



# Predicting Time to Ovarian Carcinoma Recurrence Using Protein Markers

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THE UNIVERSITY OF TEXAS

**MD Anderson  
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# Background

- Standard of care for ovarian cancer: resection, platinum based chemotherapy.
- For about 25% of patients, the tumor does not respond to platinum therapy.

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TCGA	RESISTANT	SENSITIVE	TOTAL
	90 (31.1%)	199 (68.9%)	289

- Identification of platinum resistant patients at time of diagnosis may result in a change in therapeutic approach.

# CLOVAR model predicts outcome in ovarian cancer

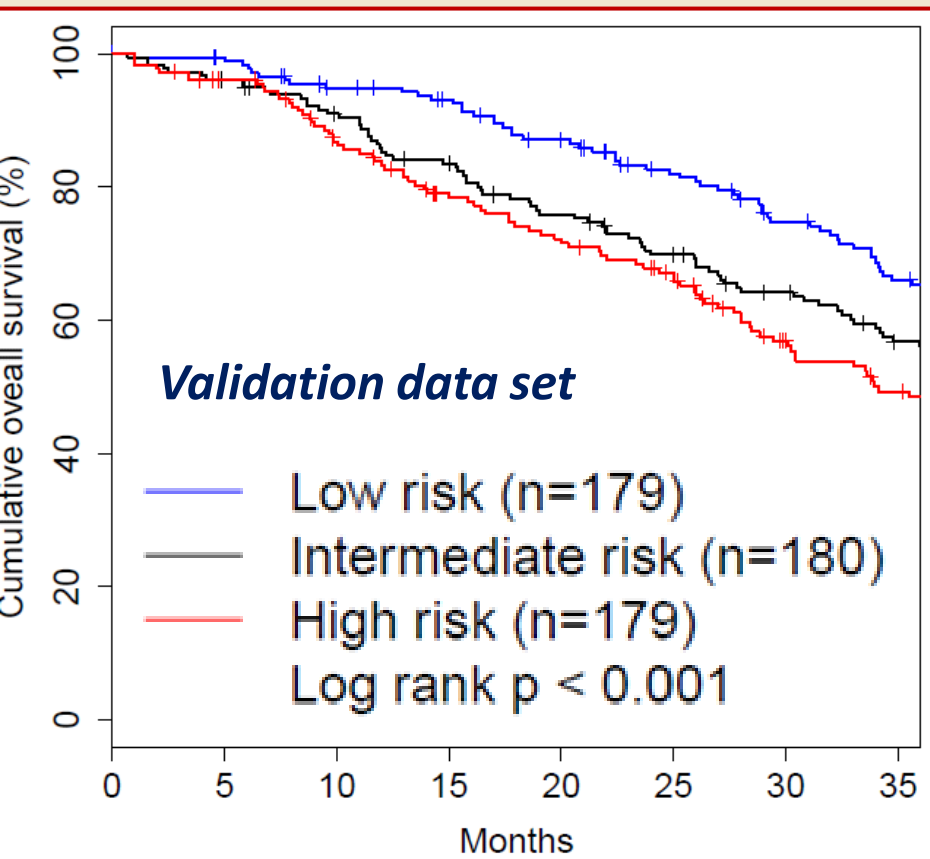


- ‘CLassification of OVARIan cancer’ (CLOVAR) predicts survival using gene expression signatures.
- Using TCGA, subtypes and survival gene expression signatures were identified, which would provide a prognostic model.

# CLOVAR model predicts outcome in ovarian cancer



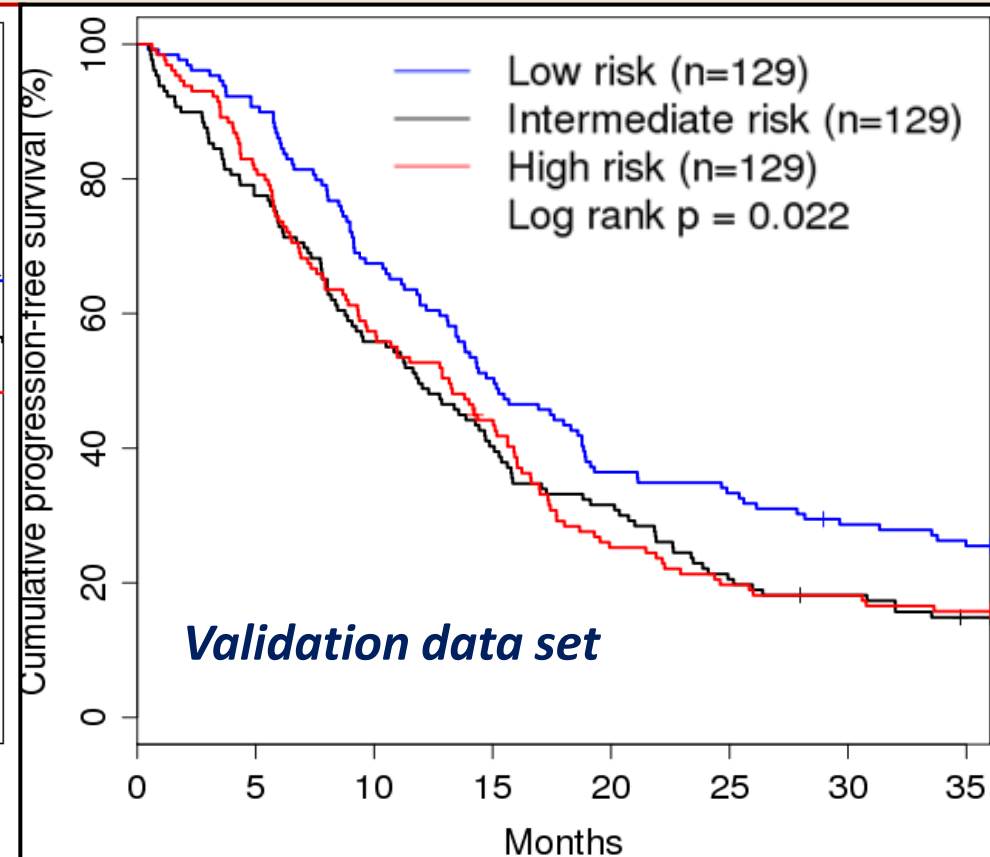
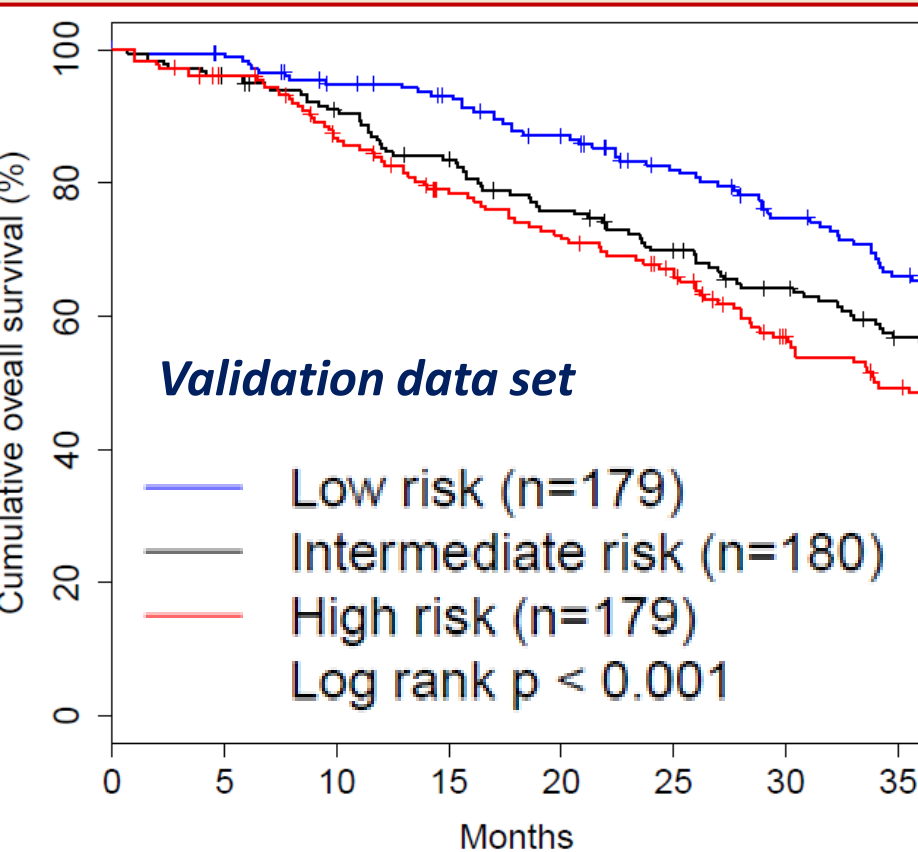
- ‘CLassification of OVARIan cancer’ (CLOVAR) predicts survival using gene expression signatures.



# CLOVAR is less effective to predict progression free survival



- ‘CLassification of OVARIan cancer’ (CLOVAR) predicts survival using gene expression signatures.



# Can we use protein markers to predict PFS?

- Aim to develop a predictor of platinum resistance that is based on **protein** markers.

- RPPA (reverse phase protein arrays):  
172 proteins and phosphoproteins in 412 TCGA samples with serous ovarian cancer.

**222** cases were included in the model construction (non-missing values for PFS, advanced stage)

# PROVAR

PRotein-driven index of OVARIan cancer

TCGA

Training set (n=222 samples)

LASSO

Selected 9 proteins most associated with PFS  
& estimated  $\beta$

PROVAR =

$$\hat{\beta}_1 X_1 + \hat{\beta}_2 X_2 + \dots + \hat{\beta}_9 X_9$$

# Feature Selection & Estimation



LASSO (Least Absolute Shrinkage and Selection Operator)

L1-constrained (lasso) Cox regression (Tibshirani, 1997), sparse interpretable models by shrinking some variables to exactly zero.



# Feature Selection & Estimation

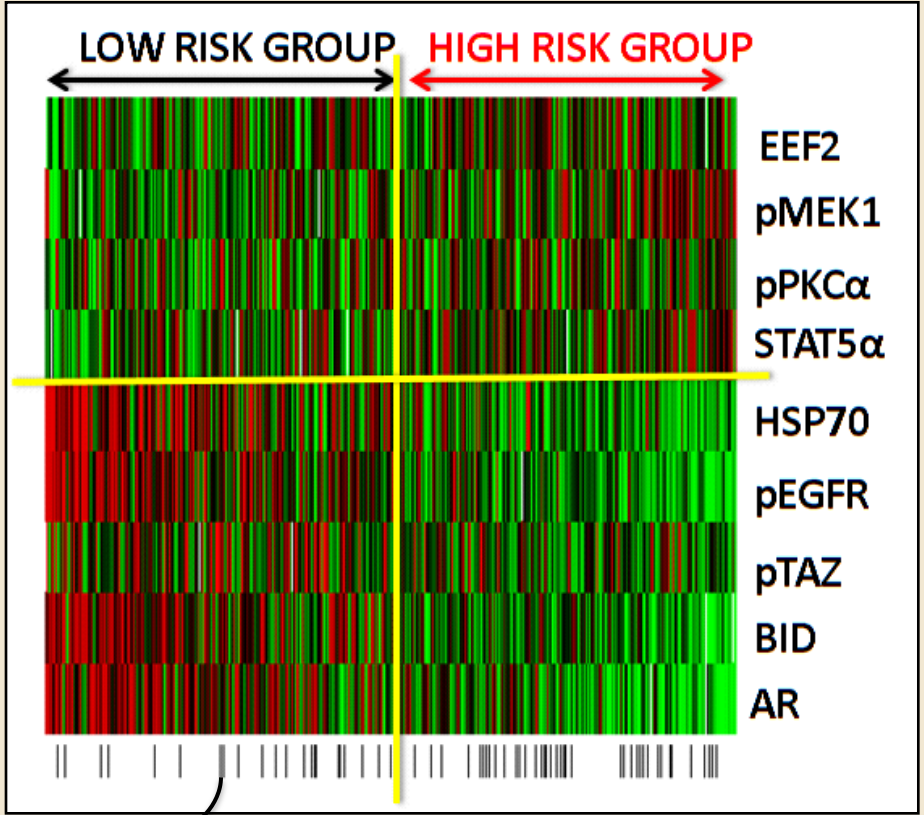


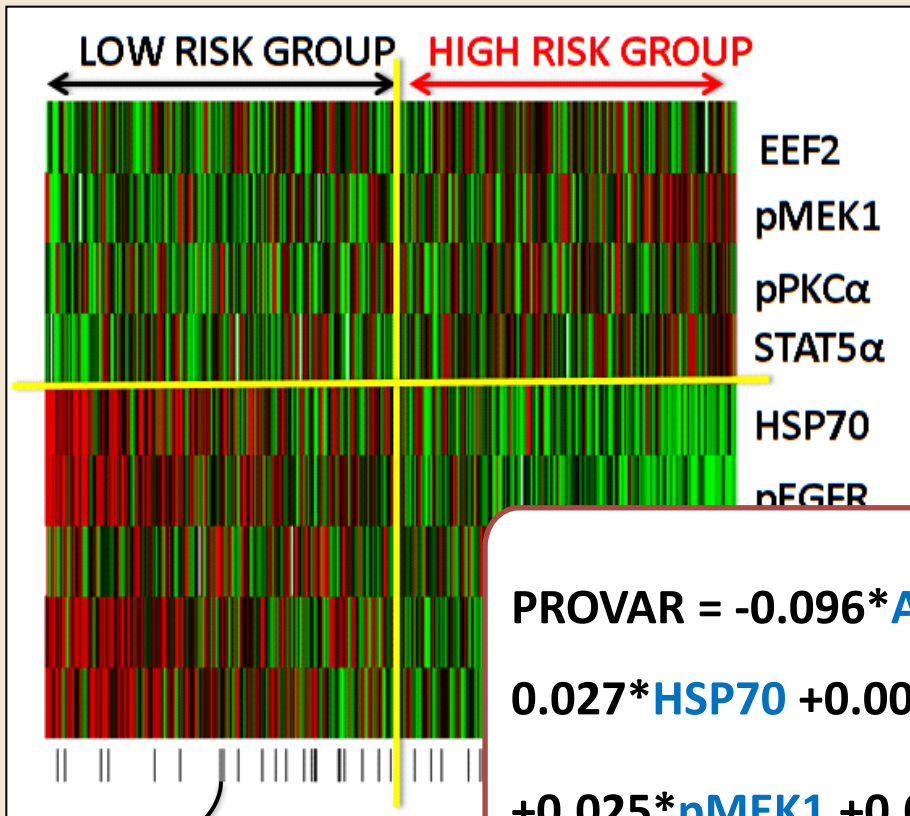
LASSO (Least Absolute Shrinkage and Selection Operator)

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AR	BID	pTAZ	pEGFR	HSP70	STAT5 $\alpha$	pPKC $\alpha$	pMEK1	EEF2
-0.096	-0.075	-0.058	-0.052	-0.027	0.004	0.022	0.025	0.029

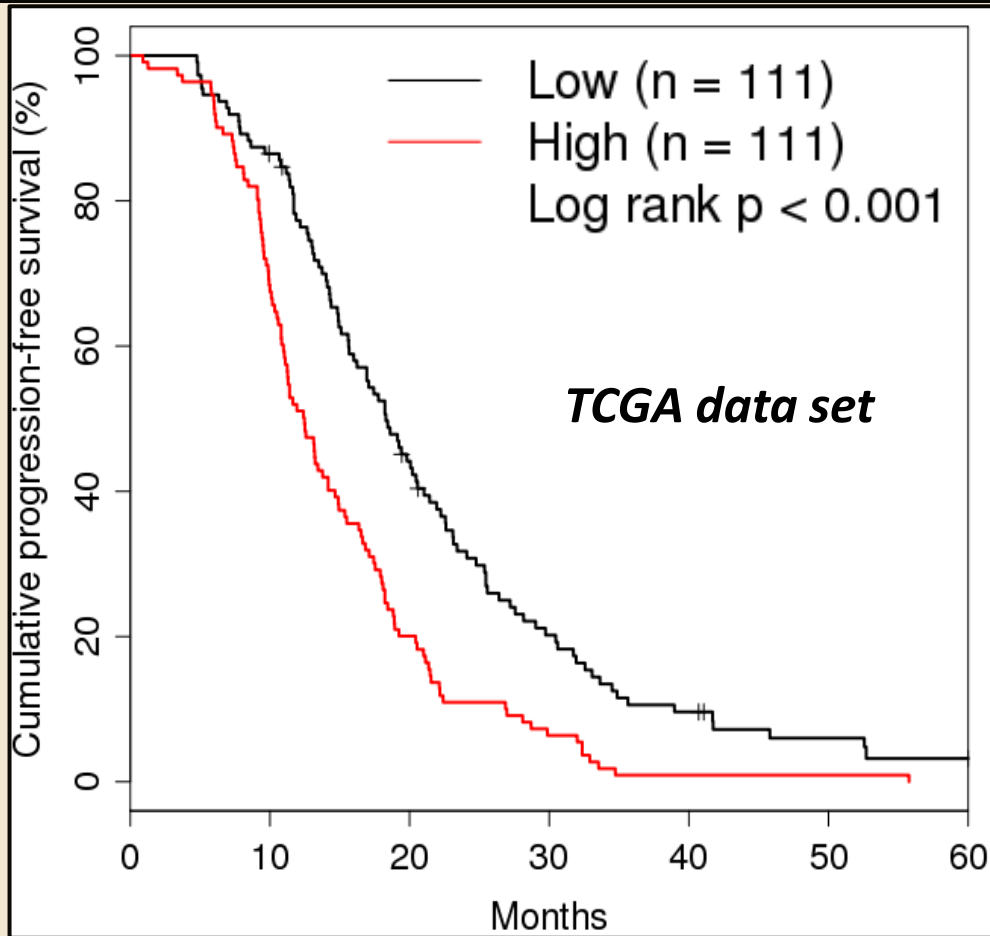
9 proteins most associated with progression-free survival.  
Numbers, lasso coefficients.



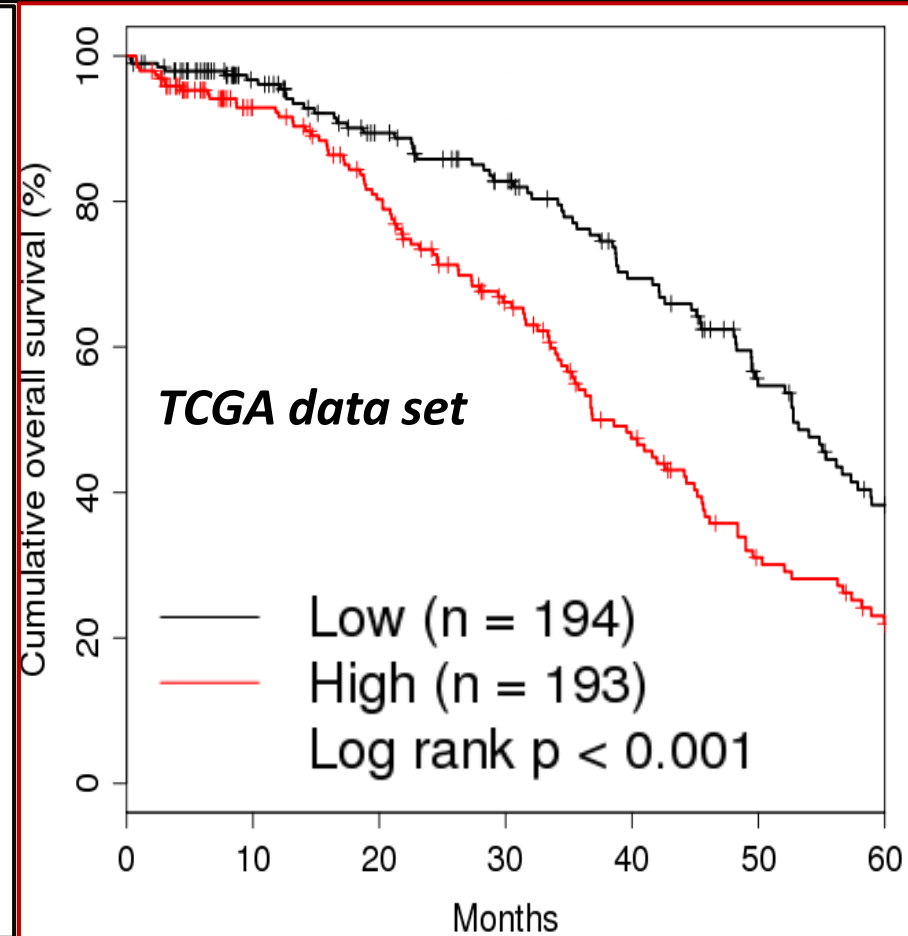


$$\begin{aligned}
 \text{PROVAR} = & -0.096 * \text{AR} - 0.075 * \text{BID} - 0.058 * \text{pTAZ} - 0.052 * \text{pEGFR} - \\
 & 0.027 * \text{HSP70} + 0.004 * \text{STAT5.ALPHA} + 0.022 * \text{pPKC.ALPHA} \\
 & + 0.025 * \text{pMEK1} + 0.029 * \text{EEF2}
 \end{aligned}$$

# PROVAR is predictive of both OS & PFS in the TCGA data set



Progression-free survival



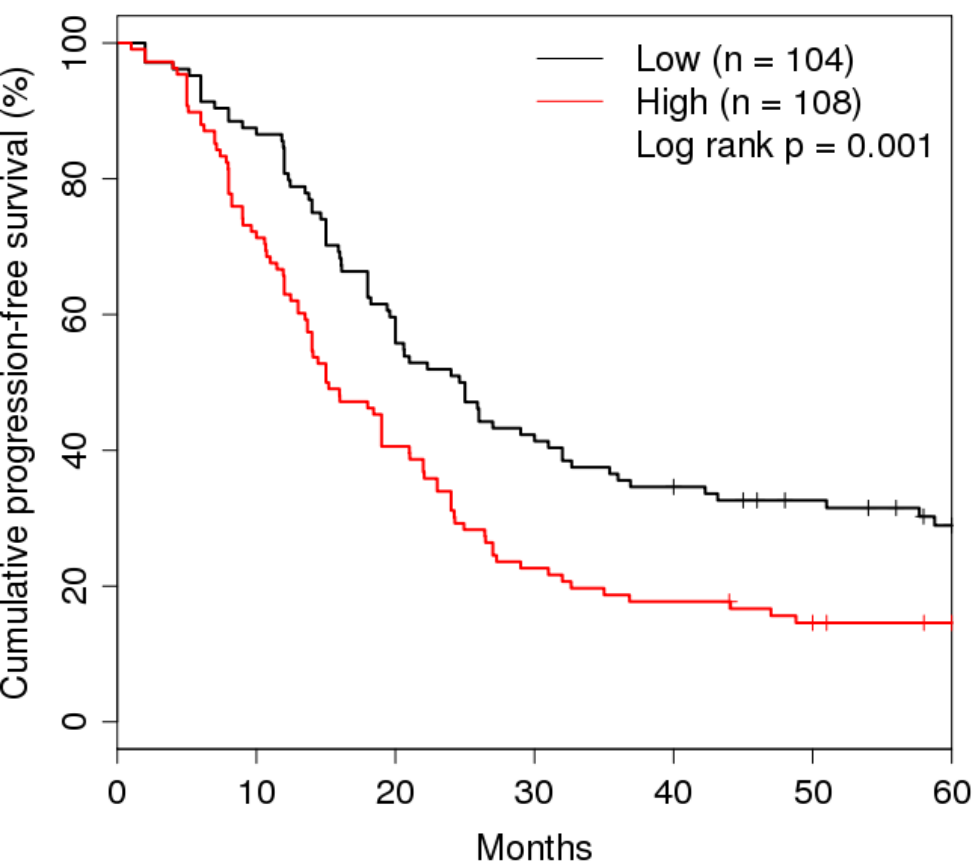
Overall survival

# Validation of PROVAR in an independent data set

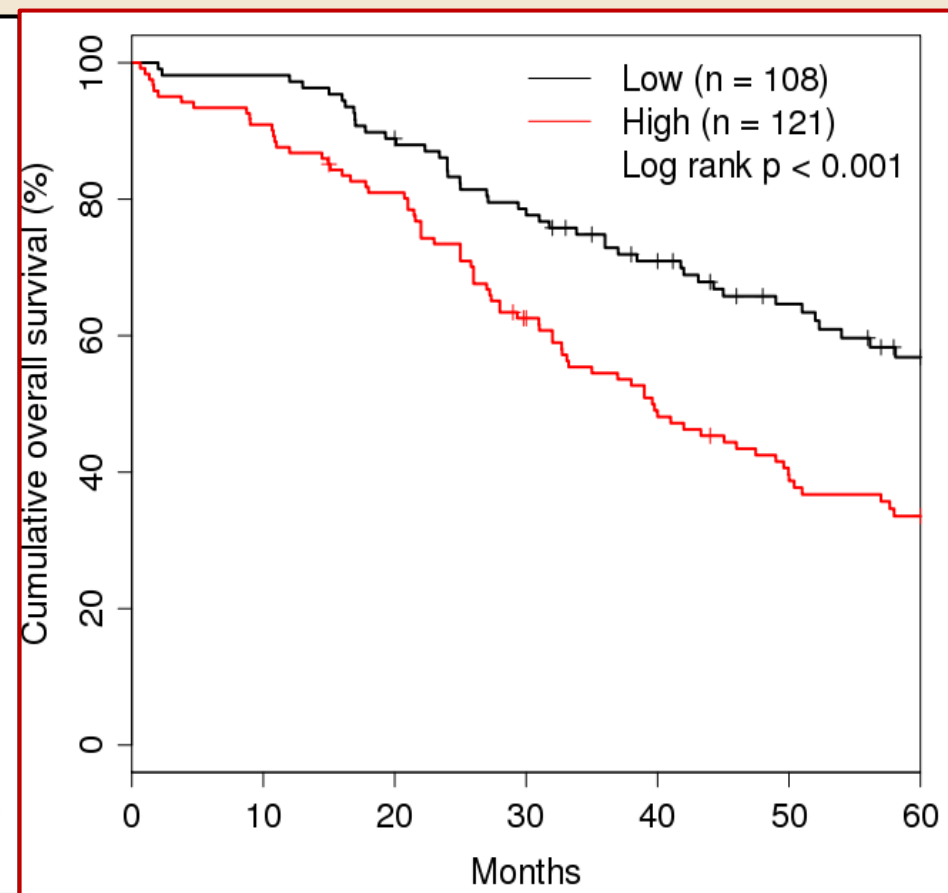


- 229 high-grade serous samples from Japan and Philadelphia.
- Expression levels of 144 proteins and phosphoproteins were measured by RPPA.

PROVAR is predictive of time to tumor recurrence in an independent dataset.



Progression-free survival



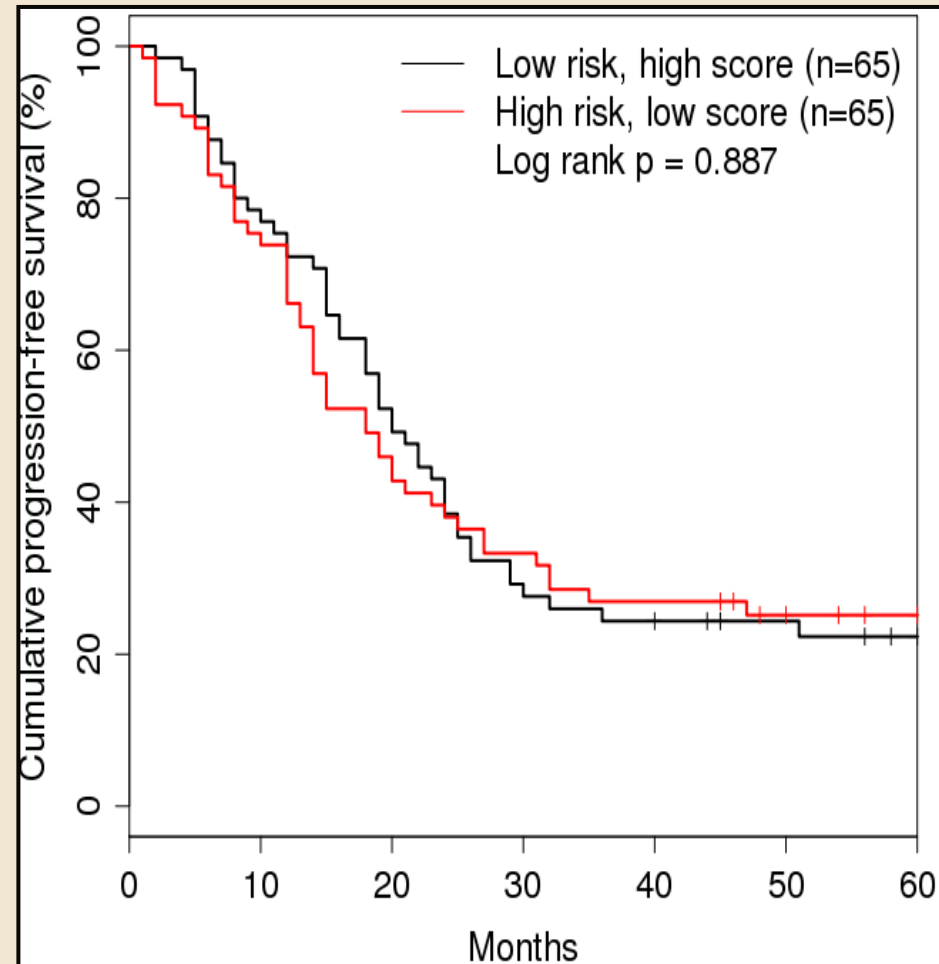
Overall survival

**PROVAR (protein-driven) vs. CLOVAR (gene-driven)**

**n=130 samples with gene expression data available**

# PROVAR (protein-driven) vs. CLOVAR (gene-driven): PFS, n=130 samples with gene expression data available

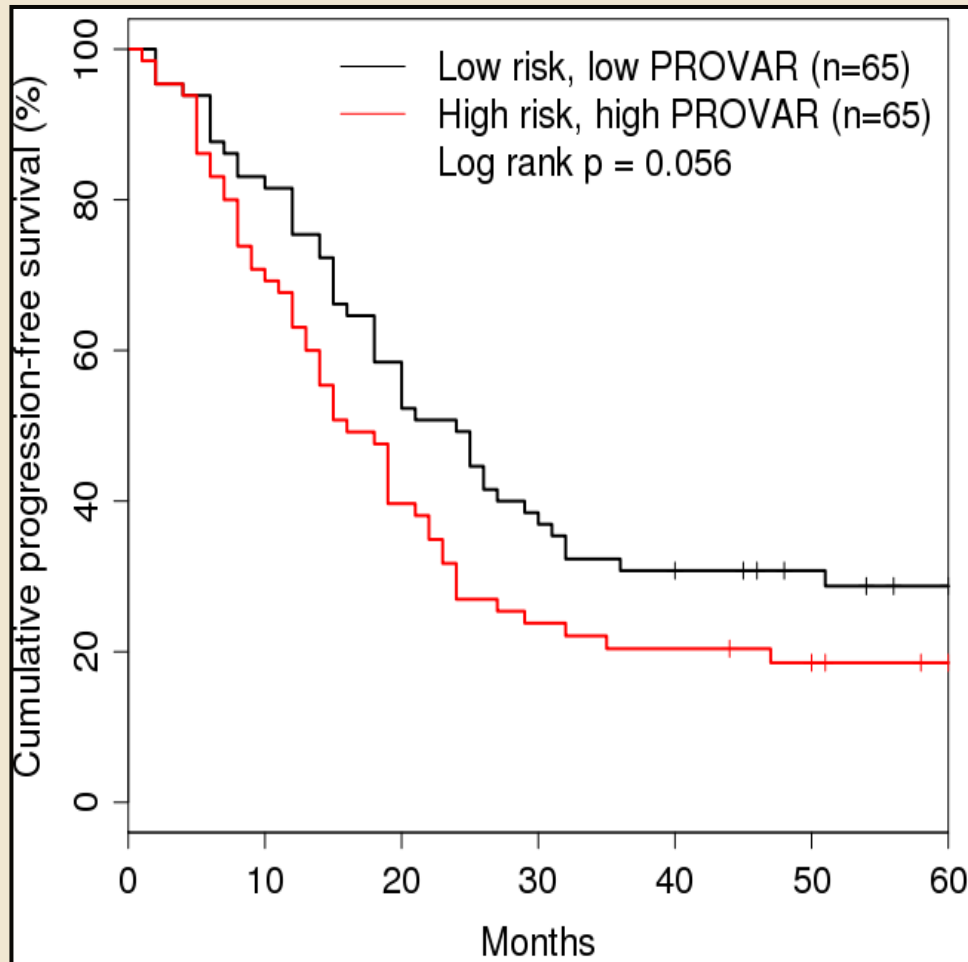
## CLOVAR



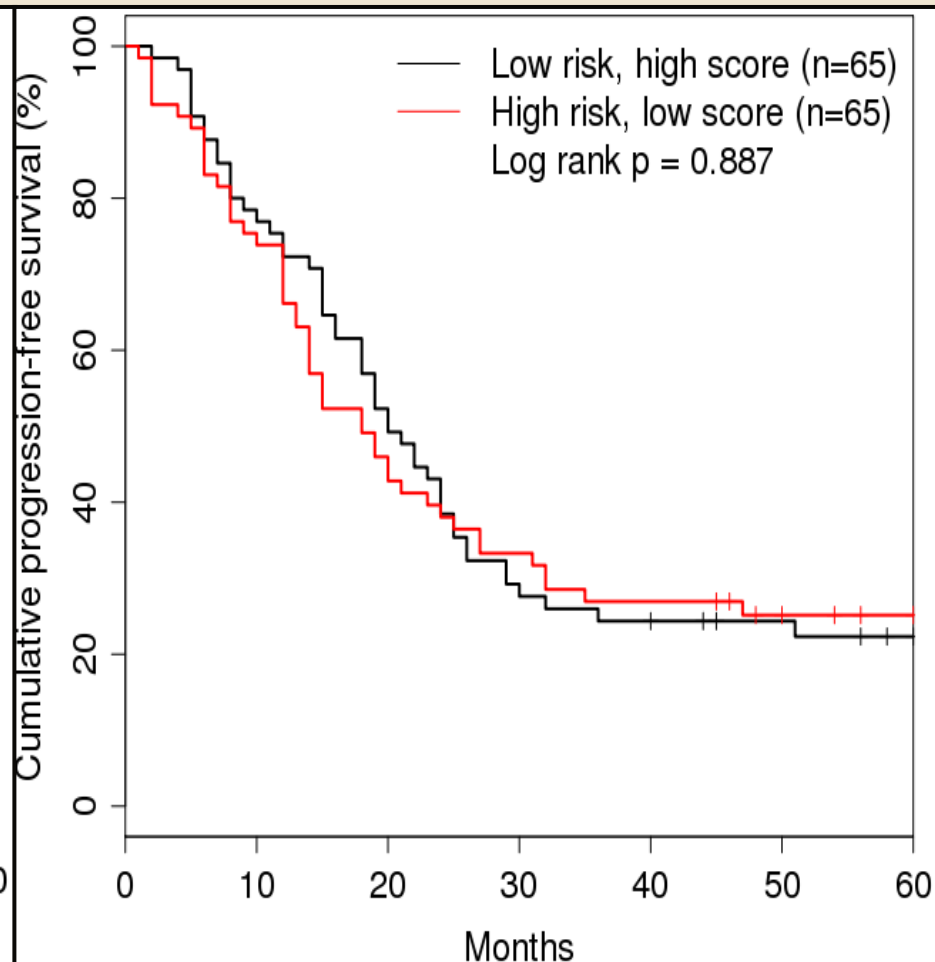


# PROVAR (protein-driven) vs. CLOVAR (gene-driven): PFS, n=130 samples with gene expression data available

## PROVAR



## CLOVAR



# **Robustness of the nine proteins markers**

# Robustness of proteins markers

Using the validation samples,

<b>COX2</b>	<b>VASP</b>	<b>CYCLINB1</b>	<b>pNFkB</b>	<b>AR</b>
<b>-0.006</b>	<b>0.057</b>	<b>0.050</b>	<b>-0.028</b>	<b>-0.014</b>

Numbers are lasso coefficients.

# Robustness of proteins markers

Using the validation samples,

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<b>COX2</b>	<b>VASP</b>	<b>CYCLINB1</b>	<b>pNFkB</b>	<b>AR</b>
<b>-0.006</b>	<b>0.057</b>	<b>0.050</b>	<b>-0.028</b>	<b>-0.014</b>

Using the TCGA samples,

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<b>BID</b>	<b>pEGFR</b>	<b>HSP70</b>	<b>EEF2</b>	<b>pPKC<math>\alpha</math></b>	<b>STAT5<math>\alpha</math></b>	<b>pTAZ</b>	<b>pMEK1</b>	<b>AR</b>
<b>-0.075</b>	<b>-0.052</b>	<b>-0.027</b>	<b>0.029</b>	<b>0.022</b>	<b>0.004</b>	<b>-0.058</b>	<b>0.025</b>	<b>-0.096</b>

Numbers are lasso coefficients.

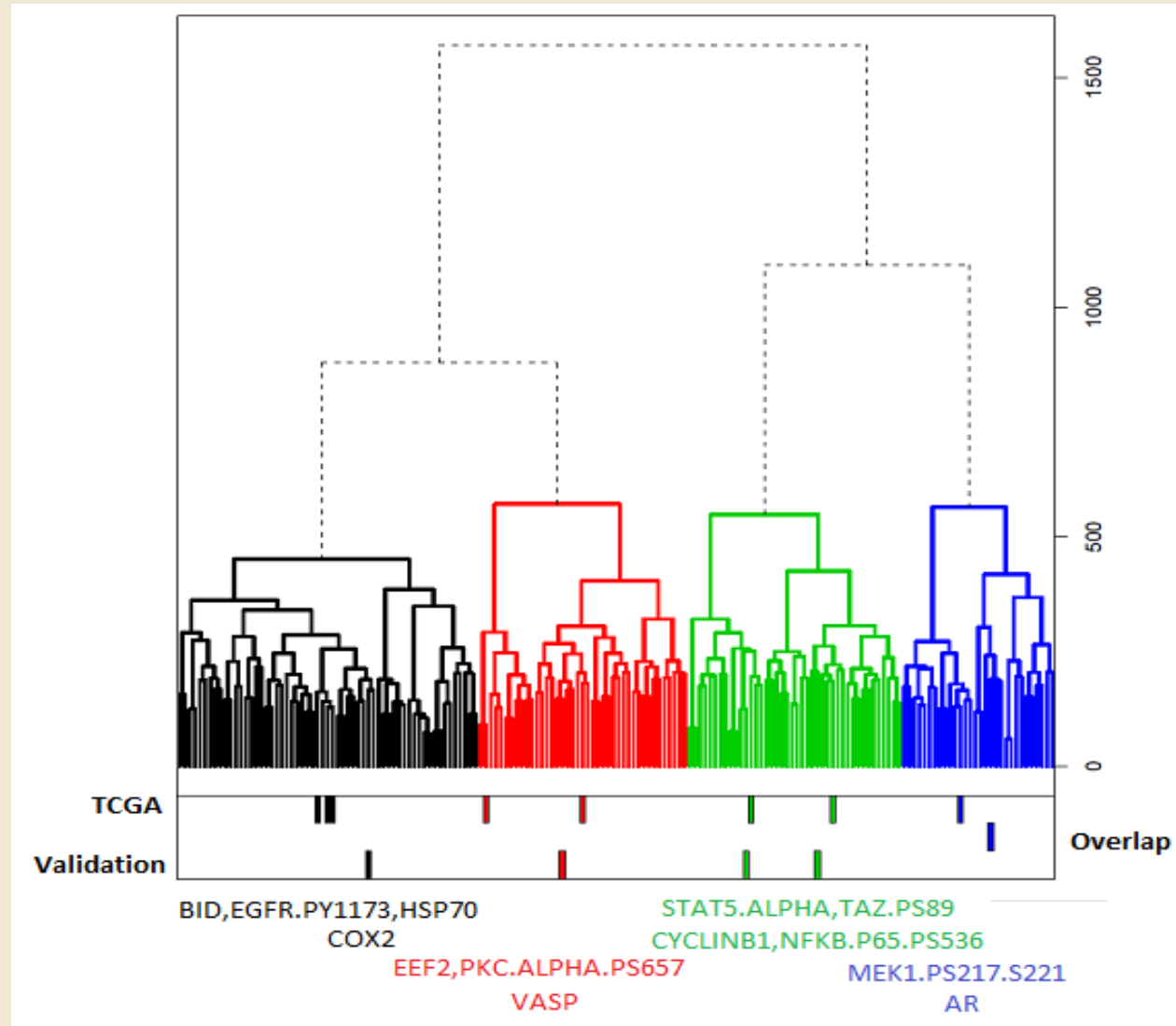
# Hierarchical clustering of 172 proteins from TCGA

**[Cluster1]**  
BID, pEGFR, HSP70  
COX2

**[Cluster2]**  
EEF2, pPKC $\alpha$   
VASP

**[Cluster3]**  
STAT5 $\alpha$ , pTAZ  
CYCLINB1, pNFKB

**[Cluster4]**  
pMEK1, AR  
(AR: overlap)



# Conclusions



- We developed a 'PRotein-driven index of OVARian cancer', PROVAR using progression-free survival,
- and successfully validated its discriminative ability to predict both progression-free and overall survival in high-grade serous ovarian cancers
- Unlike genetic signatures in previous studies that often contained a large number of genes, PROVAR is simple but still predictive of progression and survival, making it useful in clinical practice.

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- Collaborators

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