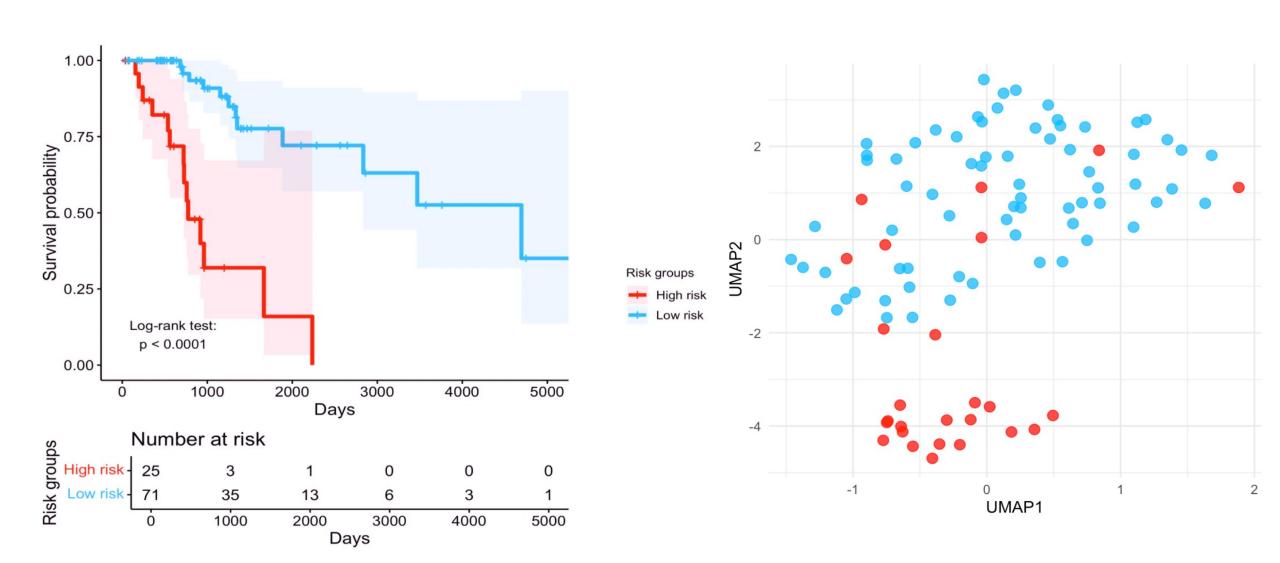




LGG

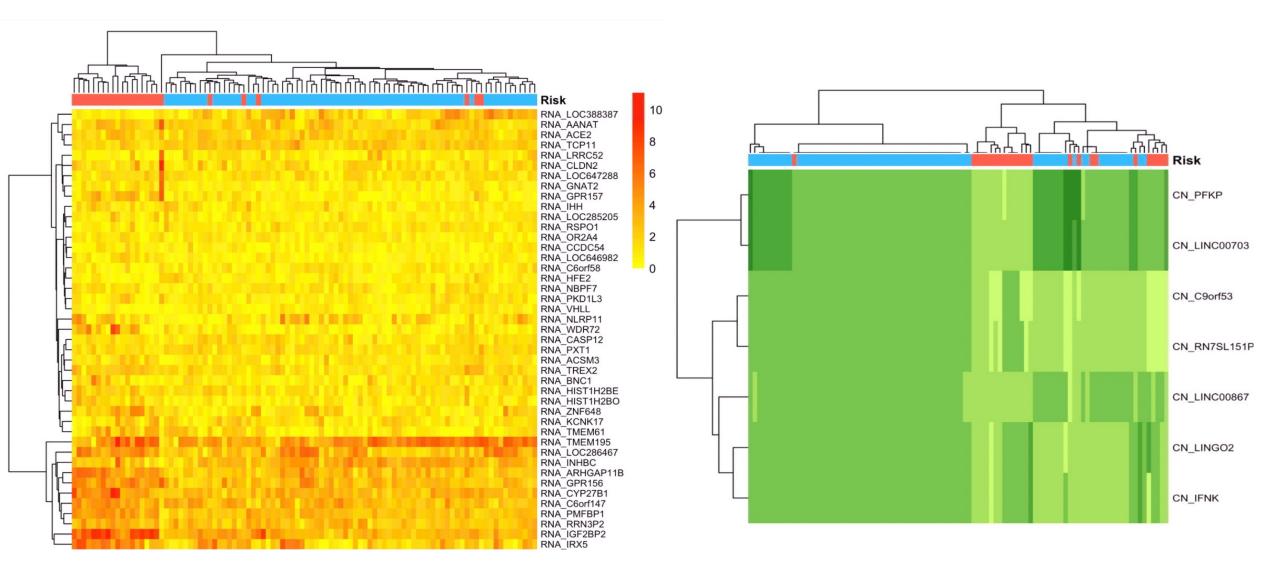


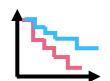
Patient stratification based on early integration with adaptive lasso-cox results



LGG Heatmaps based on model's selected features



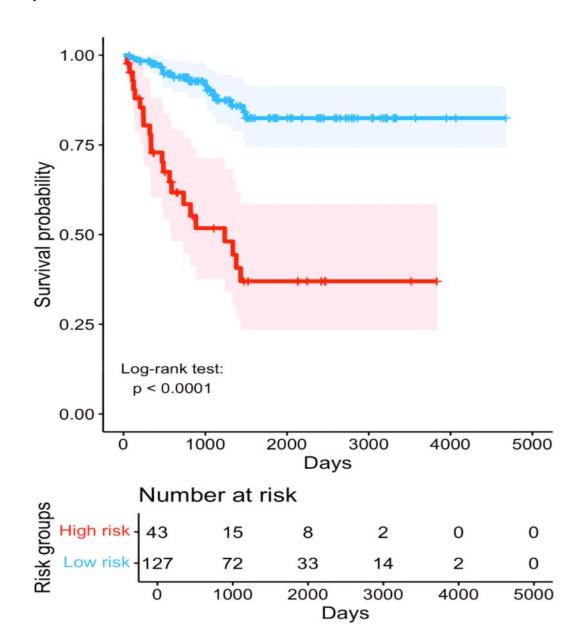


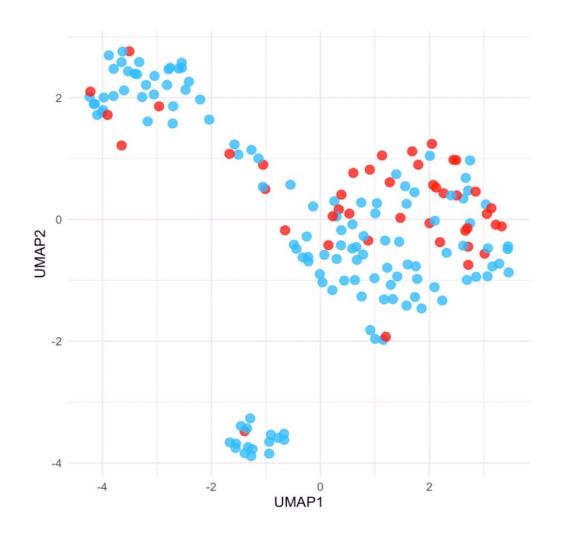


KIPAN



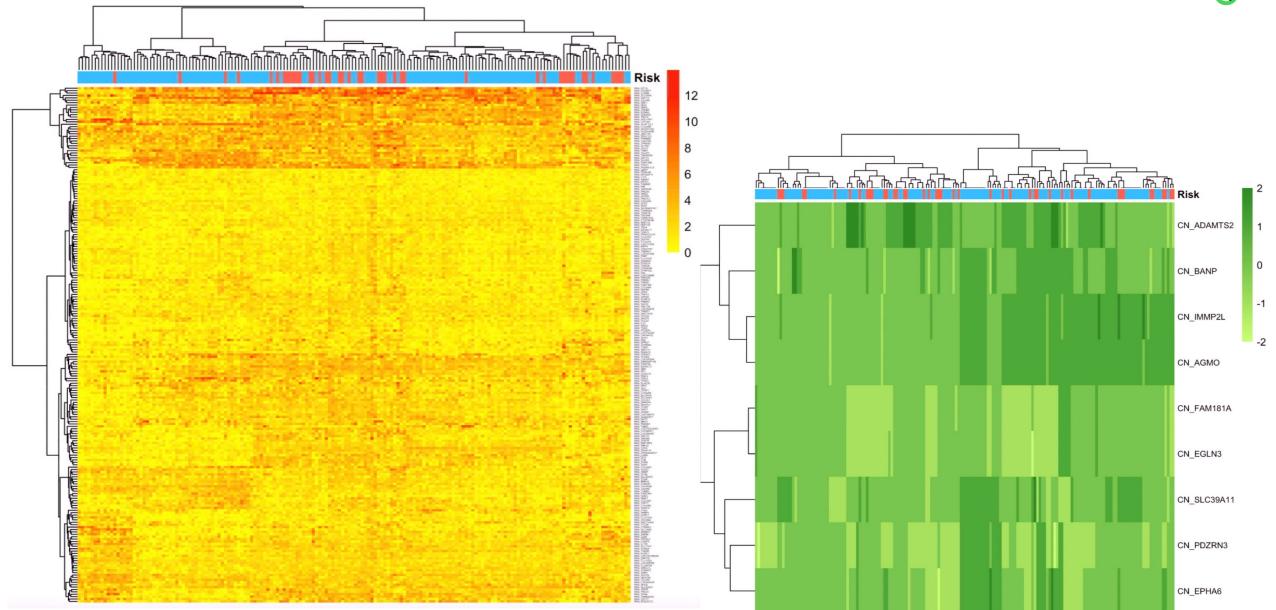
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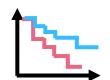






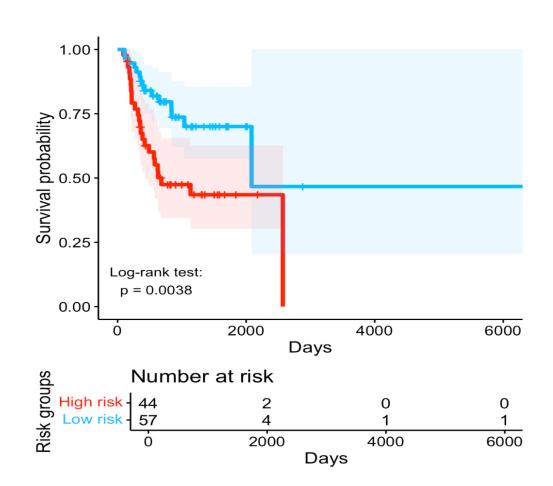


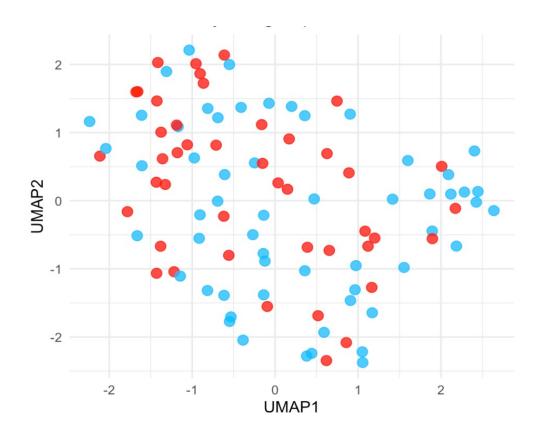






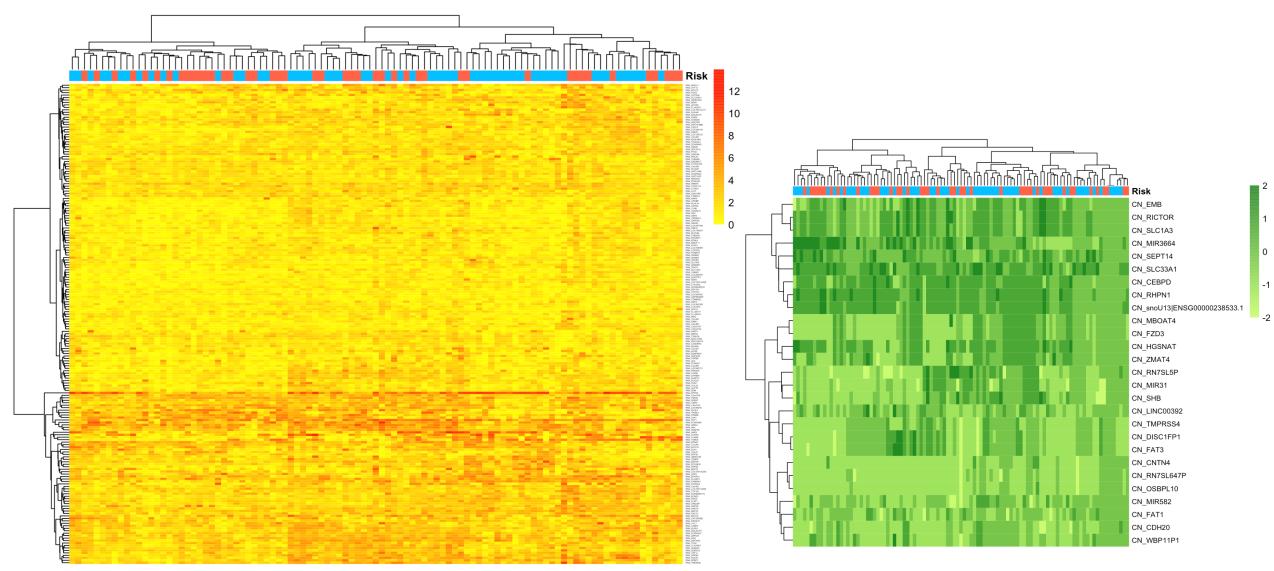
HNSC
Patient stratification based on early integration with adaptive lasso-cox results





HNSC Heatmaps based on model's selected features







THANKS FOR YOUR ATTENTION







