

# Supplementary Results

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April 19, 2007

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# 1 Survival data

## 1.1 Data collection and preprocessing

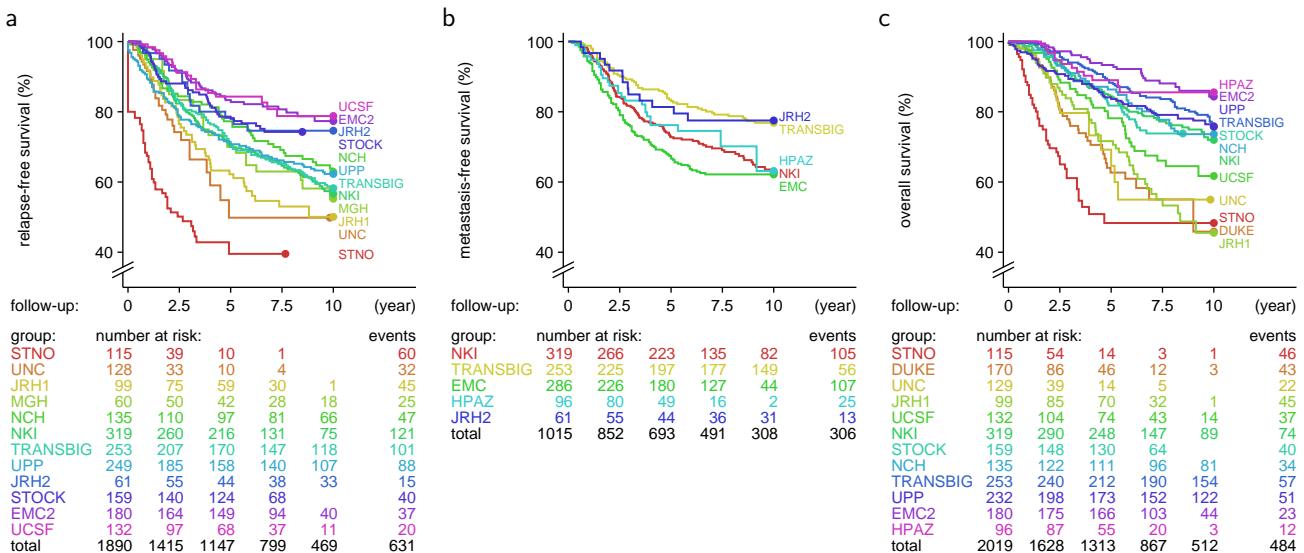
**Survival endpoints** The original studies often do not give precise description of their endpoints. The following are the definitions we used:

- **Relapse-free survival (RFS)** The endpoints are any events including local (in the same breast or the other side) or regional relapse, and distant metastasis (which is often implied by death). We also consider as RFS endpoints described as disease-free survival (DFS) in the original studies.
- **Metastasis-free survival (MFS)** The endpoints are relapses to other organs, such as liver or brain.
- **Overall survival (OS)** The endpoints are patient's deaths. Few studies specify "disease-specific survival" to distinguish death due to breast cancer, as opposed to any death. When not specified, we assume all deaths are disease related.

Note that these endpoint types are "nested". Death due to breast cancer implies distant metastasis shortly before (and often the time to metastasis is identical to the time to death). RFS typically includes MFS and OS. The events are therefore correlated. Although they obviously have different baseline hazard (RFS drops faster than MFS, which in turn drops faster than OS) (See Supplementary Result 1.2 below), most of our conclusions are concerned with relative survival between groups, which are valid for any type of events (Supplementary Result 1.2, 5.1, 5.2, 5.3, 6, 8). To save space (without affecting the conclusions), Figure 4c,d and Figure 6a,b,c in the main text are based on combined event types. Furthermore, we can not detect genes specifically associated with one type of endpoint but not the others (see Supplementary Result 7).

**Time-unit conversion and modifications** All time units were converted to "days". One year is considered to be 365.25 days. "Months" and "weeks" were converted to "years" first. Patients with missing time were excluded, but those with missing event indicator (very few) were considered censored. Median follow-up time can vary greatly between studies (e.g. very short in STNO and very long in TRANSBIG; see the numbers at risk in Supplementary Result 1.2). Therefore all survival data were truncated to 10 years (i.e. patients with longer follow-up are considered to be censored at this time point) in Cox regression analysis to ensure that the meta-analytical results pertain to similar time range in all datasets. The truncation also deals with problems with non-proportional hazard in long follow-up time (see Hilsenbeck *et al.* [1998] Time-dependence of hazard ratios for prognostic factors in primary breast cancer. *Breast Cancer Res. Treat.* **52**:227-237).

## 1.2 Cohort-specific survival for all endpoints



Cohort-specific survival curves for relapse-free survival (a), metastasis-free survival (b) and overall survival (c). Panel c is the same as figure 1b in the main text.

## 2 Coexpression modules

### 2.1 Gene Tables

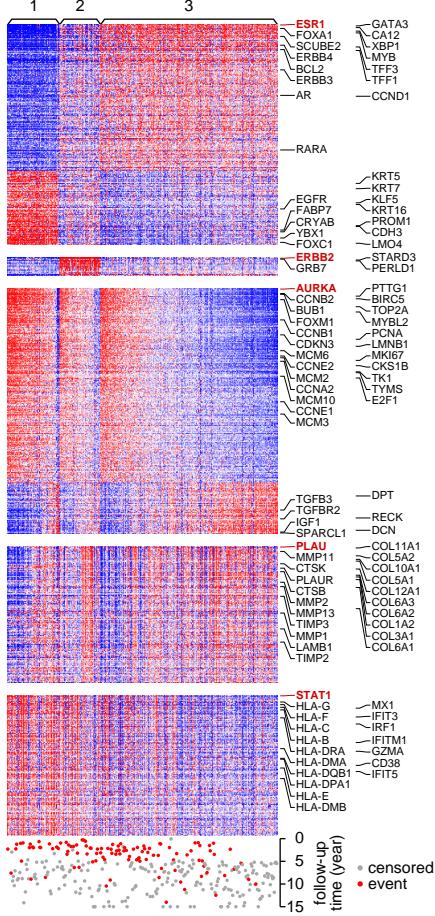
The genes identified by prototype-based coexpression analysis are listed in the file `modules.txt`, in a tab-delimited text file. For convenient browsing, open the file using spreadsheet program (e.g. MS Excel) and format the column widths to fit the contents automatically. We identified 909 genes under the specified criteria. The ordering of the genes in the table is the same as that in Figure 2b of the main text.

### 2.2 Number of genes selected and mapped to each dataset

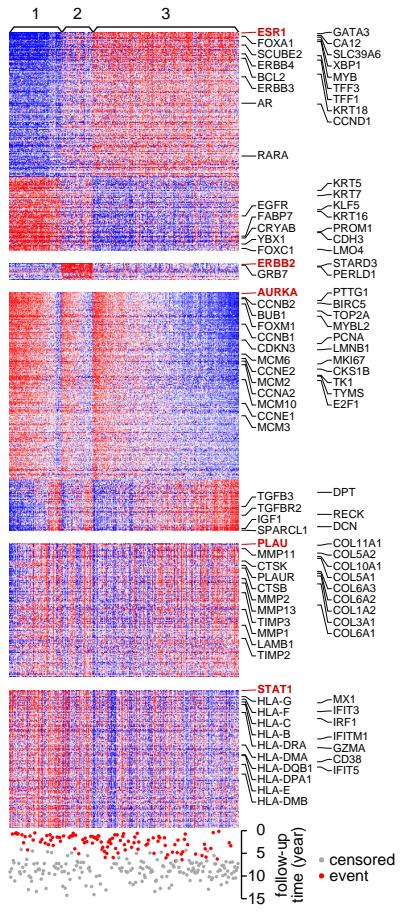
datasets	modules				
	ESR1	ERBB2	AURKA	PLAU	STAT1
NKI	250	20	278	154	159
EMC	248	18	270	151	156
UPP	268	21	294	160	166
STOCK	268	21	294	160	166
DUKE	182	14	204	130	134
UCSF	143	11	176	109	116
UNC	235	19	270	156	161
NCH	235	19	270	156	161
STNO	133	8	156	100	102
JRH1	100	8	135	90	99
JRH2	248	18	270	151	156
MGH	196	11	198	123	108
exp0	268	21	294	160	166
TGIF1	248	18	270	151	156
BWH	268	21	294	160	166
TRANSBIG	11	1	50	2	2
EMC2	1	1	10	1	0
HPAZ	2	1	13	1	0
Gene union	268	21	294	160	166

## 2.3 Heatmaps for all datasets

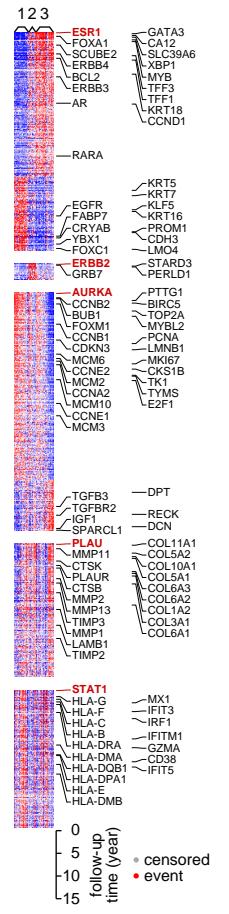
NKI



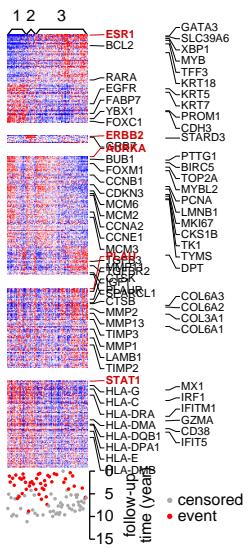
EMC



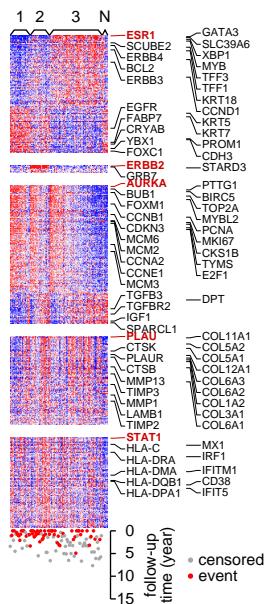
TGIF1



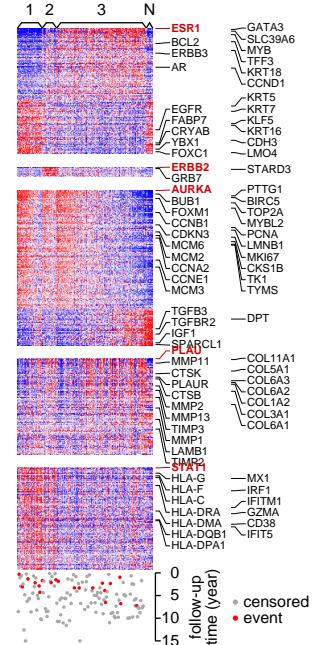
JRH1



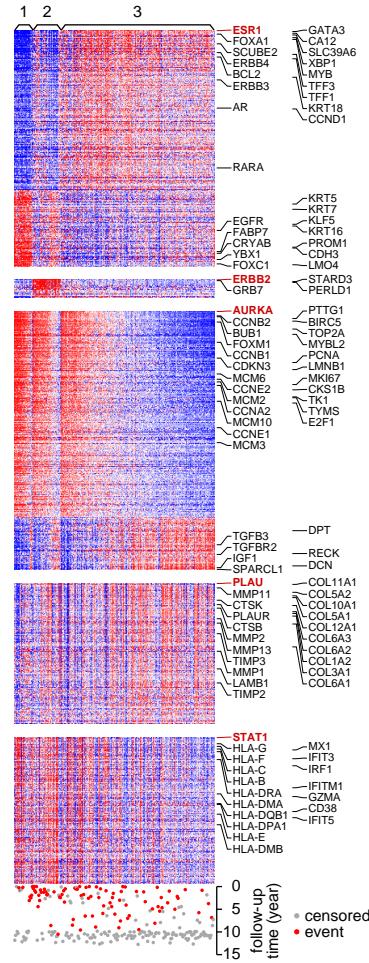
STNO



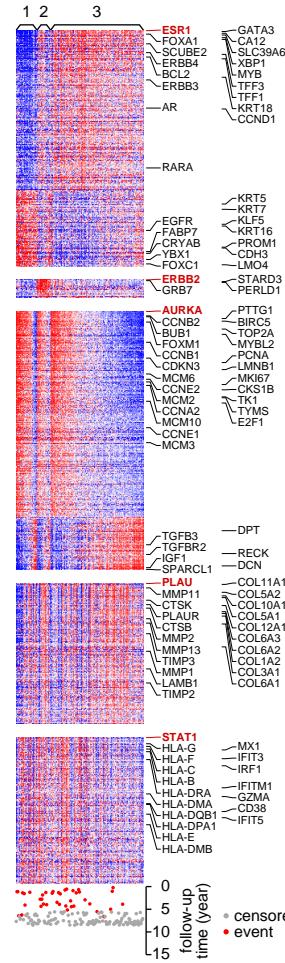
UCSF



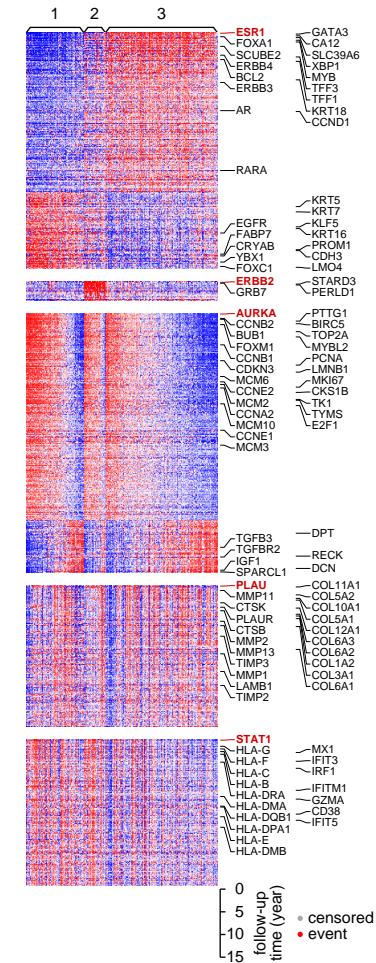
## UPP



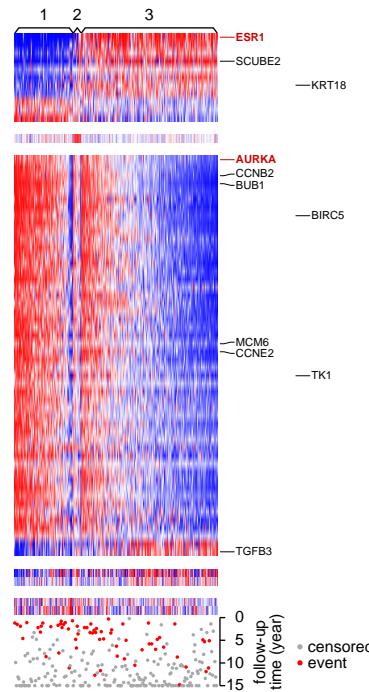
## STOCK



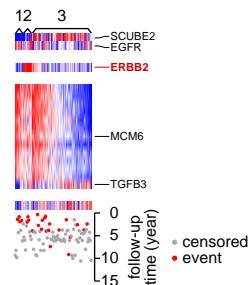
## expO



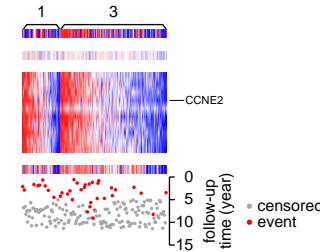
## TRANSBIG



## HPAZ



## EMC2



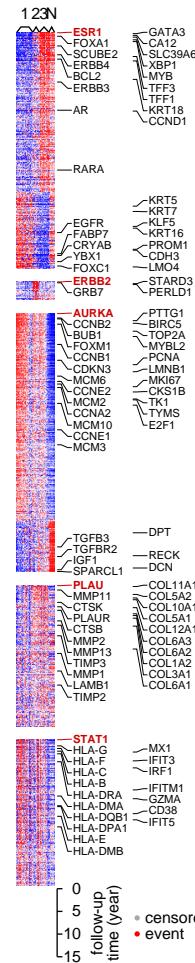
**Note** Because of the small number of genes mapped, the vertical scale of the heatmaps for TRANSBIG, HPAZ and EMC2 is larger than that of other datasets.

## BWH

UNC

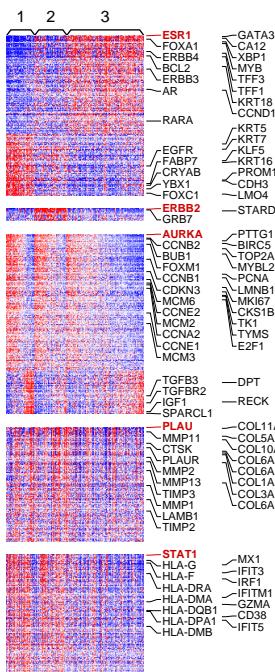
NCH

JRH2



DUKE

MGH



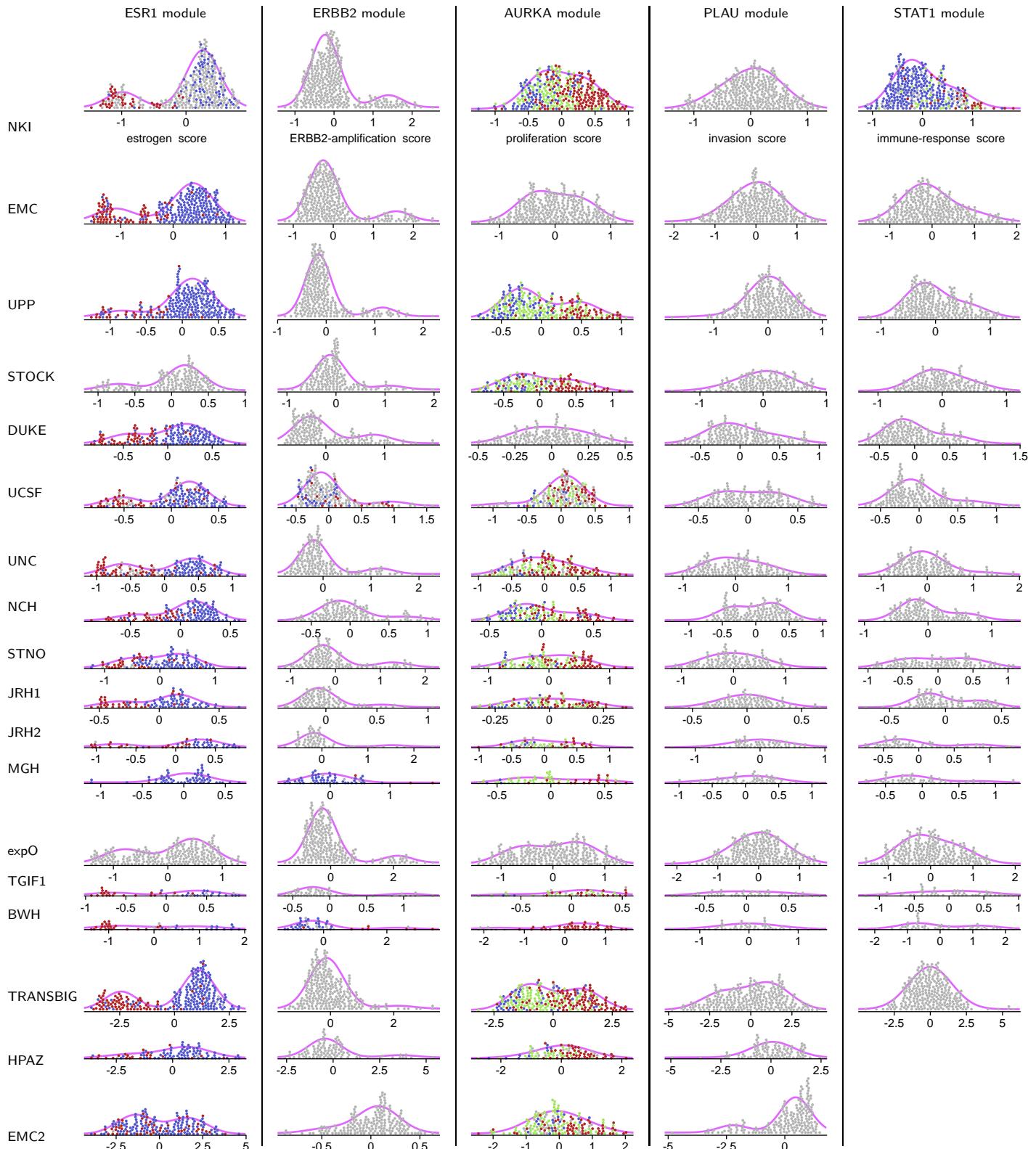
## 2.4 Biological annotation of the modules

The genes were divided into groups according to the module and the direction of the correlation (i.e., those correlated and anticorrelated with ESR1 were grouped separately; denoted by ESR1+ and ESR1-). Only ESR1 and AURKA modules have negatively correlated genes.

Each of these grouped was compared against the genesets in GO ontology database (based on annotation in Entrez Gene database tables from <ftp://ftp.ncbi.nih.gov/gene/>; version 21 January 2007) and Molecular Signature Database (MSigDB) version 2 from [http://www.broad.mit.edu/gsea/msigdb/msigdb\\_index.html](http://www.broad.mit.edu/gsea/msigdb/msigdb_index.html), Fisher's exact test was used to score the association between the modules and a gene set. The top-ranking genesets were shown in the file `go.txt` and `msigdb.txt`. The tables are in tab-delimited text format that can be opened from spreadsheet programs. The files contain Fisher's exact test p-values, odd-ratio of enrichment, geneset identifiers, and brief description.

### 3 Module score distributions

#### 3.1 Dot histograms



**Note** In datasets TRANSBIG, EMC2 and HPAZ (produced on custom diagnostic platforms), few genes were found for modules other than proliferation (AURKA), and therefore the module score might be unreliable. None can be found for STAT1-module for EMC2 and HPAZ.

### 3.2 Tests of bimodality

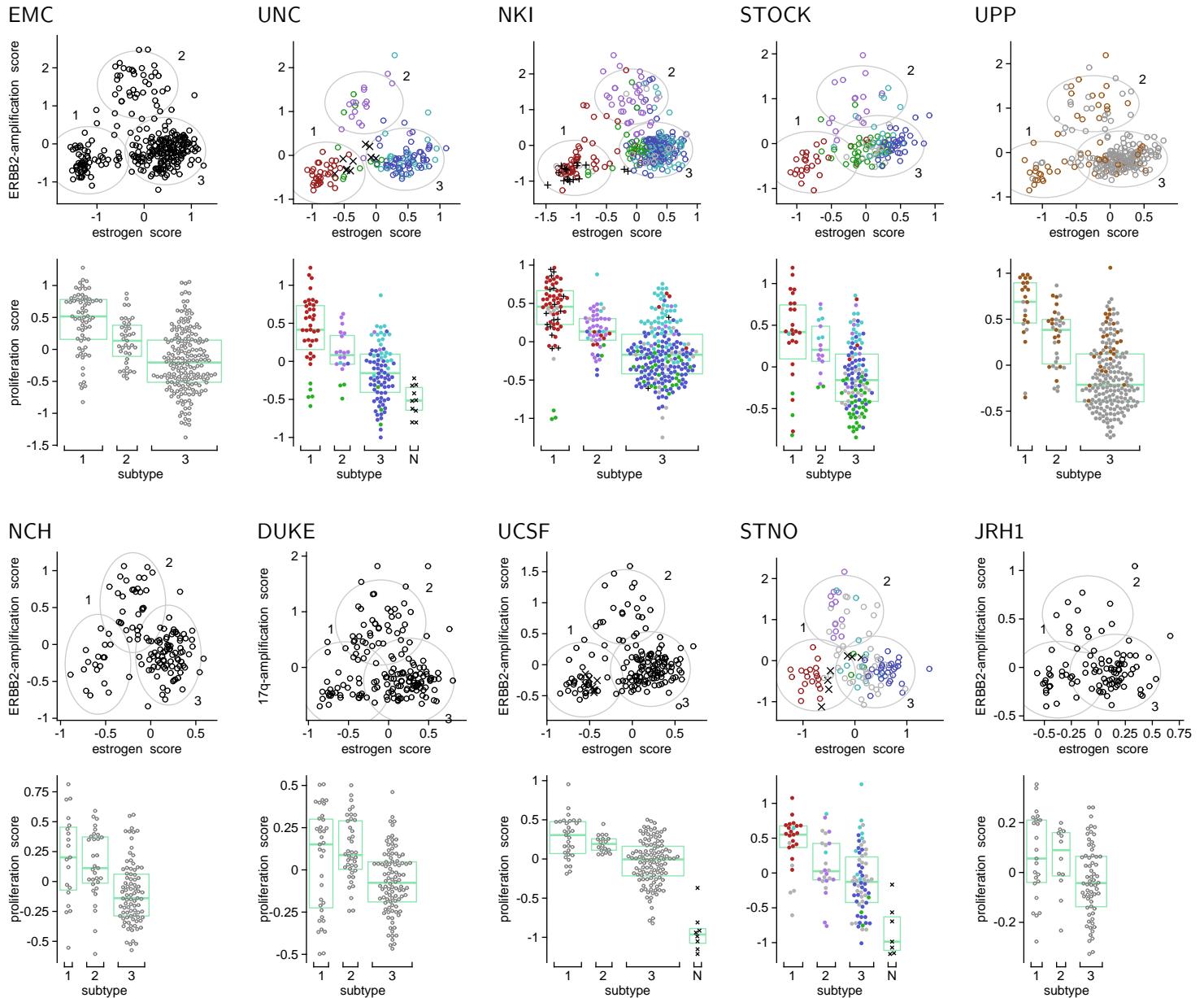
Log likelihood ratio between two- and one-component Gaussian mixture models were used as the test statistics, (as suggested in McLachlan and Peel [2000] *Finite Mixture Models*, Wiley, New York). However, the parametric bootstrap null distribution was generated from uniform distribution instead of unimodal normal distribution (following Hartigan [1985] “The dip test of unimodality.” *Ann. Statist.* **13**:30–84). This was done because the  $p$ -values based on unimodal-normal null distribution were too sensitive to non-normal, but still unimodal, densities.

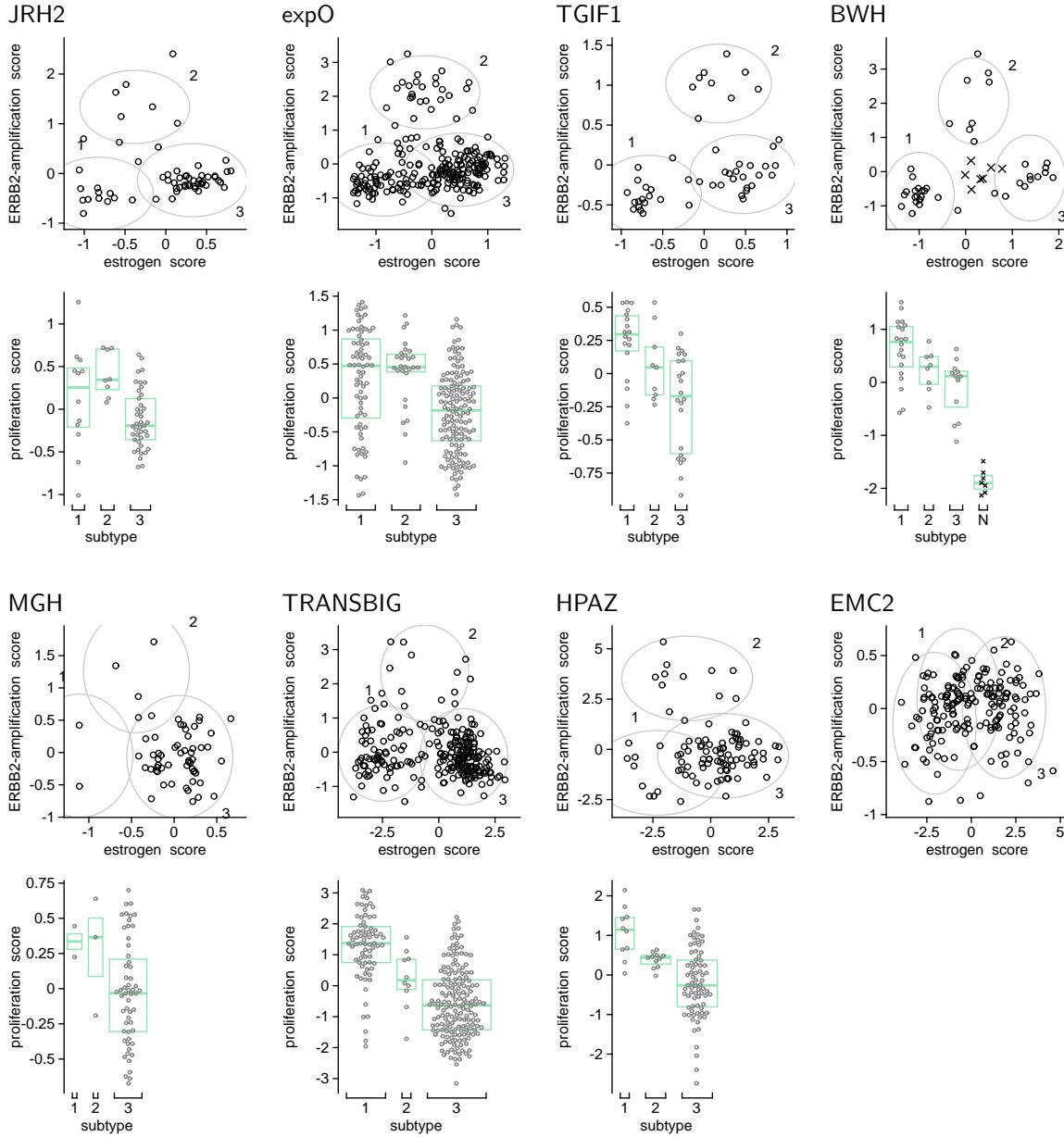
dataset	p-values of bimodality test of module scores				
	ESR1	ERBB2	AURKA	PLAU	STAT1
NKI	0.001 *	0.001 *	0.560	0.990	0.068
EMC	0.001 *	0.001 *	0.438	0.996	0.064
UPP	0.002 *	0.001 *	0.047	0.061	0.061
STOCK	0.001 *	0.001 *	0.083	0.083	0.083
DUKE	0.069	0.001 *	0.630	0.082	0.078
UCSF	0.003 *	0.001 *	0.055	0.473	0.054
UNC	0.005 *	0.001 *	0.146	0.087	0.085
NCH	0.040	0.047	0.097	0.295	0.083
STNO	0.689	0.001 *	0.412	0.573	0.571
JRH1	0.113	0.003 *	0.724	0.804	0.731
JRH2	0.003 *	0.001 *	0.119	0.531	0.010 *
MGH	0.064	0.076	0.929	0.965	0.070
exp0	0.021	0.001 *	0.673	0.574	0.481
TGIF1	0.877	0.001 *	0.088	0.875	0.919
BWH	0.927	0.001 *	0.020	0.926	0.889
TRANSBIG	0.001 *	0.001 *	0.715	0.108	0.072
EMC2	0.692	0.081	0.753	0.001 *	nan
HPAZ	0.108	0.001 *	0.109	0.097	nan

\*  $p \leq 0.01$

## 4 Three subtypes

### 4.1 Scatter plots of ERBB2 vs ESR1 module scores





## Notes

- Dataset BWH contains entirely of high-grade tumors according to pathological data. It does not have patient outcome data and therefore the difficulty in deciding high- or low-proliferation does not affect survival analysis.
- Dataset MGH contains only one ER-negative, 3 ERBB2+ and 3 low-grade tumors (out of 60) according to pathological data.
- Dataset EMC2 does not have enough genes to reliably determine the ER and ERBB2 module scores. The clusters can not be fitted well. In subsequent analysis, the pathological ER-status are used to assign the subtype (ER+ is considered type 3, ER- is type 1).

## 4.2 Tests of trimodality

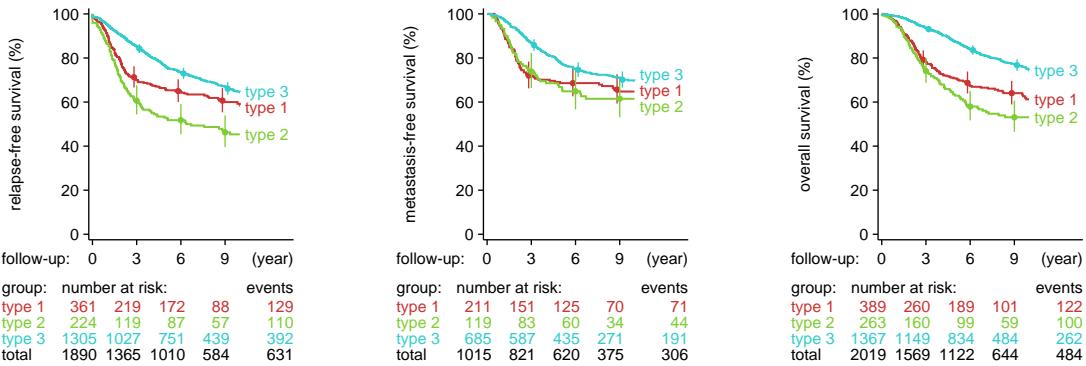
Log likelihood ratio between  $k$ - and  $(k-1)$ -component Gaussian mixture models were used as the test statistics, (as suggested in McLachlan and Peel [2000] *Finite Mixture Models*, Wiley, New York). Parametric bootstrap were used to general the null distribution, using the the parameters from the fitted null model.

p-values of tests		
for the number of components		
dataset	3 vs 2	4 vs 3
NKI	0.001 *	0.978
EMC	0.001 *	0.024
UPP	0.001 *	0.325
STOCK	0.001 *	0.014
DUKE	0.002 *	0.007 *
UCSF	0.001 *	0.012
UNC	0.001 *	0.054
NCH	0.001 *	0.589
STNO	0.001 *	0.026
JRH1	0.002 *	0.030
JRH2	0.002 *	0.088
MGH	0.932	0.671
exp0	0.001 *	0.002 *
TGIF1	0.001 *	0.268
BWH	0.001 *	0.025
TRANSBIG	0.001 *	0.035
EMC2	0.091	0.007 *
HPAZ	0.001 *	0.905

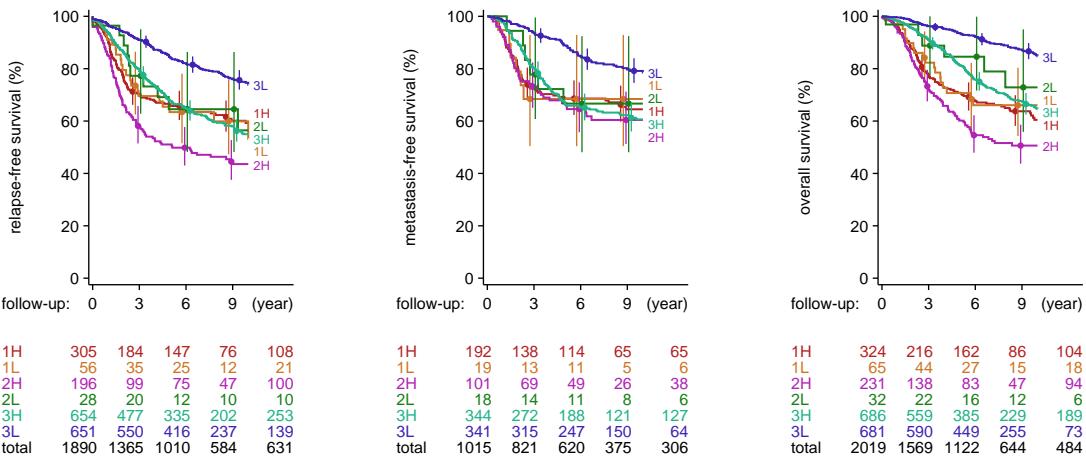
\* p <= 0.01

## 5 Additional survival analyses

### 5.1 Three subtypes

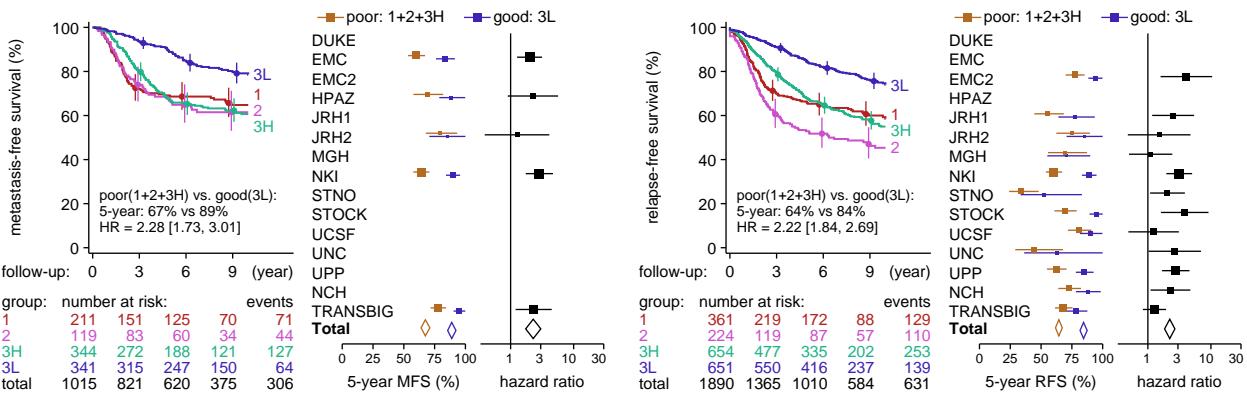


### 5.2 Three subtypes combined with high/low proliferation



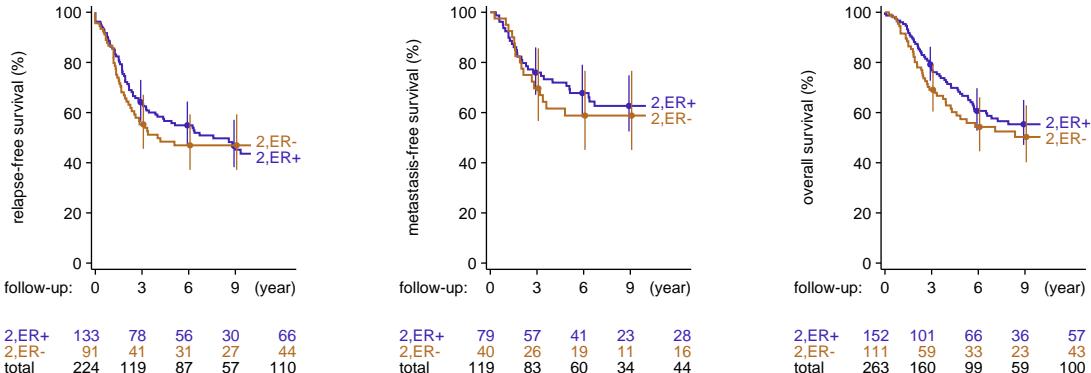
### 5.3 Separate metastasis-free survival and relapse-free survival

Figure 4c,d use metastasis-free survival, if available, or relapse-free survival otherwise. Below are the results for the two types of endpoints separately.



## 6 Prognostic value of ER-status within type-2 (ERBB2+) tumors

Because pathological ER status is not available for all patients, here ER-status is based on estrogen scores (using as cutoff the midpoint between the cluster center of type 1 and type 3).



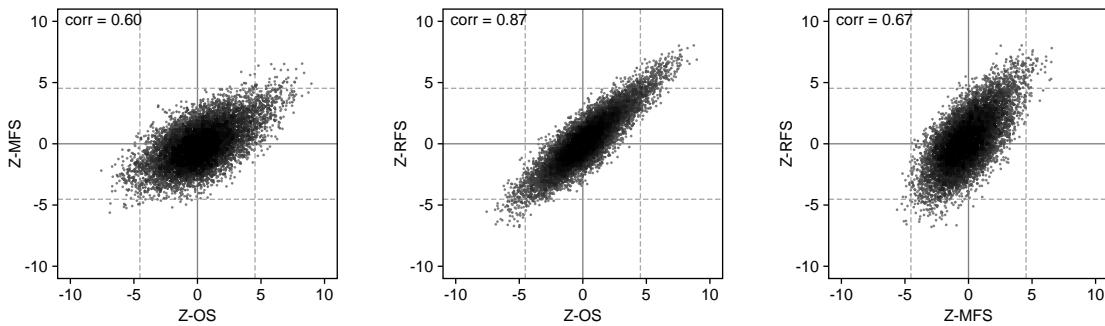
## 7 Gene-by-gene survival analysis

### 7.1 Gene Tables

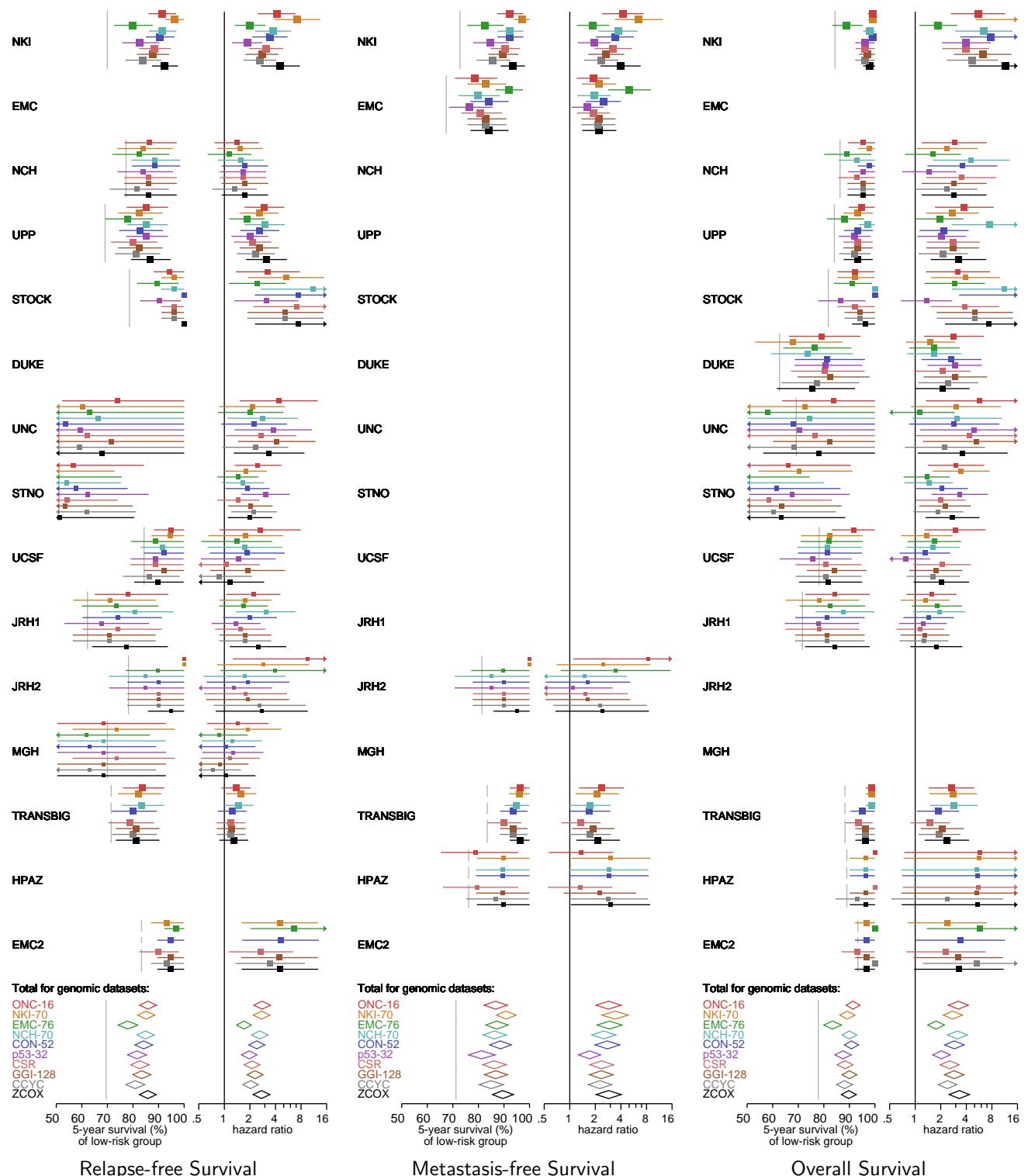
The top-ranking genes identified by meta-analytical gene-by-gene Cox regression are listed in the file `zsurvival.txt`, in a tab-delimited text file. For convenient browsing, open in spreadsheet program (e.g. MS Excel) and format the column widths to fit the contents automatically.

### 7.2 Agreement of gene associations with different survival endpoints

Here we show that the choice of endpoints (OS: overall survival, MFS: metastasis-free survival, RFS: relapse-free survival) do not substantially affect gene-by-gene survival analysis. The lower Z-scores of MFS is due mainly to smaller sample size (MFS data are available only in 5 datasets, see Figure 1a of the main text).



## 8 Forest plots of signature performance for all survival endpoints



## 9 Concordance in risk classifications

### 9.1 Tables of pairwise concordance

Whole signatures:

	NKI-70	ONC-16	EMC-76	NCH-70	CON-52	p53-32	GGI-128	CSR	CCYC	ZCOX
NKI-70	*	79.97	69.37	77.54	81.02	76.07	82.18	75.12	81.34	83.66
ONC-16	79.97	*	70.22	82.50	83.87	74.06	86.51	79.02	84.08	86.72
EMC-76	69.37	70.22	*	70.11	73.59	61.04	73.38	72.11	73.27	73.48
NCH-70	77.54	82.50	70.11	*	85.45	67.84	86.29	82.39	83.13	85.66
CON-52	81.02	83.87	73.59	85.45	*	69.11	91.04	82.29	88.82	91.14
p53-32	76.07	74.06	61.04	67.84	69.11	*	71.64	67.21	69.74	72.69
GGI-128	82.18	86.51	73.38	86.29	91.04	71.64	*	83.34	92.62	93.46
CSR	75.12	79.02	72.11	82.39	82.29	67.21	83.34	*	82.81	83.87
CCYC	81.34	84.08	73.27	83.13	88.82	69.74	92.62	82.81	*	91.25
ZCOX	83.66	86.72	73.48	85.66	91.14	72.69	93.46	83.87	91.25	*

Average pairwise concordance: 79.4% (84.5% if EMC-76 and p53-32 are excluded)

Proliferation-gene partial signatures:

	ONC-16	NKI-70	EMC-76	NCH-70	CON-52	p53-32	CSR	GGI-128	CCYC	ZCOX
ONC-16	*	84.71	82.39	85.35	89.46	84.92	84.40	90.72	89.77	90.20
NKI-70	84.71	*	83.24	83.45	88.09	80.50	83.34	88.51	88.72	89.67
EMC-76	82.39	83.24	*	84.61	86.93	78.39	83.13	86.29	86.72	85.56
NCH-70	85.35	83.45	84.61	*	87.45	81.13	87.66	87.56	86.51	87.24
CON-52	89.46	88.09	86.93	87.45	*	83.66	86.72	93.15	91.78	92.30
p53-32	84.92	80.50	78.39	81.13	83.66	*	81.34	85.98	84.29	84.40
CSR	84.40	83.34	83.13	87.66	86.72	81.34	*	86.61	86.19	87.45
GGI-128	90.72	88.51	86.29	87.56	93.15	85.98	86.61	*	94.83	94.52
CCYC	89.77	88.72	86.72	86.51	91.78	84.29	86.19	94.83	*	93.67
ZCOX	90.20	89.67	85.56	87.24	92.30	84.40	87.45	94.52	93.67	*

Average pairwise concordance: 86.7%

Non-proliferation-gene partial signatures:

	NKI-70	ONC-16	EMC-76	NCH-70	CON-52	p53-32	GGI-128	CSR	CCYC	ZCOX
NKI-70	*	67.00	59.30	65.21	62.78	69.43	69.30	65.42	62.05	73.22
ONC-16	67.00	*	58.36	63.73	60.15	68.90	65.29	62.68	59.20	70.48
EMC-76	59.30	58.36	*	59.52	60.46	54.98	61.85	64.47	56.88	57.62
NCH-70	65.21	63.73	59.52	*	66.16	58.78	64.18	65.74	56.04	65.00
CON-52	62.78	60.15	60.46	66.16	*	58.99	61.40	60.46	58.88	62.99
p53-32	69.43	68.90	54.98	58.78	58.99	*	65.07	61.52	58.67	69.74
GGI-128	69.30	65.29	61.85	64.18	61.40	65.07	*	65.74	57.40	69.97
CSR	65.42	62.68	64.47	65.74	60.46	61.52	65.74	*	62.05	67.53
CCYC	62.05	59.20	56.88	56.04	58.88	58.67	57.40	62.05	*	63.42
ZCOX	73.22	70.48	57.62	65.00	62.99	69.74	69.97	67.53	63.42	*

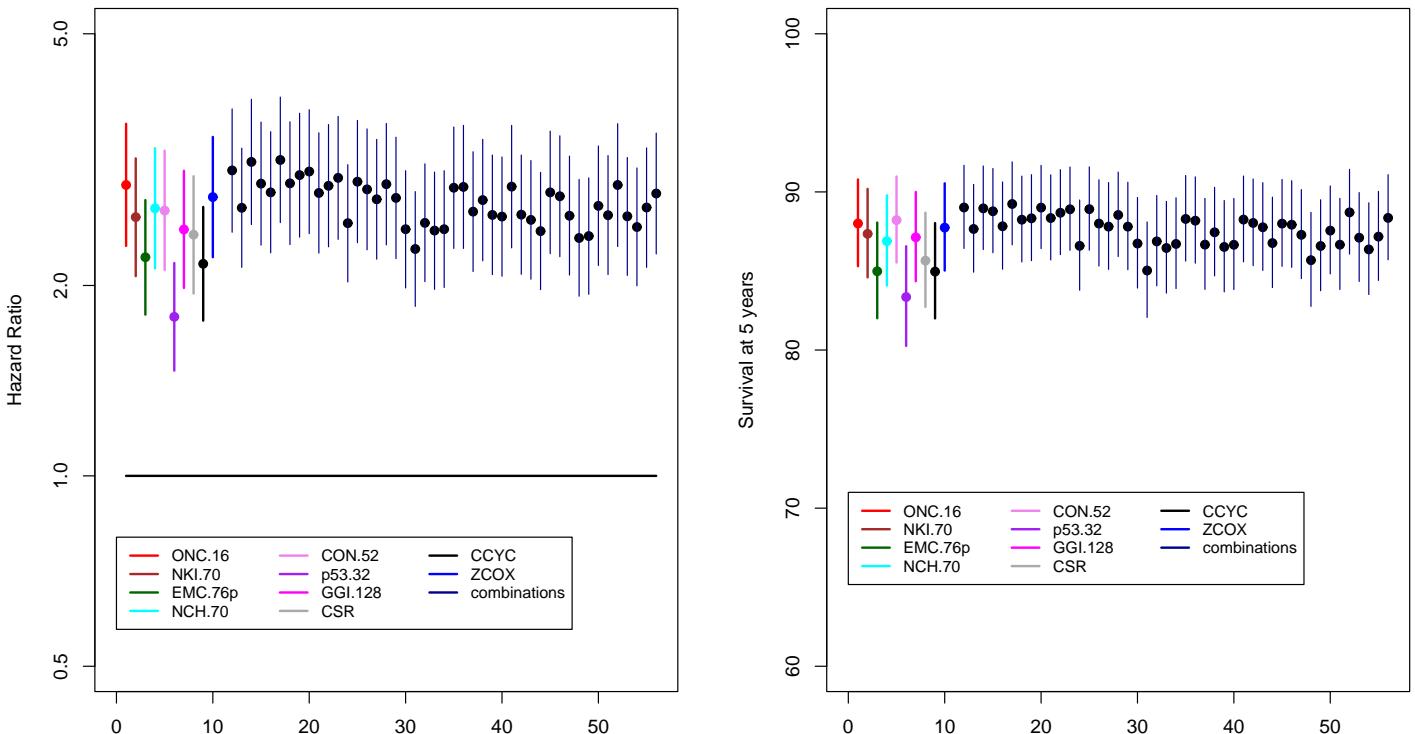
Average pairwise concordance: 63.1%

## 9.2 Combined prediction by pairs of signatures

Every distinct pair of signatures were combined in the following way:

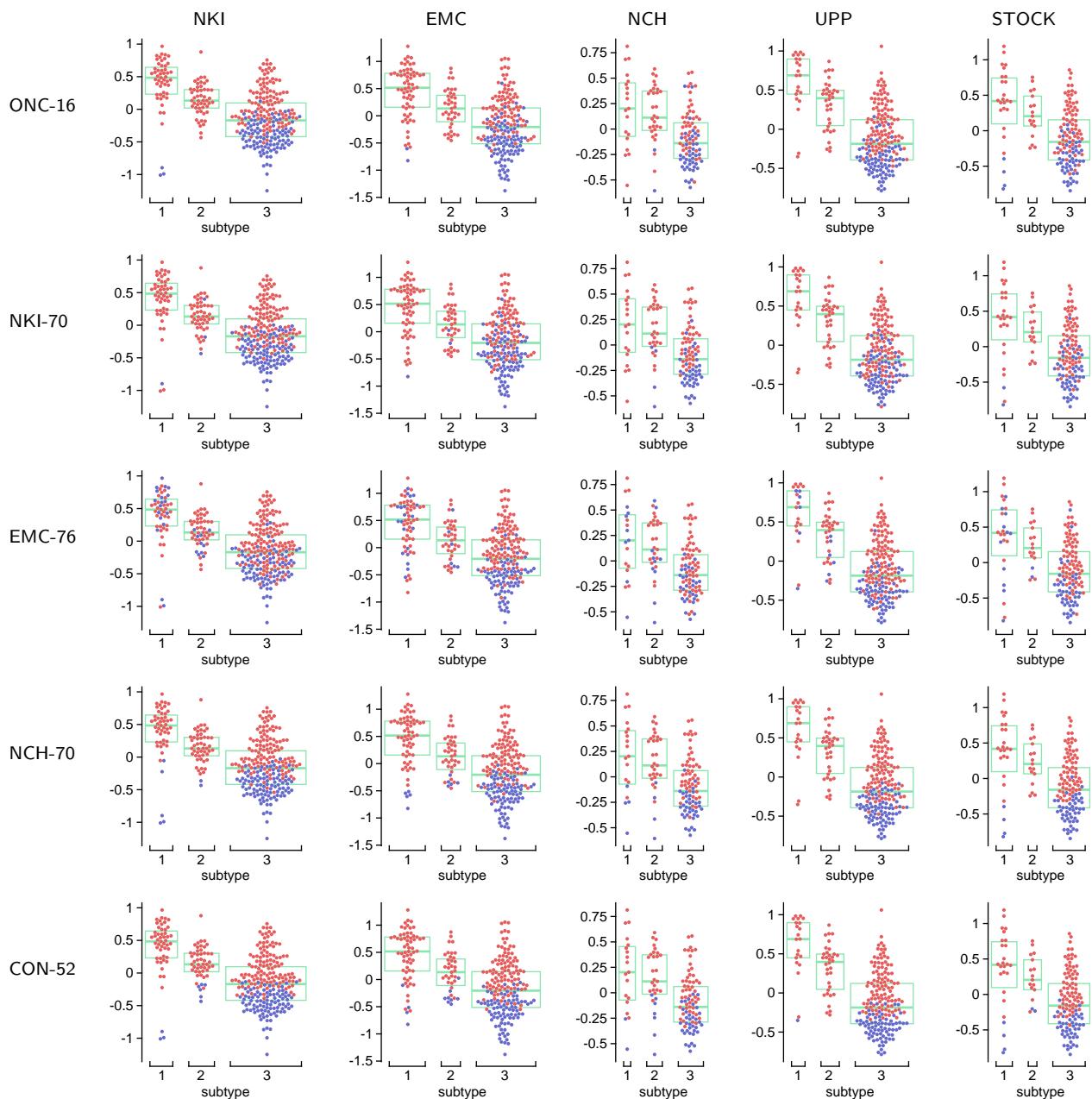
1. Their continuous prediction scores were used as explanatory variables in Cox regression (stratified by dataset).
2. The fitted coefficients were used as weights in linear combination of the two prediction scores to produce the combined prediction score
3. The combined scores were used to rank patients. 33% in each dataset were classified as low-risk.

The results for the 45 pairs are shown below as the black bars to the right of the individual signatures. The identity of each combination is not shown (the ordering from left to right is [ONC-16 + NKI-70], [ONC-16 + EMC-76p], and so on), because we only want to demonstrate that their hazard ratios and survivals are similar and within the confidence interval of the some individual signatures. Note that the combined performance may be slightly biased upward because the weights are estimated from the same data. Even so, the improvement is not clinically substantial.

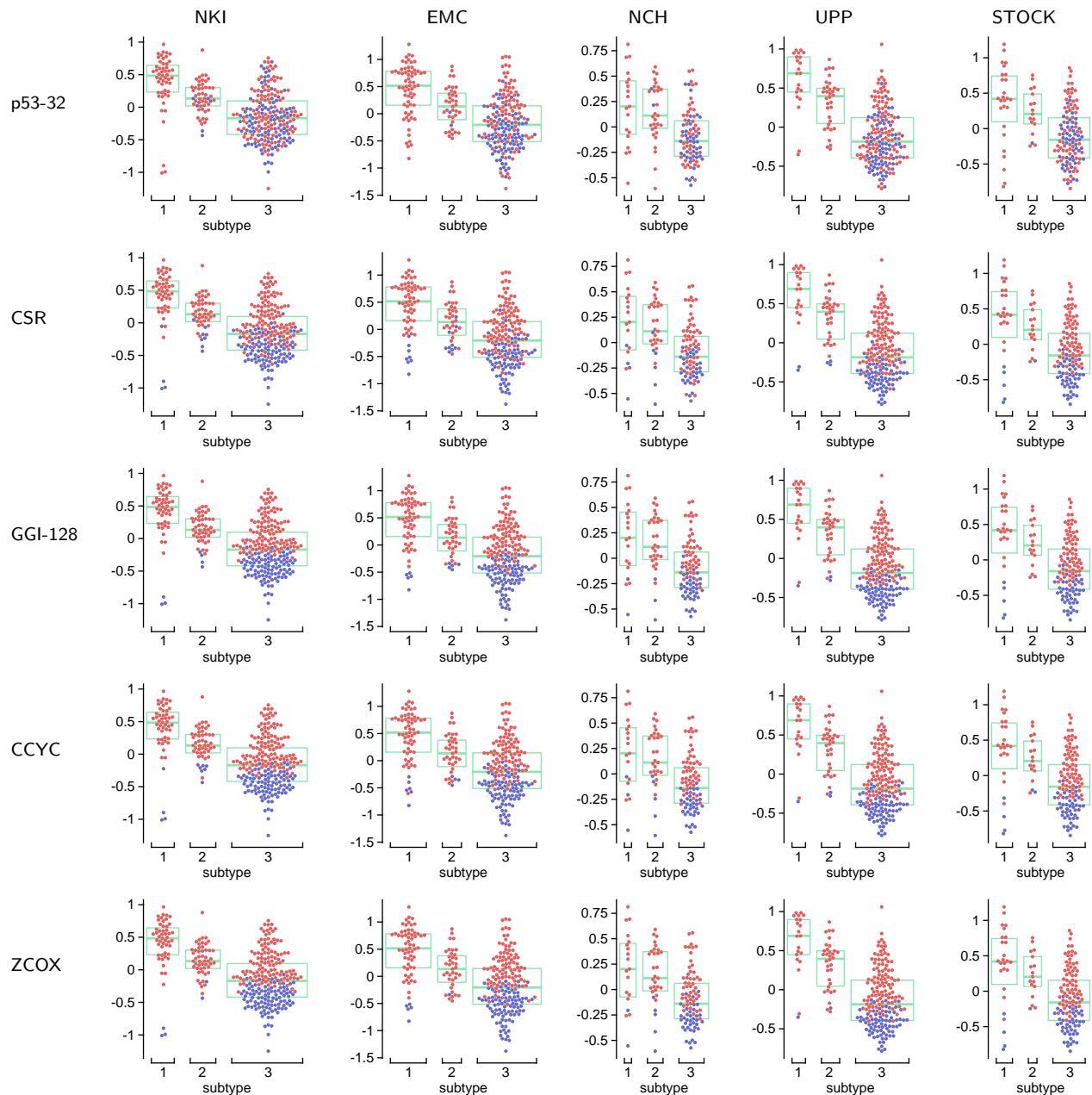


### 9.3 Patient classifications on proliferation-vs-subtype plots

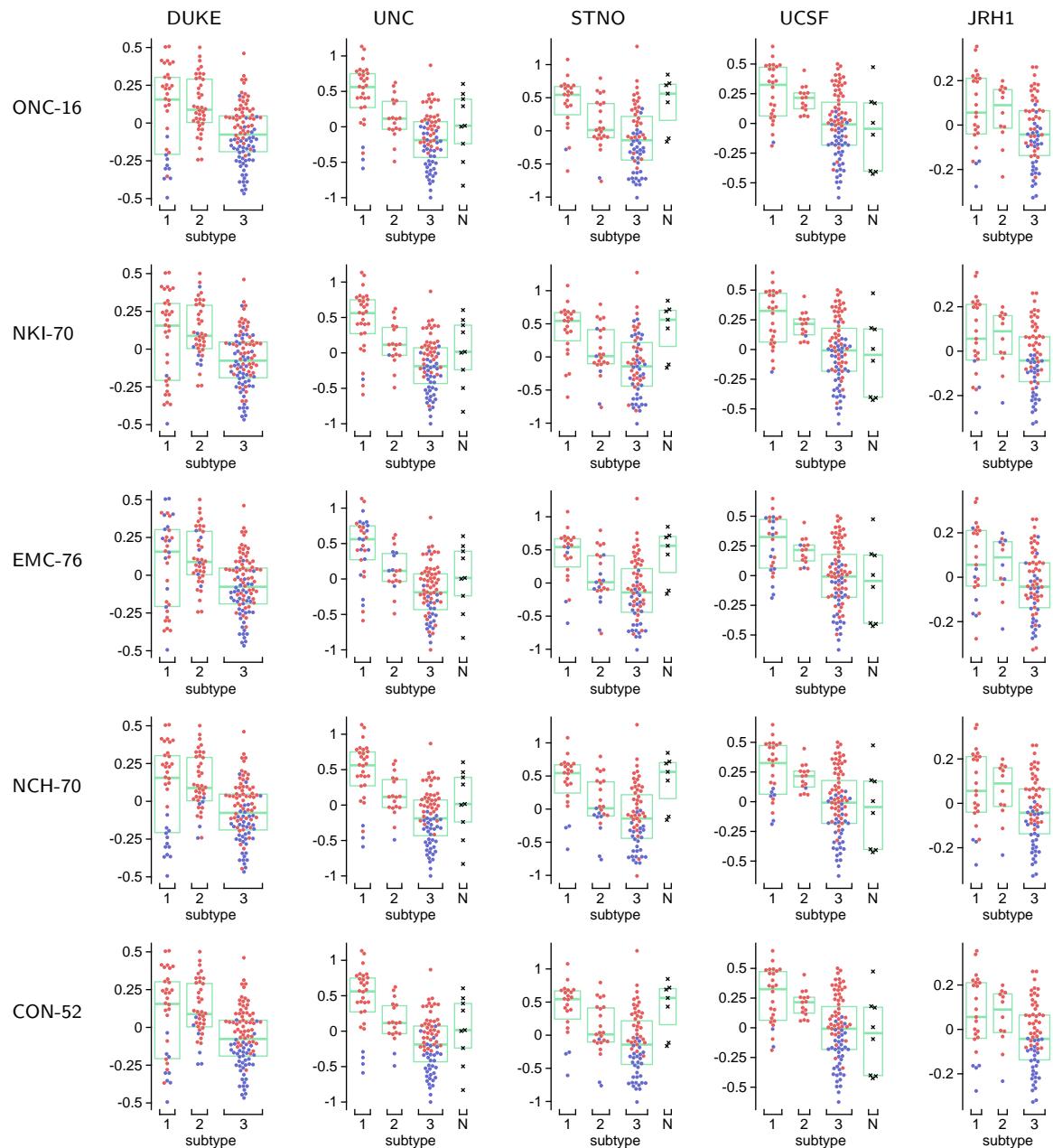
All vertical axes correspond to proliferation score.



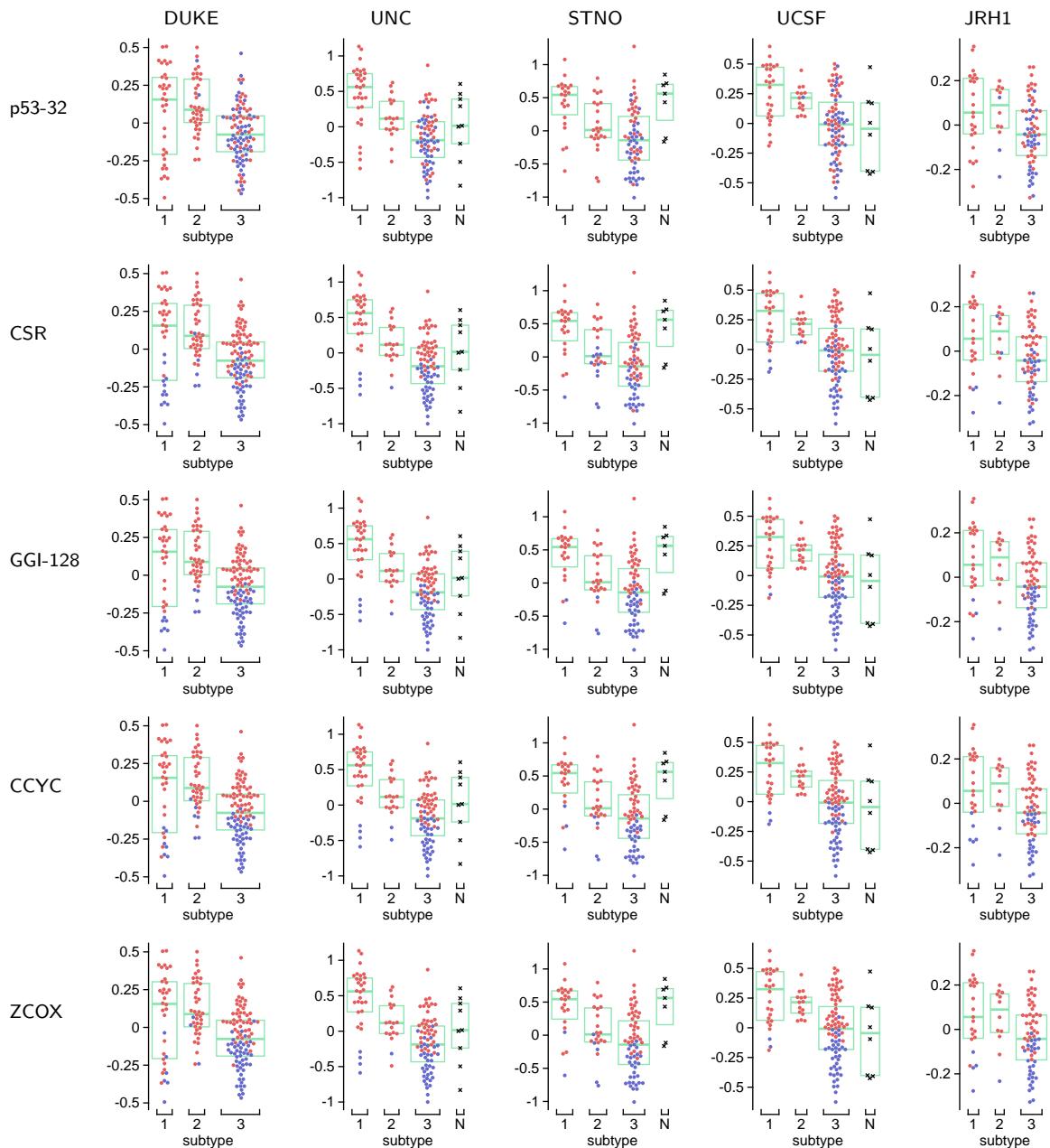
All vertical axes correspond to proliferation score.



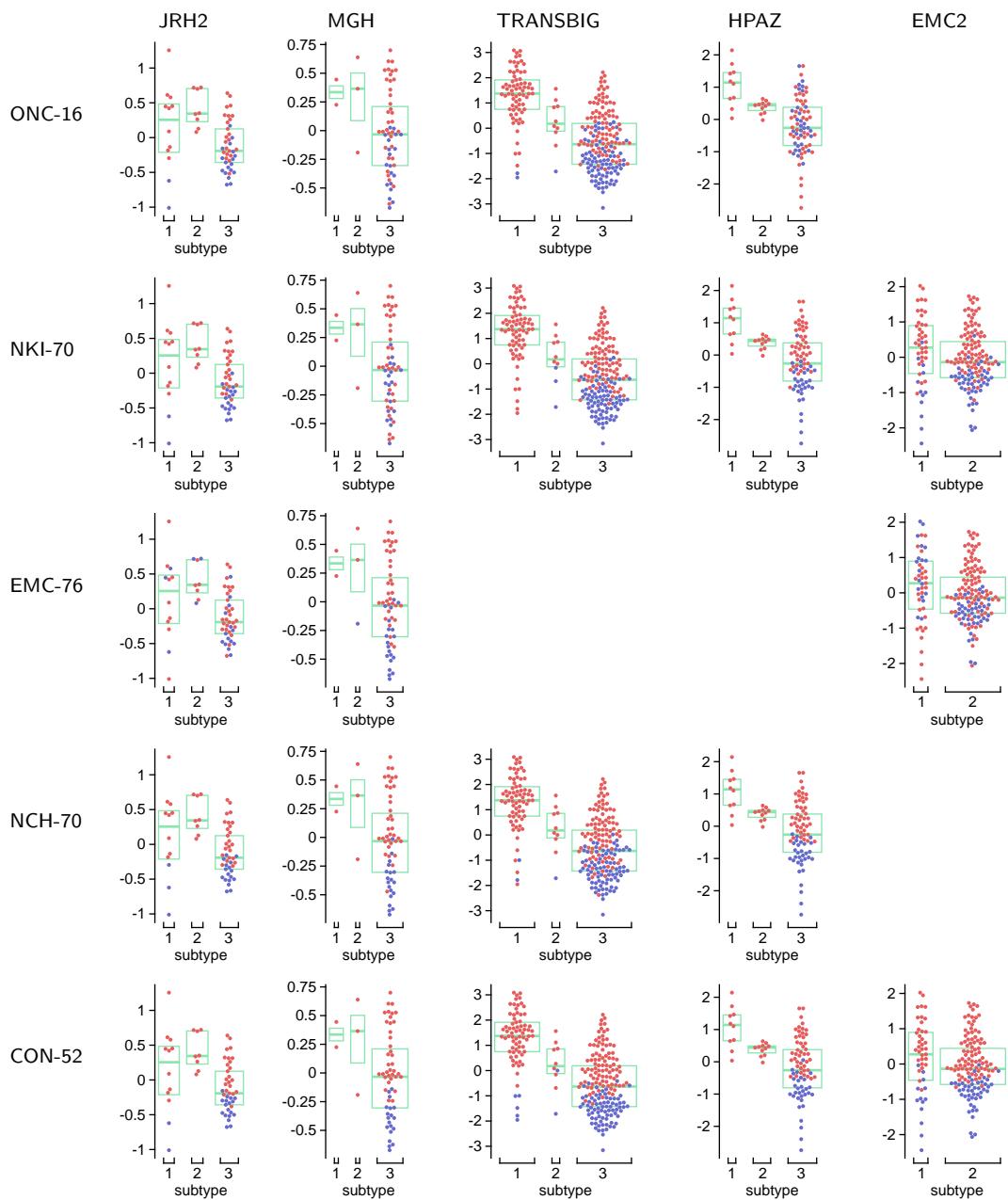
All vertical axes correspond to proliferation score.



All vertical axes correspond to proliferation score.



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