

# Liverome: a curated database of liver cancer-related gene signatures with self-contained context information

## ➤ Motivation of the study

- Our group has been performing microarray experiments on a large cohort of liver cancer patients (~300 patients).
- Needed to compare our own data with publicly available liver cancer data for prioritization of genes for further follow-up studies.
- The result from the public molecular profiling data most often comes in the form of **a list of genes**, also called **a gene signature**, reported in articles as **a table**
- These signatures are **scattered** in individual articles, **buried** in main or supplementary tables, thus all the valuable information is **underused**.
- To address this need, several **signature databases** have been constructed to serve as a repository of signatures. But several limitations were observed.

Table II. Top 30 Up-regulated Genes Distinguishing MD from WD

Predictor genes	<i>P</i> value
proteasome 26S subunit, ATPase, 5	2.774
cytochrome <i>c</i> oxidase subunit VIa polypeptide 1	1.992
<b>chaperonin containing TCP1, subunit 3</b>	<b>1.983</b>
prohibitin	1.803
<b>human D9 splice variant B mRNA</b>	<b>1.753</b>
proteasome subunit, $\beta$ , type 4	1.733
hydroxyacyl-coenzyme A dehydrogenase, type II	1.697
peptidylprolyl isomerase A	1.662
<b>adenosine deaminase, RNA-specific</b>	<b>1.654</b>
GCN5-like 1	1.591
mitochondrial ribosomal protein L12	1.493

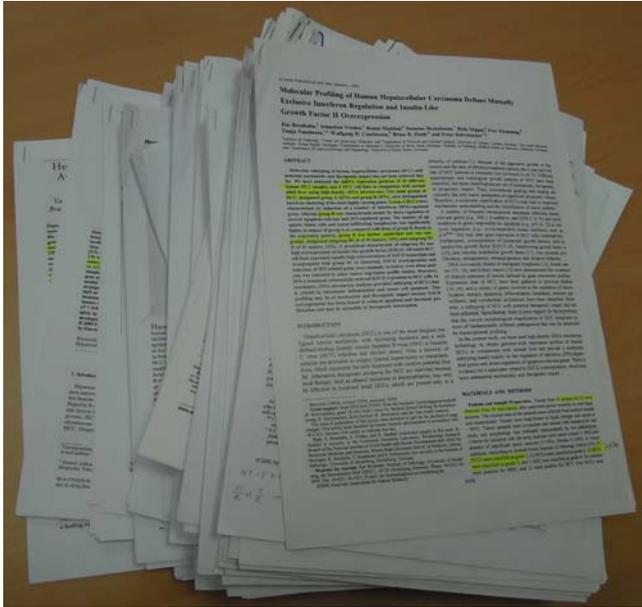
Seungwoo Hwang

Korean Bioinformation Center (KOBIC)

Korea Research Institute of Bioscience and Biotechnology (KRIBB)

# Database construction in a nutshell

~100 articles on liver cancer  
microarray and proteome studies



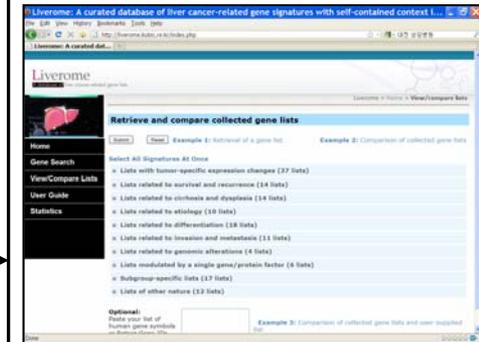
~150 gene signatures that  
appeared as tables

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Gene	100	1.449	0.963	167498
Galectin-3 binding protein (LGALS3BP)	100	1.425	0.974	163235
Paired basic amino acid cleaving enzyme (PACE)				

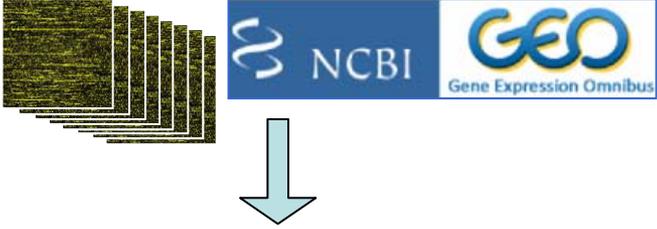
Liverome DB



- Thorough manual annotation of database content
- Comprehensive coverage
- Straightforward web interface designed for liver cancer biologists

# What is a signature database

## ➤ With respect to **data source**

	Publication-derived signatures	Raw data-derived signatures																																										
	<p><i>H. Nguyen et al. / Virology 334 (2005) 55–68</i></p> <table border="1"> <caption>Table 1 Immune response related genes were modulated by HCV Core expression</caption> <thead> <tr> <th>Gene name and probe set</th> <th>Fold change</th> <th>P value</th> </tr> </thead> <tbody> <tr> <td>LARC/MIP3A/CCL20 (liver activation regulated chemokine, 205476_AT)</td> <td>-3.15</td> <td>0.00108</td> </tr> <tr> <td>SSP1/Osteopontin (secreted phosphoprotein 1, 208975_S_AT)</td> <td>-3.04</td> <td>0.00374</td> </tr> <tr> <td>TNFSF10 (tumor necrosis factor (ligand) superfamily, member 10, 202688_AT)</td> <td>-1.89</td> <td>0.01131</td> </tr> <tr> <td>IL1RL1/IGS12L (interleukin 1 receptor-like 1 ligand, 203679_AT)</td> <td>-1.54</td> <td>0.03166</td> </tr> <tr> <td>FN1 (fibronectin 1, 212464_S_AT)</td> <td>-1.45</td> <td>0.01966</td> </tr> <tr> <td>MAP2K3 (mitogen-activated protein kinase kinase 3, 215499_AT)</td> <td>1.36</td> <td>0.03018</td> </tr> <tr> <td>B2M (Beta-2-microglobulin, 201891_S_AT)</td> <td>1.42</td> <td>0.01596</td> </tr> <tr> <td>CTSC (cathepsin C, 201487_AT)</td> <td>1.42</td> <td>0.00750</td> </tr> <tr> <td>MICB (MHC class I polypeptide-related sequence B, 206247_AT)</td> <td>1.56</td> <td>0.01557</td> </tr> <tr> <td>NK4 (NK cell transcript 4, 203828_S_AT)</td> <td>1.63</td> <td>0.01613</td> </tr> <tr> <td>PLA2G2A (phospholipase A2, group IIA, 203649_s_at)</td> <td>1.76</td> <td>0.00007</td> </tr> <tr> <td>Nuclear factor of kappa inhibitor ligand alpha (NFKBIA, 201502_s_at)</td> <td>1.90</td> <td>0.00040</td> </tr> <tr> <td>DEFB1 (defensin, beta 1, 210397_AT)</td> <td>3.04</td> <td>0.00449</td> </tr> </tbody> </table> <p>However, the expression of these genes was not significantly altered in NIH-3T3 cells stimulated by TNF-<math>\alpha</math> or prevented by the pro-apoptotic agent Bcl-2 (Jakobi et al., 2003). TNFSF10 is an apoptosis factor whose expression is induced in HCV Core protein T and human cervical cancer cells following serum starvation (Suzuki et al., 1997). The expression of Core protein in NIH-3T3 cells that can block Bax.</p> <p>HCV Core. Together with NS5A, the core protein forms a complex that affects the cell cycle in NIH-3T3 cells (De Vriese et al., 2003). DAXX was also observed to have an effect on the expression of DAXX expression in NIH-3T3 cells (Pluta et al., 1998; Yang et al., 2003; Michaelson et al., 2003). The data showed that the expression of DAXX was increased with the increased expression of DAXX. These results were shown to inhibit transactivation of NF-<math>\kappa</math>B.</p> <p>1 (SSP1/Osteopontin) and phospholipase A2 group 2A (PLA2G2A). TNFSF10 is a member of the TNF family of proteins that function as strong mediators of immune regulation and the inflammatory response. The presence of TNFSF10 was demonstrated in liver tissues of HCV infected patients and</p> <p><b>Signature tables in articles</b></p>	Gene name and probe set	Fold change	P value	LARC/MIP3A/CCL20 (liver activation regulated chemokine, 205476_AT)	-3.15	0.00108	SSP1/Osteopontin (secreted phosphoprotein 1, 208975_S_AT)	-3.04	0.00374	TNFSF10 (tumor necrosis factor (ligand) superfamily, member 10, 202688_AT)	-1.89	0.01131	IL1RL1/IGS12L (interleukin 1 receptor-like 1 ligand, 203679_AT)	-1.54	0.03166	FN1 (fibronectin 1, 212464_S_AT)	-1.45	0.01966	MAP2K3 (mitogen-activated protein kinase kinase 3, 215499_AT)	1.36	0.03018	B2M (Beta-2-microglobulin, 201891_S_AT)	1.42	0.01596	CTSC (cathepsin C, 201487_AT)	1.42	0.00750	MICB (MHC class I polypeptide-related sequence B, 206247_AT)	1.56	0.01557	NK4 (NK cell transcript 4, 203828_S_AT)	1.63	0.01613	PLA2G2A (phospholipase A2, group IIA, 203649_s_at)	1.76	0.00007	Nuclear factor of kappa inhibitor ligand alpha (NFKBIA, 201502_s_at)	1.90	0.00040	DEFB1 (defensin, beta 1, 210397_AT)	3.04	0.00449	 <p>Signatures are generated <b>by a re-analysis</b> of expression profile data from public repositories</p>
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Strength	<ul style="list-style-type: none"> <li>Utilize the end results from <b>expert analysis of individual studies</b></li> <li>Can <b>always</b> obtain signatures from articles</li> </ul>	<ul style="list-style-type: none"> <li>Consistent data processing scheme may generate signatures that are <b>more reproducible</b> across datasets</li> <li><b>Full list</b> of genes are generated</li> </ul>																																										

## ➤ With respect to **phenotype coverage**

	Specialized	All-inclusive
	for example, liver cancer	for example, all types of cancer all phenotypes
Strength	<ul style="list-style-type: none"> <li>Phenotype-specific coverage is generally high</li> </ul>	<ul style="list-style-type: none"> <li>Enables inter-phenotype comparison</li> </ul>

## Signature databases (a partial list)

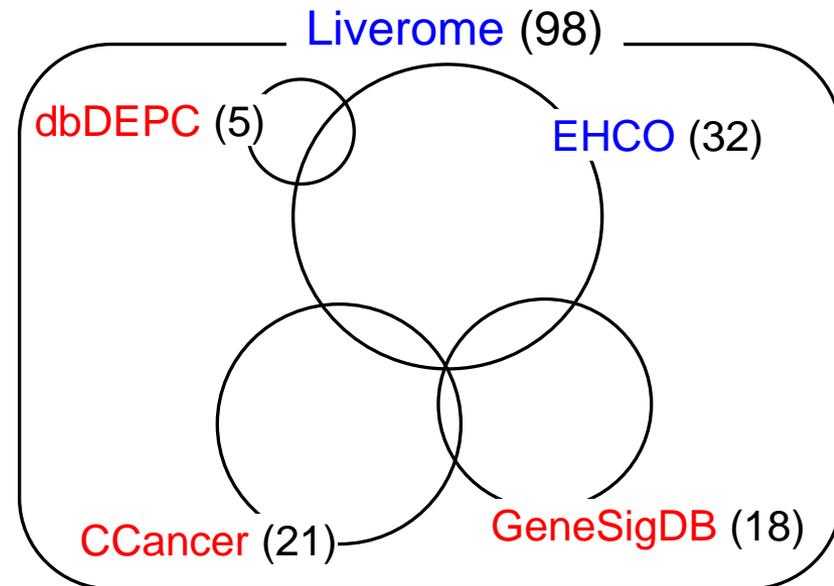
	Publication-derived	Raw data-derived
Specialized	<input type="checkbox"/> <b>Liverome</b> <input type="checkbox"/> EHCO ( <i>BMC Bioinfo</i> 2007)	<input type="checkbox"/> Pancreatic Expression DB ( <i>BMC Genomics</i> 2007; <i>NAR</i> 2011)
All-inclusive	<input type="checkbox"/> CCancer ( <i>NAR</i> 2010) <input type="checkbox"/> dbDEPC ( <i>NAR</i> 2010) <input type="checkbox"/> GeneSigDB ( <i>NAR</i> 2010) <input type="checkbox"/> MSigDB ( <i>Bioinformatics</i> 2011) <ul style="list-style-type: none"> <li>▪ Employed in GSEA</li> <li>▪ A subset “cgp” contains publication-derived signatures</li> </ul>	<input type="checkbox"/> Oncomine ( <i>Neoplasia</i> 2004, 2007) <input type="checkbox"/> GeneChaser ( <i>BMC Bioinfo</i> 2008)

# Liver cancer-specific data coverage

	Publication-derived
Specialized	<input type="checkbox"/> <b>Liverome</b> <input type="checkbox"/> EHCO ( <i>BMC Bioinfo</i> 2007)
All-inclusive	<input type="checkbox"/> CCancer ( <i>NAR</i> 2010) <input type="checkbox"/> dbDEPC ( <i>NAR</i> 2010) <input type="checkbox"/> GeneSigDB ( <i>NAR</i> 2010)

Red: All-inclusive DB  
 Blue: Liver cancer DB  
 (Number): Liver cancer-related articles

- Our data coverage is >3 times larger
- Even in large databases whose overall coverage is much higher than ours, their liver cancer-specific coverage was much lower than ours



	GeneSigDB	CCancer
Overall coverage	~10 times	~26 times
Liver cancer-specific coverage	1/5	1/5

➡ Points to the need for specialized database

# Database statistics of Liverome

## ➤ With respect to **clinicopathological category**

Category	# Signatures	# Articles
1) Tumor vs Normal comparison	37	32
2) Survival & Recurrence	14	14
3) Cirrhosis & Dysplasia	14	9
4) Etiology	10	9
5) Differentiation	18	13
6) Invasion & Metastasis	11	10
7) Genomic alterations	4	4
8) Modulated by a single gene/protein factor	6	6
9) Subgroup-specific	17	3
10) Other	12	11
<b>Total</b>	<b>143 signatures</b>	<b>98 articles</b>

## ➤ With respect to **type of experiment**

Type of experiment	# Signatures	# Articles
Transcriptomics	124	83
Proteomics	16	12
Others	3	3
<b>Total</b>	<b>143 signatures</b>	<b>98 articles</b>

# Well-annotated version of signatures: Main strength of Liverome

## Signature table as appeared in an article

Table 2 of Lau *et al* (2006) *Oncogene*

Gene name	log 2 ratios <sup>a</sup>	Compared sample groups?
Albumin (ALB)	3.8	• Tumor vs Normal?
Lactotransferrin (LTF)	3.5	• Between subtypes?
Slit homolog 3 (SLIT3)	2.3	• Something else?
. . . .		

Info is missing from table

## Uninformative form of signature from other DB (CCancer)

Table 2

Symbol
ALB
LTF
SLIT3
. . . .

Merely  
extracted  
gene IDs

↓ Read each article and derived an informative form

## Self-contained form of signature from Liverome

Lau (2006) *Oncogene* [Genes regulated by clusterin]

Symbol	Fold change (Clusterin-transfected cell line/ Control cell line)
ALB	13.900 Up
LTF	11.300 Up
SLIT3	4.900 Up
. . . .	

— Informatively named the signature

— Specifies the compared groups

— Represents fold change values  
in a scale that is more recognizable

# Well-annotated version of signatures: Main strength of Liverome (cont'd)

**Signature table as appeared in an article**

Table 2 of Okamoto *et al* (2006)

$P_1^a$	$P_2^a$	Up/down <sup>b</sup>	Symbol
.0016	.0118	Down	TRIM25
.0046	.0499	Up	EIF2S3
.0068	.0497	Down	DXYS155E
. . . .			

**Uninformative form of signature from other DB (GeneSigDB)**

Viral\_Okamoto06\_36genes

P1	P2	Up/down	Symbol
.0016	.0118	Down	TRIM25
.0046	.0499	Up	EIF2S3
.0068	.0497	Down	SFRS17A
. . . .			

  
 Extracted the signature table as-is

 Manual annotated all the information

**Self-contained form of signature from Liverome**

Okamoto (2006) Ann Sur Oncol  
 [Markers for multicentric hepatocarcinogenesis]

P-value (multicentric occurrence)	P-value (multicentric recurrence)	Change direction (Non-tumor/Normal)	Symbol
.0016	.0118	Down	TRIM25
.0046	.0499	Up	EIF2S3
.0068	.0497	Down	SFRS17A
. . . .			

Informatively named the signature

Specified the compared groups

# Summarized all the essential information underlying the signature

Iizuka (2002) Cancer Res [HBV-pos...]	
<b>Nature of list</b>	Genes that are differentially expressed between HBV-positive HCC and HCV-positive HCC
<b>Platform</b>	Affymetrix HuGeneFL Array
<b>Number of genes</b>	80 genes
<b>Samples</b>	Tumor samples from 14 HBV-positive HCC patients and from 31 HCV-positive HCC patients and 6 normal liver samples
<b>Samples Characteristics</b>	Etiology: <ul style="list-style-type: none"><li>■ HBV: 14 patients (31%)</li><li>■ HCV: 31 patients (69%)</li></ul>
<b>Data analysis method</b>	Random permutation test using Fisher ratio as a statistic ( $p < 0.05$ ) and fold change filtering ( $FC > 2$ -fold)
<b>Reference</b>	Iizuka et al (2002) Comparison of gene expression profiles between hepatitis B virus- and hepatitis C virus-infected hepatocellular carcinoma by oligonucleotide microarray data on the basis of a supervised learning method. <i>Cancer Res.</i> <a href="#">PubMed.gov</a>
<b>Source</b>	Table 2

## Main point:

- Made extensive manual annotation efforts to contain **all context information within the database**
- Should enable **easier database browsing** without a need to refer to the original publication

# Straightforward web interface: Gene search

➤ A gene search result for “CES2 (carboxylesterase 2)”

Description of gene list	Evidence	
Chaerkady (2008) <i>J Proteome Res</i> Tumor vs Non-tumor	Fold change (Tumor/Non-tumor)	1.667 Down
Lee (2004) <i>Hepatology</i> Genes associated with survival	Hazard Ratio	0.663
	P-value (Wald test)	2.200E-4
Kato (2005) <i>Nucleic Acids Res</i> HBV-tumor vs HCV-tumor	P-value	0.025
Kato (2005) <i>Nucleic Acids Res</i> Tumor vs Non-tumor	P-value	0.021
Chiang (2008) <i>Cancer Res</i> Genes specific to proliferation subgroup	SAM score	-14.350
	Fold change (Proliferation subgroup/Other subgroups)	5.260 Down
	q-value	0
Hsu (2007) <i>BMC Bioinformatics</i> HCC-related genes from PubMed text mining	Related to	HCC
Iizuka (2002) <i>Cancer Res</i> HBV-tumor vs HCV-tumor	Fold change (HBV-tumor/Normal)	1.890 Down
	Fold change (HCV-tumor/Normal)	1.158 Up
	Fold change (HBV-tumor/HCV-tumor)	2.189 Down
Kurokawa (2003) <i>J Hepatol</i> Non-tumor vs Normal liver	P-value	0.007
	Change direction (Non-tumor/Normal)	Down

Survival

DE in T vs NT

Viral infection status

Subtype-specific

# Straightforward web interface: Signature comparison

⊕ Lists with tumor-specific expression changes (37 lists)

⊕ Lists related to survival and recurrence (14 lists)

⊕ Lists related to cirrhosis and dysplasia (14 lists)

⊕ Lists related to etiology (10 lists)

⊕ Lists related to differentiation (18 lists)

⊕ Lists related to invasion and metastasis (11 lists)

☐ Lists related to genomic alterations (4 lists)

<input checked="" type="checkbox"/>	Skawran	(2008) Mod Pathol	Genes up-regulated in HCC compared to HCA, and located in amplified chromosomal regions	17 genes
<input type="checkbox"/>	Skawran	(2008) Mod Pathol	HCC with 13q loss vs HCC without 13q loss	22 genes
<input checked="" type="checkbox"/>	Tsai	(2006) J Biomed Sci	Genes under-expressed in tumor and located within frequently deleted loci	17 genes
<input checked="" type="checkbox"/>	Woo	(2009) Cancer Res	Potential driver genes of HCC	50 genes

⊕ Lists modulated by a single gene/protein factor (6 lists)

⊕ Subgroup-specific lists (17 lists)

⊕ Lists of other nature (12 lists)

## Optional:

Paste your list of human gene symbols or Entrez Gene IDs (max: 1000 genes)



**Example 3:** Comparison of collected gene lists and user-supplied list

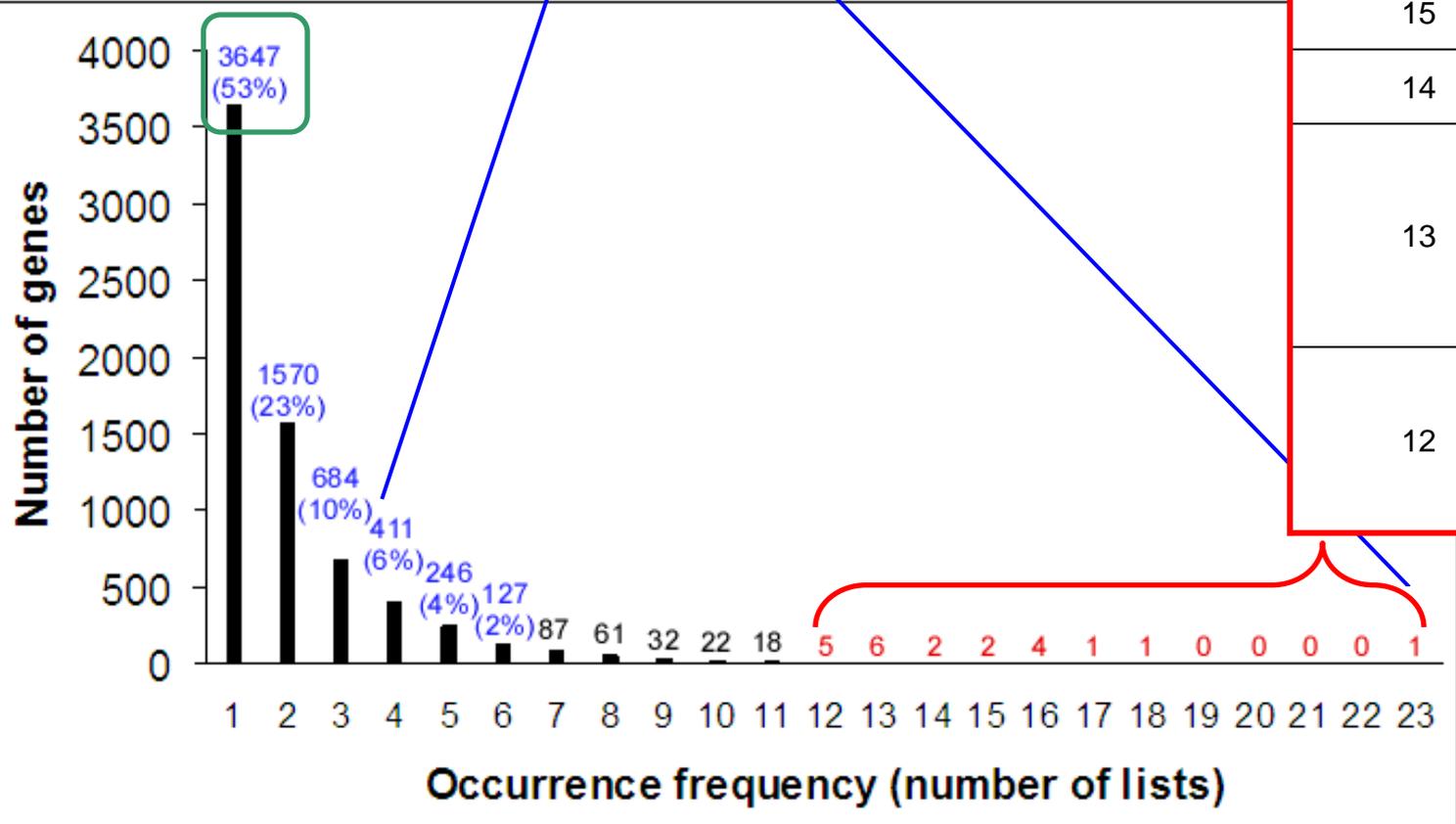
# Prioritization of genes according to occurrence frequency

~20 genes occur **very frequently** in  $\geq 12$  signatures

~1,000 genes occur **frequently** in  $\geq 4$  signatures

~ A half of the genes occur in only one signature

Occurrence	Symbol
23	ECHS1
18	ADH1B
17	GPC3
16	ALB
	BHMT
	PLG
	VIM
15	RGN
	TF
14	FABP1
	HPD
13	ACADSB
	CAT
	MTHFD1
	RPSA
	SLC22A1
	TDO2
12	ADH4
	CP
	CYP2E1
	PCK1
	SPARC



# Construction of co-occurrence network of genes

## A network analysis using Liverome-collected signatures

### Method

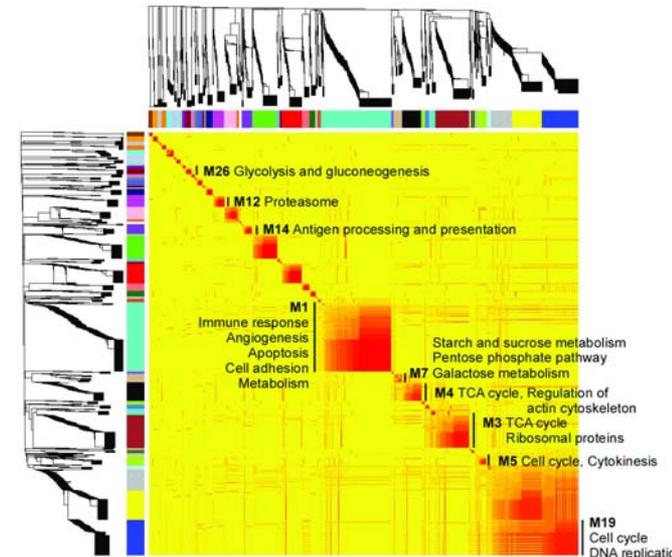
- Used **WGCNA** (**Weighted Gene Co-expression Network Analysis**) Langfelder & Horvath, 2007, BMC Bioinformatics
- Usually used to construct **co-expression network** from expression profile data
- Here, used it to construct **co-occurrence network** from signature data
- A pair of genes are considered as similar if they **co-occur in many of the signatures**

	Signature1	Signature2	Signature3	Signature4	Signature5	Signature6	...	Signature N
Gene A								
Gene B								

$$\text{Co-occurrence (Jaccard similarity coefficient)} = \frac{\# \text{ Signatures containing both gene A and gene B}}{\# \text{ All other signatures}}$$

### Result: a co-occurrence network

- Genes are shown on rows and columns
- Color coding represents the similarity measure
- Each block represents a module which consists of a set of genes that have **similar liver cancer signature membership**
- Cancer-related pathways are enriched in the modules (glycolysis, cell cycle, apoptosis, etc.)
- Co-occurrence network constructed from liver cancer signature data alone recapitulates known liver cancer biology



# Summary

- **Comprehensive** collection of liver cancer-related gene signatures
- All the database content was made into a **self-contained form** by extensive manual annotation
- **Limitations:** All limitations inherent to publication-based signature database
  - Each signature contains **only a few selected genes** above the significance cutoff
  - Each signature was derived from its own data processing scheme, which **may decrease reproducibility** across datasets
- **Usefulness**
  - Most useful **to retrieve known differential expression information of a gene**
  - To **compare your own gene list** with previously reported gene lists
  - An interesting **bioinformatics analysis** may be possible using the DB contents

## Main contributors

Korea Research Institute of Bioscience & Biotechnology	<a href="#">Hyang-Sook Yoo</a>	Project conception & supervision
	Langho Lee	Database and web programming
Pfizer	Kai Wang	Co-occurrence network analysis
	Gang Li	Biological discussion
Personal Genome Institute	Jong Bhak	Project initiation

Also thank to the authors of the articles that we collected and included in the database