## Gene Expression Analysis – Gene Set Analysis Approach CANB 7640

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http://tanlab.ucdenver.edu/labHomePage/teaching/CANB7640/

## Outline

- Methods for Identifying Gene Set Analysis
  - Motivation
  - Statistics
  - Gene Set Enrichment Analysis (GSEA)

## Limitations of Candidate Gene Analysis

- Functional genomics technologies such as expression profiling using microarrays provide a global approach to understanding cellular. processes in different biological phenotypes.
- Candidate genes analyses
  - Gene lists
  - Number of genes range from hundred to thousands
  - Sifting through gene list is a daunting task to group these genes into functional groups (*ad hoc* analysis)
  - Bias and require expert knowledge

## Motivation

- Genes must act in concert to drive various cellular processes.
- Gene expression alterations might be revealed at the level of biological pathways or co-regulated gene sets (functional groups).
- Gene set analysis
  - more objective and robust.
  - able to discover sets of coordinated differentially expressed genes among pathway members and their association to a specific biological phenotype.
  - provide new insights linking biological phenotypes to their underlying molecular mechanisms.
  - suggesting new hypotheses about pathway membership and connectivity.

## How to find Sets of Co-ordinately Differentially Expressed Genes

- High-throughput "omics" data (e.g. microarray gene expression, RNA-seq etc) full with noise.
- Real biological signals might be subtle.
- (Assumption) Genes with similar function or participate in a biological process have similar expression patterns.
- Goal: Find these sets of genes from highthroughput "omics" data

## Gene Set Analysis

- Samples with known biological phenotypes (e.g. *class labels*)
- High-throughput measurements of data points (e.g. gene expressions)
- Set of genes involved in biological processes or cellular functions or pathways (e.g. gene sets)
- Compare the <u>gene expressions</u> of various <u>class labels</u> to find differentially expressed <u>gene sets</u>.

## Gene Set Enrichment Analysis (GSEA)

- *Goal*: to detect modest but coordinated expression changes in pre-specified sets of related genes (gene sets).
- Gene set can be all the genes involved in
  - specific pathway (obtained from Pathway databases such as KEGG, BioCARTA, REACTOME etc)
  - specific gene ontology class (obtained from Gene Ontology category)
  - specific chromosome locations
  - specific transcriptional regulated targets (e.g. transcription factor targets, miRNA targets)
  - specific gene signatures (obtained from published papers or your own experiments)
  - Specific drug targets (obtained from experiments of druggene interactions)

## **Original Publication of GSEA**

# PGC-1α-responsive genes involved in oxidative phosphorylation are coordinately downregulated in human diabetes

Vamsi K Mootha<sup>1,2,3,10</sup>, Cecilia M Lindgren<sup>1,4,10</sup>, Karl-Fredrik Eriksson<sup>4</sup>, Aravind Subramanian<sup>1</sup>, Smita Sihag<sup>1</sup>, Joseph Lehar<sup>1</sup>, Pere Puigserver<sup>5</sup>, Emma Carlsson<sup>4</sup>, Martin Ridderstråle<sup>4</sup>, Esa Laurila<sup>4</sup>, Nicholas Houstis<sup>1</sup>, Mark J Daly<sup>1</sup>, Nick Patterson<sup>1</sup>, Jill P Mesirov<sup>1</sup>, Todd R Golub<sup>1,5</sup>, Pablo Tamayo<sup>1</sup>, Bruce Spiegelman<sup>5</sup>, Eric S Lander<sup>1,6</sup>, Joel N Hirschhorn<sup>1,7,8</sup>, David Altshuler<sup>1,2,7,9,11</sup> & Leif C Groop<sup>4,11</sup>

DNA microarrays can be used to identify gene expression changes characteristic of human disease. This is challenging, however, when relevant differences are subtle at the level of individual genes. We introduce an analytical strategy, Gene Set Enrichment Analysis, designed to detect modest but coordinate changes in the expression of groups of functionally related genes. Using this approach, we identify a set of genes involved in oxidative phosphorylation whose expression is coordinately decreased in human diabetic muscle. Expression of these genes is high at sites of insulin-mediated glucose disposal, activated by PGC-1 $\alpha$  and correlated with total-body aerobic capacity. Our results associate this gene set with clinically important variation in human metabolism and illustrate the value of pathway relationships in the analysis of genomic profiling experiments.

### NATURE GENETICS VOLUME 34 | NUMBER 3 | JULY 2003 267-273 [Citations: >4700]

## **Original Publication of GSEA**

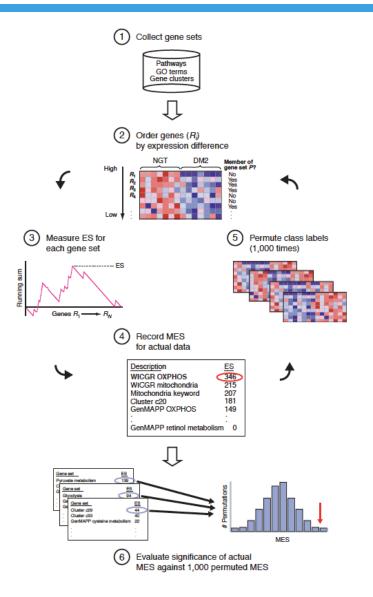


Figure 1 Schematic overview of GSEA. The goal of GSEA is to determine whether any a priori defined gene sets (step 1) are enriched at the top of a list of genes ordered on the basis of expression difference between two classes (for example, highly expressed in individuals with NGT versus those with DM2). Genes  $R_1, \dots R_N$  are ordered on the basis of expression difference (step 2) using an appropriate difference measure (for example, SNR). To determine whether the members of a gene set S are enriched at the top of this list (step 3), a Kolmogorov-Smirnov (K-S) running sum statistic is computed: beginning with the top-ranking gene, the running sum increases when a gene annotated to be a member of gene set S is encountered and decreases otherwise. The ES for a single gene set is defined as the greatest positive deviation of the running sum across all N genes. When many members of S appear at the top of the list, ES is high. The ES is computed for every gene set using actual data, and the MES achieved is recorded (step 4). To determine whether one or more of the gene sets are enriched in one diagnostic class relative to the other (step 5), the entire procedure (steps 2-4) is repeated 1,000 times, using permuted diagnostic assignments and building a histogram of the maximum ES achieved by any pathway in a given permutation. The MES achieved using the actual data is then compared to this histogram (step 6, red arrow). providing us with a global P value for assessing whether any gene set is associated with the diagnostic categorization.

## **Original Publication of GSEA**

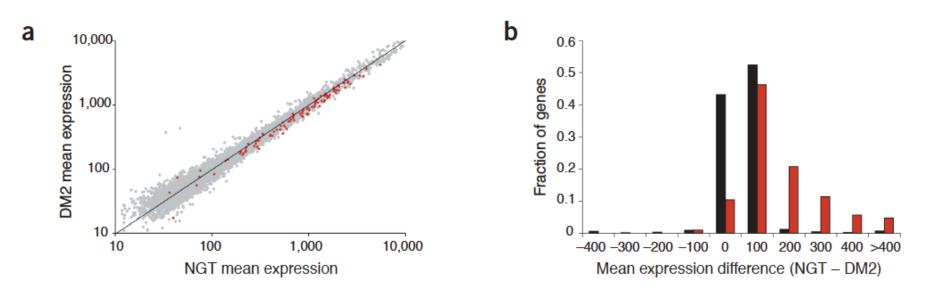


Figure 2 OXPHOS gene expression is reduced in diabetic muscle. (a) The mean expression of all genes (gray) and of OXPHOS genes (red) is plotted for individuals with DM2 versus those with NGT. (b) Histogram of mean gene expression level differences between individuals with NGT and DM2, using the data from a, for all genes (black) and for OXPHOS genes (red).

## GSEA paper (PNAS 2005)

### Gene set enrichment analysis: A knowledge-based approach for interpreting genome-wide expression profiles

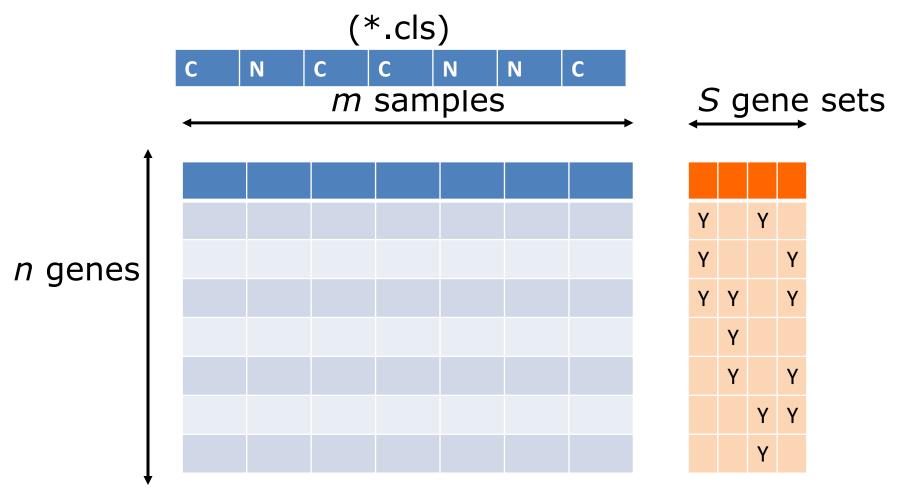
Aravind Subramanian<sup>a,b</sup>, Pablo Tamayo<sup>a,b</sup>, Vamsi K. Mootha<sup>a,c</sup>, Sayan Mukherjee<sup>d</sup>, Benjamin L. Ebert<sup>a,e</sup>, Michael A. Gillette<sup>a,f</sup>, Amanda Paulovich<sup>g</sup>, Scott L. Pomeroy<sup>h</sup>, Todd R. Golub<sup>a,e</sup>, Eric S. Lander<sup>a,c,i,j,k</sup>, and Jill P. Mesirov<sup>a,k</sup>

PNAS October 25, 2005 vol. 102 no. 43 15545–15550 [Citations: >14000]

## GSEA Algorithm: Step 1

- Collect gene sets (\*.gmt) from databases
- Compile gene expression data (\*.gct)
- Define class labels for your samples (\*.cls)
- Microarray chip definition file (\*.chip) [Not applicable to RNA-seq]

## Required Files for GSEA



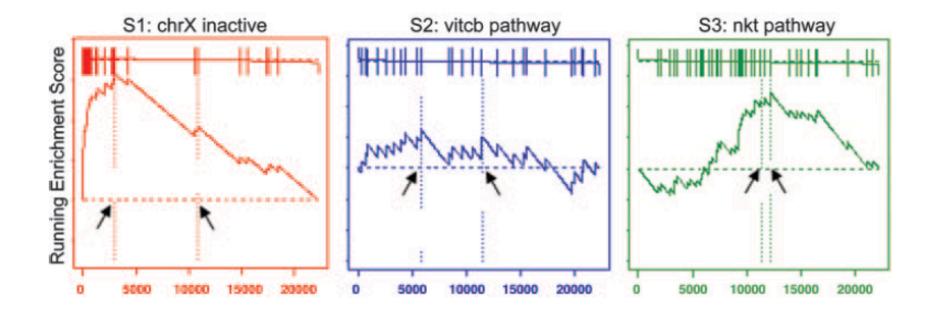
(\*.gct)

(\*.gmt)

## GSEA Algorithm: Step 2

- Rank genes based on their expression differences between the two phenotypes (in GSEA, the measurement is Signal-to-noise ratio)
- Compute Enrichment Score (ES)
  - Compute cumulative sum over ranked genes:
    - Increase sum when gene in set, decrease it otherwise.
    - Magnitude of increment depends on correlation of gene with phenotype.
  - Maximum deviation from zero = enrichment score

## GSEA Algorithm: Step 2 Enrichment Plot



## Results on SETPATHWAY (Enrichment Plot)

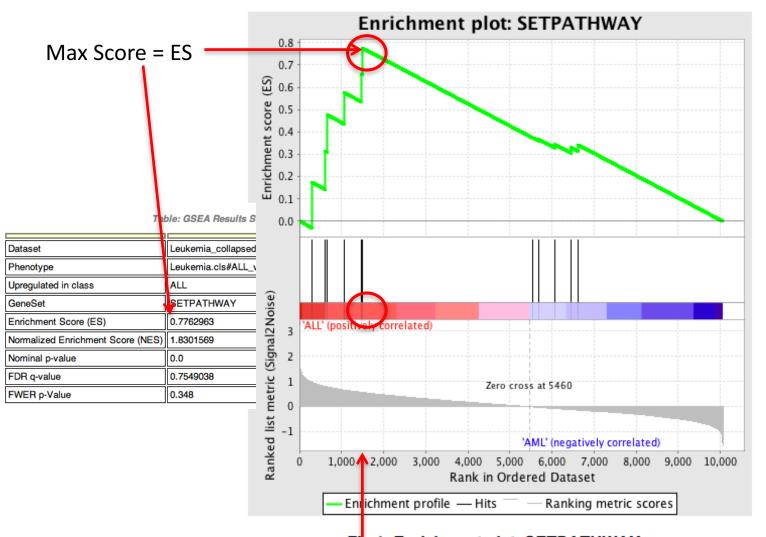


Fig 1: Enrichment plot: SETPATHWAY Profile of the Running ES Score & Positions of GeneSet Members on the Rank Ordered List

## GSEA Algorithm: Step 3

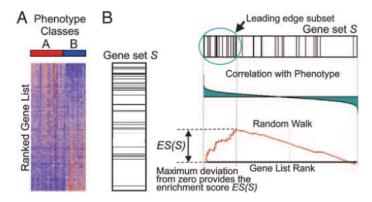
- Compute the Significance of the Gene Sets by Permutation Test
- Permutation Test (*n* times)
  - Permute phenotype labels
  - Permute gene sets
- For each permutation, compute ES score
- Compare ES score for actual data to distribution of ES scores from permuted data

## GSEA Algorithm: Step 4

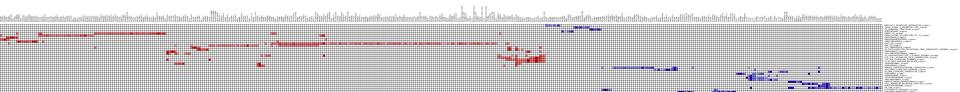
- Adjustment for multiple hypothesis testing:
  - Normalize the ES accounting for size of each gene set, yielding normalized enrichment score (NES)
  - Control proportion of false positives by calculating FDR corresponding to each NES, by comparing tails of the observed and null distributions for the NES.

## GSEA Leading Edge Analysis

- Genes might be involved in different pathways/gene sets.
- Selected core genes might be identified in top enriched gene sets.



**Fig. 1.** A GSEA overview illustrating the method. (*A*) An expression data set sorted by correlation with phenotype, the corresponding heat map, and the "gene tags," i.e., location of genes from a set *S* within the sorted list. (*B*) Plot of the running sum for *S* in the data set, including the location of the maximum enrichment score (*ES*) and the leading-edge subset.



**GSEA** Homepage

### http://www.broadinstitute.org/gsea/index.jsp

e Set Enrichment Analysis GSEA Home	Downloads	Molecular	Signatures Database	Documentation	Contact
Overview		ſ	Molecular Profile Data		
<ul> <li>Gene Set Enrichment Analysis (GSEA) is a completermines whether an a priori defined set of generis significant, concordant differences between two bio (e.g. phenotypes).</li> <li>From this web site, you can:</li> <li>Download the GSEA software and additional reannotate and interpret enrichment results.</li> <li>Explore the Molecular Signatures Database annotated gene sets for use with GSEA software</li> <li>View documentation describing GSEA and MS</li> </ul>	s shows statistically logical states esources to analyze, e (MSigDB), a collectic e.		Gene Set Database	Run GSEA	
What's New					

19-Sep-2016: The Documentation section of our website is temporarily offline. We are working to resolve the issue and get it back as soon as possible.

15-Aug-2016: The first beta of the next major GSEA Desktop release is available, with SVG plots, Cytoscape 3.3+ support and much more.

17-Jun-2016: You can now follow @GSEA\_MSigDB on Twitter!

29-Feb-2016: The Sunday 28-Feb-2016 maintenance is complete on the GSEA/MSigDB website. Thanks for your patience!

13-Jan-2016: Version 5.1 of the Molecular Signatures Database (MSigDB) is now available. It includes the addition of 2,962 gene sets to the C7 collection of immunologic signatures, as well as a number of updates and corrections. See the Release Notes for details.

23-Dec-2015: Our paper describing the generation of the Hallmarks collection and examples of its use for GSEA was published in Cell Systems.

10-Dec-2015: We have confirmed that GSEA v2.2.0 and newer are compatible with Java 8 and produce equivalent results. Its use is highly recommended.

### Registration

Please register to download the GSEA software and view the MSigDB gene sets. After registering, you can log in at any time using your email address. Registration is free. Its only purpose is to help us track usage for reports to our funding agencies.

#### Contributors

GSEA and MSigDB are maintained by the GSEA team with the support of our MSigDB Scientific Advisory Board. Our thanks to our many contributors. Funded by: National Cancer Institute, National Institutes of Health, National Institute of General Medical Sciences.



### Citing GSEA

To cite your use of the GSEA software, please reference Subramanian, Tamayo, et al. (2005, PNAS 102, 15545-15550) and Mootha, Lindgren, et al. (2003, Nat Genet 34, 267-273).

## **GSEA** Implementation

### http://www.broadinstitute.org/gsea/index.jsp

### Software

There are several options for GSEA software. All options implement exactly the same algorithm. Usage recommendations and installation instructions are listed below. Current Java implementations of GSEA require Java 7 or 8. Java 8 is recommended.

javaGSEA	<ul> <li>Easy-to-use graphical user interface</li> </ul>	Launch with
Desktop Application	<ul> <li>Runs on any desktop computer (Windows, Mac OS X, Linux etc.) that supports</li> </ul>	1GB (for 32 or 64-bit Java)
	Java 7 or 8. Java 8 is recommended.	memory:
	<ul> <li>Produces richly annotated reports of enrichment results</li> </ul>	🔮 Launch
	<ul> <li>Integrated gene sets browser to view gene set annotations, search for gene sets and map gene sets between platforms</li> </ul>	
avaGSEA	Command line or offline usage. See our User Guide for details.	download
Java Jar file	Runs on any platform that supports Java 7 or 8. Java 8 is recommended.	gsea2-2.2.2.jar
	$\blacktriangleright$ We recommend using the 'Launch' buttons above instead of this mode for most users	
BETA	The first Beta version of the next major GSEA Desktop release, with SVG plots, Cytoscape 3.3+ support for the Enrichment Map, and much more.	BETA
javaGSEA Java Jar file	Our tests show this Beta version produces equivalent results, but use the Production version if you have concerns. At a minimum, verification with the Production version before publication is strongly recommended.	download gsea2-3.0_beta_1.jar
	<ul> <li>Please contact us with bugs or other feedback. We will aim to address problems as soon as possible in future Beta releases.</li> </ul>	
	Runs only on the command line. See our User Guide for details.	
	Runs on any platform that supports Java 7 or 8. Java 8 is recommended.	
R-GSEA	Usage from within the R programming environment	download
R Script	<ul> <li>Easily inspect, learn and tweak the algorithm</li> </ul>	GSEA-P-R.1.0.zip
	<ul> <li>Incorporate GSEA into your own data analysis pipeline</li> </ul>	
	<ul> <li>Programmatically call the open source GSEA R API</li> </ul>	
	Note that this script has not been updated since 2005 and may not work as-is with modern R distributions.	
	<ul> <li>Click here to learn more about the R-GSEA script</li> </ul>	
GenePattern GSEA Module	Use GSEA from within GenePattern	GenePattern site
	<ul> <li>Use GSEA in concert with a large suite of other analytics found in GenePattern (a powerful and flexible analysis platform developed at the Broad Institute)</li> </ul>	

For details on the GSEA algorithm and software refer to the Documentation. For details on the latest release refer to the Release Notes.

-	<b>MSigDB</b>
	Molecular Signat Database

Molecular Signatures Database v5.1

#### Overview

The Molecular Signatures Database (MSigDB) is a collection of annotated gene sets for use with GSEA software. From this web site, you can

Signatures

- Search for gene sets by keyword.
- Browse gene sets by name or collection.
- Examine a gene set and its annotations. See, for example, the ANGIOGENESIS gene set page.
- Download gene sets.
- Investigate gene sets:
  - Compute overlaps between your gene set and gene sets in MSigDB.
  - Categorize members of a gene set by gene families.
  - View the expression profile of a gene set in any of the three provided public expression compendia.

#### Registration

Please register to download the GSEA software and view the MSigDB gene sets. After registering, you can log in at any time using your email address. Registration is free. Its only purpose is to help us track usage for reports to our funding agencies.

#### **Current Version**

MSigDB database v5.1 updated January 2016. Release notes. GSEA/MSigDB web site v5.0 released March 2015

#### Contributors

The MSigDB is maintained by the GSEA team with the support of our MSigDB Scientific Advisory Board. We also welcome and appreciate contributions to this shared resource and encourage users to submit their gene sets to genesets@broadinstitute.org. Our thanks to our many contributors.

Funded by: National Cancer Institute, National Institutes of Health, National Institute of General Medical Sciences.



### Collections

The MSigDB gene sets are divided into 8 major collections:

hallmark gene sets are coherently expressed signatures derived by aggregating many MSigDB п gene sets to represent well-defined biological states or processes.

positional gene sets for each human chromosome and cytogenetic band.

curated gene sets from online pathway databases, publications in PubMed, and knowledge of domain experts.

motif gene sets based on conserved cis-regulatory motifs from a comparative analysis of the human, mouse, rat, and dog genomes.

computational gene sets defined by mining large collections of cancer-oriented microarray data.

GO gene sets consist of genes annotated by the same GO terms.

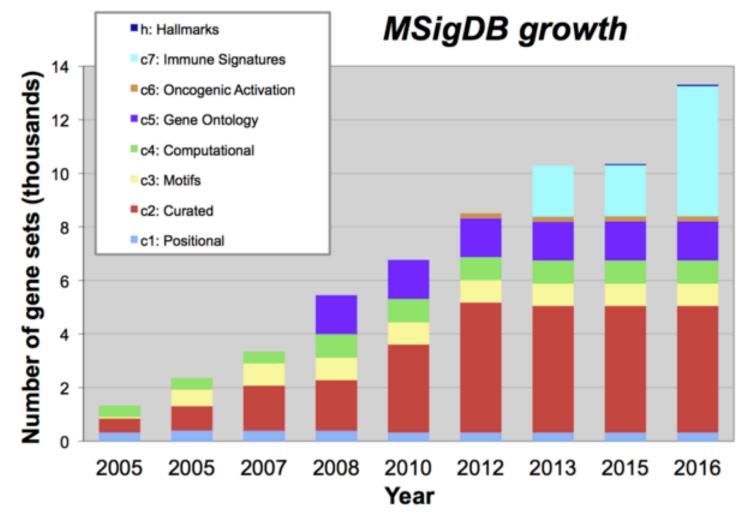
oncogenic signatures defined directly from Cb microarray gene expression data from cancer gene perturbations.

immunologic signatures defined directly from microarray gene expression data from immunologic studies.

#### Citing the MSigDB

To cite your use of the Molecular Signatures Database (MSigDB), please reference Subramanian, Tamayo, et al. (2005, PNAS 102, 15545-15550) and also the source for the gene set as listed on the gene set page.

Currently contains 13,311 gene sets organized into 8 categories.



### **Browse Gene Sets**

Gene set name:	Search	
By first letter:	1 2 3 4 5 6 7 8 9 0 A B C D E F G H I J K L M N O P Q R S T U V W X Y Z	
By collection:	[about the MSigDB collections]	
	H (hallmark gene sets, 50 gene sets)	
	C1 (positional gene sets, 326 gene sets)	
	by chromosome: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 X Y	
	C2 (curated gene sets, 4726 gene sets) 2	
	CGP (chemical and genetic perturbations, 3396 gene sets) 12	
	CP (Canonical pathways, 1330 gene sets)	
	CP:BIOCARTA (BioCarta gene sets, 217 gene sets) 2	
	CP:KEGG (KEGG gene sets, 186 gene sets)	
	CP:REACTOME (Reactome gene sets, 674 gene sets)	
	C3 (motif gene sets, 836 gene sets) 2	
	MIR (microRNA targets, 221 gene sets) 2	
	TFT (transcription factor targets, 615 gene sets) 2	
	C4 (computational gene sets, 858 gene sets) 2	
	CGN (cancer gene neighborhoods, 427 gene sets) 2	
	CM (cancer modules, 431 gene sets)	
	C5 (GO gene sets, 1454 gene sets) 2	
	BP (GO biological process, 825 gene sets) 2	
	CC (GO cellular component, 233 gene sets)	
	MF (GO molecular function, 396 gene sets) 2	
	C6 (oncogenic signatures, 189 gene sets)	
	C7 (immunologic signatures, 4872 gene sets) 12	

Click on a gene set name to view its gene set page.

HALLMARK\_ADIPOGENESIS HALLMARK\_ALLOGRAFT\_REJECTION HALLMARK\_ANDROGEN\_RESPONSE HALLMARK\_ANGIOGENESIS HALLMARK\_APICAL\_JUNCTION HALLMARK\_APICAL\_SURFACE HALLMARK APOPTOSIS HALLMARK BILE ACID METABOLISM HALLMARK\_CHOLESTEROL\_HOMEOSTASIS HALLMARK\_COAGULATION HALLMARK\_COMPLEMENT HALLMARK\_DNA\_REPAIR HALLMARK\_E2F\_TARGETS HALLMARK\_EPITHELIAL\_MESENCHYMAL\_TRA NSITION HALLMARK\_ESTROGEN\_RESPONSE\_EARLY HALLMARK\_ESTROGEN\_RESPONSE\_LATE HALLMARK\_FATTY\_ACID\_METABOLISM

HALLMARK\_G2M\_CHECKPOINT HALLMARK\_GLYCOLYSIS HALLMARK\_HEDGEHOG\_SIGNALING HALLMARK\_HEME\_METABOLISM HALLMARK HYPOXIA HALLMARK\_IL2\_STAT5\_SIGNALING HALLMARK IL6 JAK STAT3 SIGNALING HALLMARK\_INFLAMMATORY\_RESPONSE HALLMARK\_INTERFERON\_ALPHA\_RESPONSE HALLMARK\_INTERFERON\_GAMMA\_RESPONSE HALLMARK\_KRAS\_SIGNALING\_DN HALLMARK\_KRAS\_SIGNALING\_UP HALLMARK\_MITOTIC\_SPINDLE HALLMARK\_MTORC1\_SIGNALING HALLMARK\_MYC\_TARGETS\_V1 HALLMARK\_MYC\_TARGETS\_V2 HALLMARK\_MYOGENESIS

HALLMARK\_NOTCH\_SIGNALING HALLMARK\_OXIDATIVE\_PHOSPHORYLATION HALLMARK\_P53\_PATHWAY HALLMARK\_PANCREAS\_BETA\_CELLS HALLMARK\_PEROXISOME HALLMARK\_PI3K\_AKT\_MTOR\_SIGNALING HALLMARK PROTEIN SECRETION HALLMARK REACTIVE OXIGEN SPECIES PA THWAY HALLMARK\_SPERMATOGENESIS HALLMARK\_TGF\_BETA\_SIGNALING HALLMARK\_TNFA\_SIGNALING\_VIA\_NFKB HALLMARK\_UNFOLDED\_PROTEIN\_RESPONSE HALLMARK\_UV\_RESPONSE\_DN HALLMARK\_UV\_RESPONSE\_UP HALLMARK\_WNT\_BETA\_CATENIN\_SIGNALING HALLMARK\_XENOBIOTIC\_METABOLISM

### Gene Set: HALLMARK\_KRAS\_SIGNALING\_UP

Standard name	HALLMARK_KRAS_SIGNALING_UP	~
Systematic name	M5953	
Brief description	Genes up-regulated by KRAS activation.	
Full description or abstract		11
Collection	H: hallmark gene sets	
Source publication		
Exact source		
Related gene sets	(show 14 founder gene sets for this hallmark gene set)	
External links		
Organism	Homo sapiens	
Contributed by	Arthur Liberzon (Broad Institute)	
Source platform	HUMAN_GENE_SYMBOL	
Dataset references	(show 5 hallmark refinement datasets)	
	(show 1 hallmark validation datasets)	
Download gene set	format: grp   text   gmt   gmx   xml	
Compute overlaps 👔	(show collections to investigate for overlap with this gene set)	
Compendia expression profiles 🛛	Human tissue compendium (Novartis) Global Cancer Map (Broad Institute) NCI-60 cell lines (National Cancer Institute)	
Advanced query	Further investigate these 200 genes	
Gene families 🔽	Categorize these 200 genes by gene family	
Show members	(show 200 members mapped to 200 genes)	
Version history	5.0: First introduced	

(	преготпарре	ed to 200 gene	5)
Original Member	Entrez Gene Id	Gene Symbol	Gene Description
ABCB1	5243	ABCB1	ATP-binding cassette, sub-family B (MD
ACE	1636	ACE	angiotensin I converting enzyme (pepti
ADAM17	6868	ADAM17	ADAM metallopeptidase domain 17
ADAM8	101	ADAM8	ADAM metallopeptidase domain 8
ADAMDEC1	27299	ADAMDEC1	ADAM-like, decysin 1
AKAP12	9590	AKAP12	A kinase (PRKA) anchor protein 12
AKT2	208	AKT2	v-akt murine thymoma viral oncogene ho
ALDH1A2	8854	ALDH1A2	aldehyde dehydrogenase 1 family, membe
ALDH1A3	220	ALDH1A3	aldehyde dehydrogenase 1 family, membe
AMMECR1	9949	AMMECR1	Alport syndrome, mental retardation, m
ANGPTL4	51129	ANGPTL4	angiopoietin-like 4
ANKH	56172	ANKH	ankylosis, progressive homolog (mouse)
ANO1	55107	ANO1	anoctamin 1, calcium activated chlorid
ANXA10	11199	ANXA10	annexin A10
APOD	347	APOD	apolipoprotein D
ARG1	383	ARG1	arginase, liver
ATG10	83734	ATG10	ATG10 autophagy related 10 homolog (S
AVL9	23080	AVL9	AVL9 homolog (S. cerevisiase)
BIRC3	330	BIRC3	baculoviral IAP repeat containing 3
BMP2	650	BMP2	bone morphogenetic protein 2
BPGM	669	BPGM	2,3-bisphosphoglycerate mutase
BTBD3	22903	BTBD3	BTB (POZ) domain containing 3
BTC	685	BTC	betacellulin

### Search Gene Sets

Search by keyword, collection, organism, or contributor. 12

See the Browse Gene Sets page for an alphabetical list of gene sets and collections, or to search by gene set name.

### Keywords:

### Search Filters:

#### KRAS

(supports boolean operators AND and OR, and wildcard searches with \*)

search

collection	organism	contributor
all collections	all organisms	all contributors
H: hallmark gene sets	Danio rerio	Aristoteles University of Thessaloniki
C1: positional gene sets	Homo sapiens	BioCarta
C2: curated gene sets	Macaca mulatta	Broad Institute
CGP: chemical and genetic perturbations	Mus musculus	Columbia University
CP: Canonical pathways	Rattus norvegicus	Dana-Farber Cancer Institute
CP:BIOCARTA: BioCarta gene sets		Giannina Gaslini Institute
CP:KEGG: KEGG gene sets		GO
CP:REACTOME: Reactome gene sets		Johns Hopkins University School of Medicine
C3: motif gene sets		KEGG

control-click to select multiple lines

\$

### click on rows to select gene sets, click a gene set name to view the gene set page

select all 305 0 gene sets selected Select An Action...

<< < 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 > >> 10 +

name	# genes	description	collections	organism	contributor
AACTTT_UNKNOWN	1890	Genes with promoter regions [-2kb,2kb] around transcription start site containing motif AACTTT. Motif does not match any known transcription factor	C3 TFT	Homo sapiens	Broad Institute
ABE_VEGFA_TARGETS_30MIN	29	Genes up-regulated in HUVEC cells (endothelium) at 30 min after VEGFA [GeneID=7422] stimulation.	C2 CGP	Homo sapiens	University of Washington
ACCATTT,MIR-522	160	Targets of MicroRNA ACCATTT, MIR-522	C3 MIR	Homo sapiens	Broad Institute
ACTGAAA,MIR-30A-3P,MIR- 30E-3P	201	Targets of MicroRNA ACTGAAA, MIR-30A-3P, MIR-30E-3P	C3 MIR	Homo sapiens	Broad Institute
AGCATTA,MIR-155	134	Targets of MicroRNA AGCATTA, MIR-155	C3 MIR	Homo sapiens	Broad Institute
ATAGGAA,MIR-202	102	Targets of MicroRNA ATAGGAA, MIR-202	C3 MIR	Homo sapiens	Broad Institute
ATATGCA, MIR-448	212	Targets of MicroRNA ATATGCA, MIR-448	C3 MIR	Homo sapiens	Broad Institute
ATGCAGT,MIR-217	115	Targets of MicroRNA ATGCAGT, MIR-217	C3 MIR	Homo sapiens	Broad Institute
BENPORATH_CYCLING_GENES	648	Genes showing cell-cycle stage-specific expression [PMID=12058064].	C2 CGP	Homo sapiens	Broad Institute
BIOCARTA_TEL_PATHWAY	18	Telomeres, Telomerase, Cellular Aging, and Immortality	C2 CP CP:BIOCARTA	Homo sapiens	BioCarta

### Investigate Gene Sets

Gain further insight into the biology behind a gene set by using the following tools:

- compute overlaps with other gene sets in MSigDB (more...)
- display the gene set expression profile based on a selected compendium of expression data (more...)
- categorize members of the gene set by gene families (more...)

### **Gene Identifiers**

### Compute Overlaps

- 🔲 H: hallmark gene sets 🚺 C1: positional gene sets 2
- C2: curated gene sets 2
- CGP: chemical and genetic perturbations 1
- CP: Canonical pathways 🖬
- CP:BIOCARTA: BioCarta gene sets 1
- CP:KEGG: KEGG gene sets 2
- CP:REACTOME: Reactome gene sets
- 🔲 C3: motif gene sets 😰
- MIR: microRNA targets TFT: transcription factor targets 1
- C4: computational gene sets 12
- CGN: cancer gene neighborhoods 2
- CM: cancer modules 2
- 🔲 C5: GO gene sets 🖬
- BP: GO biological process 1
- CC: GO cellular component 🖬
- MF: GO molecular function 12
- 🔲 C6: oncogenic signatures 🖬
- 🔲 C7: immunologic signatures 🖬
- show top 10 \$ genesets

with FDR q-value below 0.05

### compute overlaps

### **Compendia expression profiles**

- Human tissue compendium (Novartis)
- Global Cancer Map (Broad Institute)
- NCI-60 cell lines (National Cancer Institute)

display expression profile

### Gene families

show gene families

### Investigate Gene Sets

Gain further insight into the biology behind a gene set by using the following tools:

- compute overlaps with other gene sets in MSigDB (more...)
- display the gene set expression profile based on a selected compendium of expression data (more...)
- categorize members of the gene set by gene families (more...)

#### **Gene Identifiers**

### Compute Overlaps

KRAS	H: hallmark gene sets	i.
BRAF	C1: positional gene sets 2	
NRAS MAP2K1	C2: curated gene sets 2	
MAP2K2 MAPK1	✓ CGP: chemical and genetic perturbations	
CCND1	CP: Canonical pathways	
PIK3CA AKT1	CP:BIOCARTA: BioCarta gene sets	
PTEN	CP:KEGG: KEGG gene sets	
MTOR	CP:REACTOME: Reactome gene sets	
	C3: motif gene sets 2	
	MIR: microRNA targets 2	
	TFT: transcription factor targets 1	
	C4: computational gene sets 2	
	CGN: cancer gene neighborhoods 2	
	CM: cancer modules 2	
	🔲 C5: GO gene sets 🖬	
	BP: GO biological process 2	
	CC: GO cellular component 2	
1	MF: GO molecular function 2	
	C6: oncogenic signatures 2	1
	C7: immunologic signatures 2	
	show top 10 🗧 genesets	
	with FDR q-value below 0.05	
	compute overlaps	

### Compendia expression profiles

 Human tissue compendium (Novartis)
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 display expression profile

### Gene families

show gene families

### Compute Overlaps for Selected Genes

Collections	# Overlaps Shown	# Gene Sets in Collections	# Genes in Comparison (n)	# Genes in Universe (N)
С2, Н	10	4776	11	45956

Click the gene set name to see the gene set page. Click the number of genes [in brackets] to download the list of genes.

Color bar shading from light green to black, where lighter colors indicate more significant FDR q-values (< 0.05) and black indicates less significant FDR q-values (>= 0.05).

Save to: Excel | CenomeSpace

Gene Set Name [# Genes (K)]	Description	# Genes in Overlap (k)	k/K	p-value 🛐	FDR q-value 🛐
KEGG_GLIOMA [65]	Glioma	11		1.85 e <sup>-32</sup>	8.85 e <sup>-29</sup>
KEGG_PROSTATE_CANCER [89]	Prostate cancer	11		7.56 e <sup>-31</sup>	1.8 e <sup>-27</sup>
KEGG_ENDOMETRIAL_CANCER [52]	Endometrial cancer	10		1.5 e <sup>-29</sup>	2.39 e <sup>-26</sup>
KEGG_ACUTE_MYELOID_LEUKEMIA [60]	Acute myeloid leukemia	10		7.16 e <sup>-29</sup>	8.55 e <sup>-26</sup>
KEGG_MELANOMA [71]	Melanoma	10		4.39 e <sup>-28</sup>	4.19 e <sup>-25</sup>
REACTOME_SIGNALING_BY_FGFR [112]	Genes involved in Signaling by FGFR	10		5.37 e <sup>-26</sup>	4.28 e <sup>-23</sup>
KEGG_NON_SMALL_CELL_LUNG_CANCER [54]	Non-small cell lung cancer	9		1.16 e <sup>-25</sup>	7.91 e <sup>-23</sup>
REACTOME_SIGNALING_BY_FGFR_IN_DISEASE [127]	Genes involved in Signaling by FGFR in disease	10		1.98 e <sup>-25</sup>	1.18 e <sup>-22</sup>
REACTOME_NGF_SIGNALLING_VIA_TRKA_FROM_ OM_THE_PLASMA_MEMBRANE [137]	Genes involved in NGF signalling via TRKA from the plasma membrane	10		4.35 e <sup>-25</sup>	2.31 e <sup>-22</sup>
KEGG_PATHWAYS_IN_CANCER [328]	Pathways in cancer	11		2.07 e <sup>-24</sup>	9.19 e <sup>-22</sup>

### Gene/geneset overlap matrix

	Gene	KEGG_GLIOMA	<pre><cccrete_cancer< pre=""></cccrete_cancer<></pre>	KEGG_ENDOMETRIAL_CANCER	KEGG_ACUTE_MYELOID_LEUKEMIA	KEGG_MELANOMA	REACTOM E_SIGNALING_BY_FGFR	KEGG_NON_SMALL_CELL_LUNG_CANCER	REACTOM E_SIGNALING_BY_FGFR_IN_DISEASE	REACTOME_NGF_SIGNALLING_VIA_TRKA_FROM_THE_PLASMA_MEMBRANE	KEGG_PATHWAYS_IN_CANCER	Entrez	rce	Gene
Entrez Gene Id	Symbol	KEG	Ĕ	Ξ.	Ξ.		~	2	2	~	Ξ.	E	Source	Description
		KEG	Ĕ	X	ž	Z	~	¥	2	2	_	S E		
Gene Id	Symbol	KEG	KE	¥	¥	Z	~	¥	2	R			S	Description
<b>Gene Id</b> 5594 5604	Symbol MAPK1	KEG	KE	KE	X	¥	~	×	R	R		w w w	S S	Description mitogen-activated protein kinase 1
<b>Gene Id</b> 5594 5604 3845	Symbol MAPK1 MAP2K1	KEG	KE	K	N N N N N N N N N N N N N N N N N N N	×		×	R.	R		w w w w	S S S	Description mitogen-activated protein kinase 1 mitogen-activated protein kinase kinase 1
<b>Gene Id</b> 5594 5604 3845 4893	Symbol MAPK1 MAP2K1 KRAS	KEG	KE	KE	KE	×	~		R			22 22 23 23 23 23 23 23	S S S S S	Description mitogen-activated protein kinase 1 mitogen-activated protein kinase kinase 1 v-Ki-ras2 Kirsten rat sarcoma viral oncogene homolog neuroblastoma RAS viral (v-ras) oncogene homolog mitogen-activated protein kinase kinase 2
<b>Gene Id</b> 5594 5604 3845 4893 5605 207	Symbol MAPK1 MAP2K1 KRAS NRAS MAP2K2 AKT1			KE	KE			¥					S S S S S S	Description mitogen-activated protein kinase 1 mitogen-activated protein kinase kinase 1 v-Ki-ras2 Kirsten rat sarcoma viral oncogene homolog neuroblastoma RAS viral (v-ras) oncogene homolog mitogen-activated protein kinase kinase 2 v-akt murine thymoma viral oncogene homolog 1
Gene Id 5594 5604 3845 4893 5605 207 5290	Symbol MAPK1 MAP2K1 KRAS NRAS MAP2K2 AKT1 PIK3CA		KE	KE	KE								S S S S S S S	Description mitogen-activated protein kinase 1 mitogen-activated protein kinase kinase 1 v-Ki-ras2 Kirsten rat sarcoma viral oncogene homolog neuroblastoma RAS viral (v-ras) oncogene homolog mitogen-activated protein kinase kinase 2 v-akt murine thymoma viral oncogene homolog 1 phosphoinositide-3-kinase, catalytic, alpha polypeptide
Gene Id 5594 5604 3845 4893 5605 207 5290 673	Symbol MAPK1 MAP2K1 KRAS NRAS MAP2K2 AKT1 PIK3CA BRAF		KE	KE	KE								S S S S S S S S S	Description mitogen-activated protein kinase 1 mitogen-activated protein kinase kinase 1 v-Ki-ras2 Kirsten rat sarcoma viral oncogene homolog neuroblastoma RAS viral (v-ras) oncogene homolog mitogen-activated protein kinase kinase 2 v-akt murine thymoma viral oncogene homolog 1 phosphoinositide-3-kinase, catalytic, alpha polypeptide v-raf murine sarcoma viral oncogene homolog B1
Gene         Id           5594         5594           5604         3845           3845         605           207         5290           573         595	Symbol MAPK1 MAP2K1 KRAS NRAS MAP2K2 AKT1 PIK3CA BRAF CCND1			KE	KE								S S S S S S S S S S S	Description mitogen-activated protein kinase 1 mitogen-activated protein kinase kinase 1 v-Ki-ras2 Kirsten rat sarcoma viral oncogene homolog neuroblastoma RAS viral (v-ras) oncogene homolog mitogen-activated protein kinase kinase 2 v-akt murine thymoma viral oncogene homolog 1 phosphoinositide-3-kinase, catalytic, alpha polypeptide v-raf murine sarcoma viral oncogene homolog B1 cyclin D1
Gene Id 5594 5604 3845 4893 5605 207 5290 573	Symbol MAPK1 MAP2K1 KRAS NRAS MAP2K2 AKT1 PIK3CA BRAF			KE									S S S S S S S S S S S	Description mitogen-activated protein kinase 1 mitogen-activated protein kinase kinase 1 v-Ki-ras2 Kirsten rat sarcoma viral oncogene homolog neuroblastoma RAS viral (v-ras) oncogene homolog mitogen-activated protein kinase kinase 2 v-akt murine thymoma viral oncogene homolog 1 phosphoinositide-3-kinase, catalytic, alpha polypeptide v-raf murine sarcoma viral oncogene homolog B1

### **Gene Identifiers Compute Overlaps** 🔲 H: hallmark gene sets 🔽 KRAS BRAF C1: positional gene sets 🖬 NRAS MAP2K1 C2: curated gene sets 1 MAP2K2 CGP: chemical and genetic perturbations 1 MAPK1 CCND1 CP: Canonical pathways 1 **РІКЗСА** CP:BIOCARTA: BioCarta gene sets 1 AKT1 PTEN CP:KEGG: KEGG gene sets MTOR CP:REACTOME: Reactome gene sets 1 🔲 C3: motif gene sets 🔽 MIR: microRNA targets 2 TFT: transcription factor targets 1 C4: computational gene sets 12 🔲 CGN: cancer gene neighborhoods 😰 CM: cancer modules 1 🔲 C5: GO gene sets 🖬 BP: GO biological process 1 CC: GO cellular component 1 MF: GO molecular function 1 C6: oncogenic signatures 1 🔲 C7: immunologic signatures 😰 show top 10 \$ genesets with FDR q-value below 0.05 compute overlaps

# Compendia expression profiles Human tissue compendium (Novartis) Global Cancer Map (Broad Institute) NCI-60 cell lines (National Cancer Institute) display expression profile

### Gene families

show gene families



Key to lineages: gene families

kinases oncogenes translocated genes tumor suppressors

### View Gene Families for Selected Genes

The following table provides a functional overview of the MSigDB gene sets by categorizing their genes into a small number of carefully chosen "gene families". To categorize the genes in a gene set, use the gene set page or the Investigate Gene Sets page.

	cytokines and growth factors	transcription factors	homeodomain proteins	cell differentiation markers	protein kinases	translocated cancer genes	oncogenes	tumor suppressors
tumor suppressors	0	0	0	0	0	0	0	1
oncogenes	0	0	0	0	2	2	6	
translocated cancer genes	0	0	0	0	1	2		
protein kinases	0	0	0	0	6			
cell differentiation markers	0	0	0	0				
homeodomain proteins	0	0	0					
transcription factors	0	0						
cytokines and growth factors	0							

Click on a gene family or gene family intersection to retrieve annotations for those genes.

Members of these "gene families" share a common feature such as homology or biochemical activity. They do not necessarily have common origins. For the source of each "gene family" definition, click here.

#### Curated Gene Signatures (GeneSigDB) http://genesigdb.org/genesigdb/

				igDB Signatures		
Home	Browse	Analyze My C	ienes	Download	Support	Contact Us
Publication Se Search the full text of ar gene signatures they de as author name, article	ticles to retrieve a list o escribe. Enter one or m	ore search terms, such		Gene Search ? Search gene annotations signatures.		ed in GeneSigDB gene

The **Gene Signature DataBase** is a searchable database of fully traceable, standardized, annotated gene signatures which have been manually curated from publications that are indexed in <u>PubMed</u>. Enter a search term above to get started.

News	GeneSigDB Data Release 4
September, 2011: GeneSigDB Data and Website Update We continue to expand. So far we have read and processed almost 3,000 publications to extract 3,515 genes signatures from 1,604 publications. See <u>GeneSigDB Release 4 release notes</u> We have a new tag cloud <u>Browse</u> feature to enable easy browsing of GeneSigDB. Additional <u>download</u> formats. Download GeneSigDB as an R/Bioconductor data file, gmt or compressed flat file formats.	Gene Signatures: 3515 Published Articles: 1604 Genes (Human): 20,523 Tissues and Diseases: More than 50 Species: 3

#### DSigDB http://tanlab.ucdenver.edu/DSigDB

Bioinformatics, 31(18), 2015, 3069–3071 doi: 10.1093/bioinformatics/btv313 Advance Access Publication Date: 19 May 2015 Applications Note

OXFORD

Systems biology

#### DSigDB: drug signatures database for gene set analysis

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<sup>†</sup>The authors wish it to be known that, in their opinion, the first two authors should be regarded as Joint First Authors. Associate Editor: Jonathan Wren

Received on February 17, 2015; revised on April 30, 2015; accepted on May 13, 2015

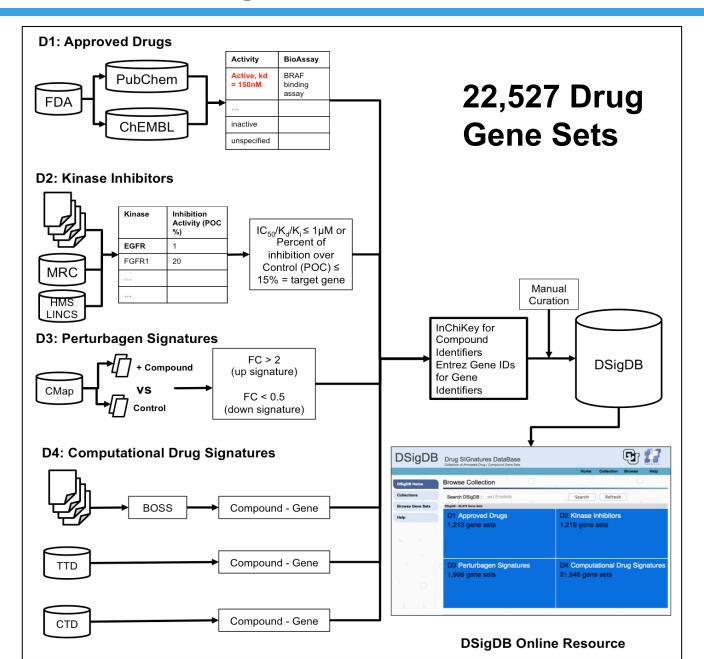
#### Abstract

**Summary:** We report the creation of Drug Signatures Database (DSigDB), a new gene set resource that relates drugs/compounds and their target genes, for gene set enrichment analysis (GSEA). DSigDB currently holds 22527 gene sets, consists of 17389 unique compounds covering 19531 genes. We also developed an online DSigDB resource that allows users to search, view and download drugs/compounds and gene sets. DSigDB gene sets provide seamless integration to GSEA software for linking gene expressions with drugs/compounds for drug repurposing and translational research.

**Availability and implementation:** DSigDB is freely available for non-commercial use at http://tan lab.ucdenver.edu/DSigDB.

Supplementary information: Supplementary data are available at *Bioinformatics* online. Contact: aikchoon.tan@ucdenver.edu

### **DSigDB** Workflow

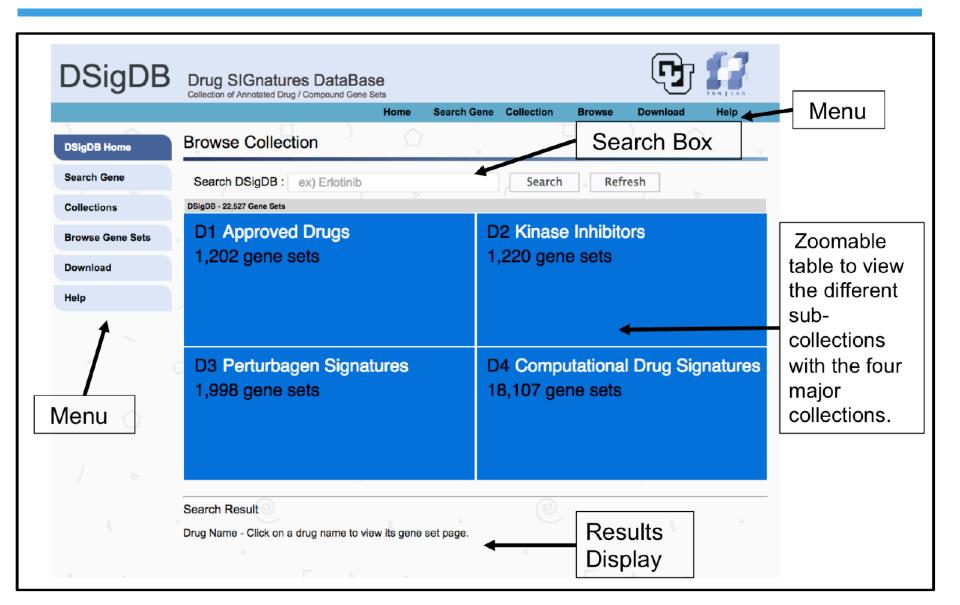


### DSigDB

DSigDB	Drug SIGnatures DataBase Collection of Annotated Drug / Compound Gene Sets	
	Home	Search Gene Collection Browse Download Help
DSigDB Home	Browse Collection	
Search Gene	Search DSigDB : ex) Erlotinib	Search Refresh
Collections	DSigDB - 22,527 Gene Sets	
Browse Gene Sets	D1 Approved Drugs	D2 Kinase Inhibitors
Download	1,202 gene sets	1,220 gene sets
Help		
	D3 Perturbagen Signatures 1,998 gene sets	D4 Computational Drug Signatures 18,107 gene sets

#### http://tanlab.ucdenver.edu/DSigDB/

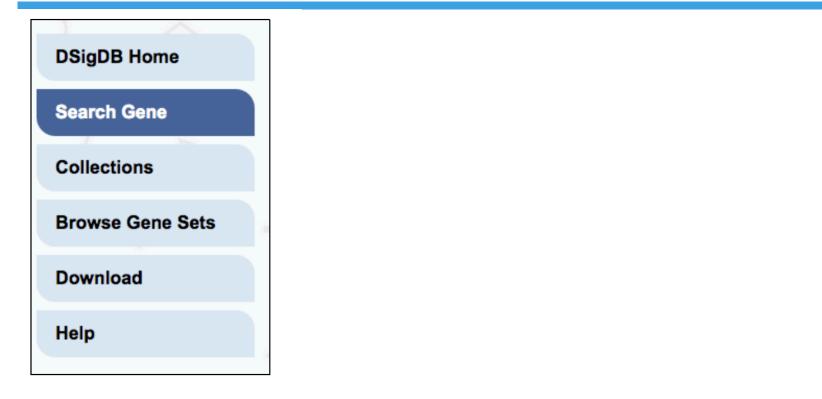
### DSigDB



## **Drug Search**

Search DSig[	DB : Erlotinib		Search	
SigDB - 22,527 Gen		1 2		
	oved Drugs		D2 Kinase Inhibitors	
1,202 ge	ne sets		1,220 gene sets	
				0'
	rbagen Sign	latures	D4 Computational Drug	Signatures
1,998 ge	ne sets		18,107 gene sets	
earch Result	0			
	k on a drug name to	view its gene set page.		
	k on a drug name to Source	view its gene set page. Representative Name	Synonym	
rug Name - Clic	-		Synonym Erlotinib Hydrochloride	•
rug Name - Clic Collection	Source	Representative Name		•
rug Name - Clic Collection D1	Source	Representative Name Erlotinib Hydrochloride	Erlotinib Hydrochloride	- - -
rug Name - Clic Collection D1	Source D1 FDA	Representative Name Erlotinib Hydrochloride Erlotinib	Erlotinib Hydrochloride Erlotinib	
rug Name - Clic Collection D1	Source D1 FDA Kinome Scan	Representative Name Erlotinib Hydrochloride Erlotinib Erlotinib	Erlotinib Hydrochloride Erlotinib Erlotinib	
rug Name - Clic Collection D1 D2	Source D1 FDA Kinome Scan RBC	Representative Name Erlotinib Hydrochloride Erlotinib Erlotinib Erlotinib	Erlotinib Hydrochloride Erlotinib Erlotinib Erlotinib	
rug Name - Clic Collection D1 D2	Source D1 FDA Kinome Scan RBC BOSS	Representative Name Erlotinib Hydrochloride Erlotinib Erlotinib Erlotinib Erlotinib	Erlotinib Hydrochloride Erlotinib Erlotinib Erlotinib Erlotinib	

#### Gene Search



Search Gene (19,531) :	ex) EGFR		Search		
	1 2	T	bturth		
	Source	c	Chemical Name		
		c our gene name	Chemical Name	3	•
now 15 ¢ entries Gene ▲ Gene		our gene name	Chemical Name	3	•

#### Gene Search Result

Search Gene	1			$\bigcirc$	
Search Gene (19,5	31): EGFR		Search		
Show 15 ¢ entries					
Gene	Source		Chemical Name		*
EGFR	D1	chlorpromazine			
EGFR	D1	afatinib			
EGFR	D1	thioridazine			
EGFR	D1	vandetanib			
EGFR	D1	baciguent			
EGFR	D1	levodopa			
EGFR	D1	hexachlorophene			
EGFR	D1	zafirlukast			
EGFR	D1	erlotinib hydrochloride			
EGFR	D1	miconazole			
EGFR	D1	tamoxifen			
EGFR	D1	crystal violet			
EGFR	D1	methyldopa			
EGFR	D1	dobutamine			
EGFR	D1	crizotinib			
Gene	Source		Chemical Name		
Page 1 of 42 ( Total 616 D	Data Sets)	P	revious 1 2 3	4 5 42	Next

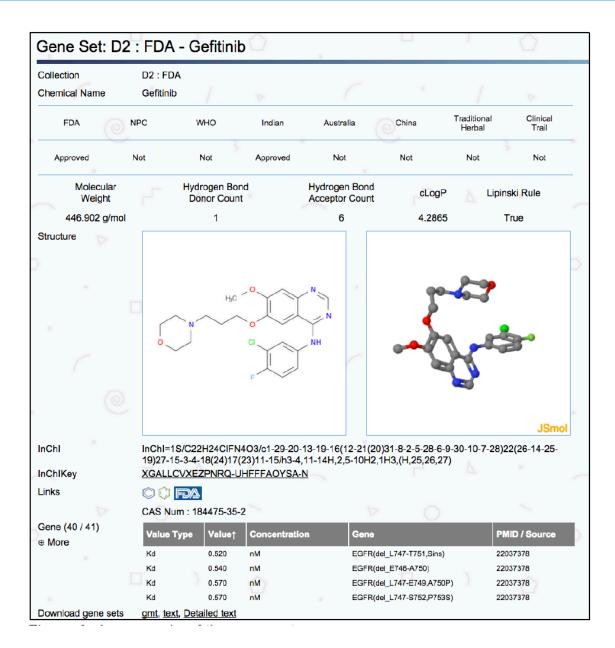
#### **Browsing Collection**

<b>(A)</b>	DSigDB - 22,527 Gene Sets / D2	7	1	/ >>
$\langle \gamma \gamma \rangle$	FDA		HMS LINCS	
	28 gene sets		90 gene sets	<b>1</b>
	MRC	Roche	)	
	157 gene sets	570 ge	ene sets	
		Kinom	e Scan	RBC
	GSK	72 gei	ne sets	99 gene sets
	204 gene sets			
	Ŭ			
<b>(B)</b>	DSigDB - 22,527 Gene Sets / D2			
(-)	FDA		HMS LINCS	
	28 gene sets		90 gene sets	
	MRC	Roche	1	
	157 gene sets	570 gene sets		
			e Scan	RBC
	GSK	72 ger	ne sets	99 gene sets
	GSK 204 gene sets	72 ger	ie sets	99 gene sets
		72 ger	ne sets	99 gene sets

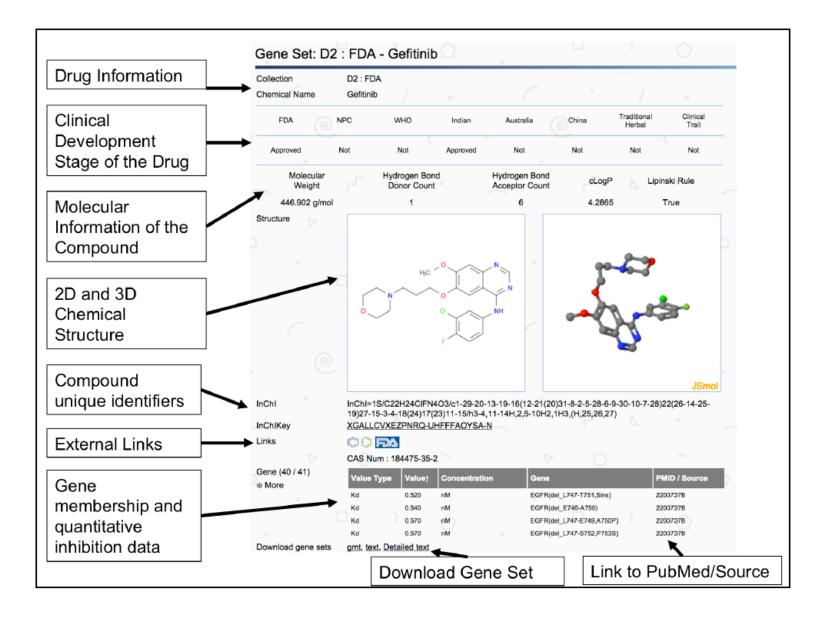
### **Browsing Collection**

FDA 28 gene set	ts 🐂	6	HMS LING 90 gene s			
MRC 157 gene sets			Roche 570 gene sets			
GSK 204 gene sets			ie Scan ne sets	RBC 99 gene sets		
			0			
Search Result : FD Drug Name - Click on	A a drug name to view its ge	ene set page.	٢	3	•	

#### **Compound Webpage**



#### **Compound Webpage**



#### Collections

Collection	Description	Unique Number of Genes	Number of Gene Sets	Download
DSigDB	All Gene Sets.	19,531	22,527	GMT File
D1 : FDA Approved ( browse 1,202 gene sets )	FDA Approved Drug Gene Sets.	1,288	1,202	GMT File
D2 : Kinase Inhibitors	Kinase Inhibitors Gene Sets based on in vitro kinase profiling assays.	407	1,220	GMT File
FDA ( browse 28 gene sets )	FDA Approved Kinase Inhibitors.	341	28	GMT File
HMS LINCS ( browse 90 gene sets )	Kinase inhibition assays extracted from HMS LINCS database.	381	90	GMT File
MRC ( browse 157 gene sets )	Kinase inhibition assays extracted from MRC Kinome Inhibition database.	137	157	GMT File
GSK ( browse 204 gene sets )	GSK Published Kinase Inhibitor Set (PKIS), kinase inhibitors used as chemical probes.	116	204	GMT File
Roche ( browse 570 gene sets )	Kinase Inhibitors profiled by Roche.	153	570	GMT File
RBC ( browse 99 gene sets )	Kinase Inhibitors profiled by Reaction Biology Corporation.	246	99	GMT File
KinomeScan ( browse 72 gene sets )	Kinase Inhibitors profiled by DiscoveryRx using KinomeScan technology.	374	72	GMT File
D3 : Perturbagen Signatures ( browse 1,998 gene sets )	7,064 gene expression profiles from three cancer cell lines perturbed by 1,309 compounds from CMap (build 02).	11,137	1,998	GMT File
CMAP ( browse 1,998 gene sets )	7,064 gene expression profiles from three cancer cell lines perturbed by 1,309 compounds from CMap (build 02).	11,137	1,998	GMT File
D4 : Computational Drug Signatures	Drug signatures extracted from literatures using a mixture of manual curation and by automatic computational approaches.	18,854	18,107	GMT File
BOSS ( browse 2,114 gene sets )	Text mining approach of drug-gene targets using Biomedical Object Search System (BOSS).	3,354	2,114	GMT File
CTD ( browse 5,163 gene sets )	Curation of targets from Comparative Toxicogenomics Database (CTD).	18,700	5,163	GMT File
TTD ( browse 10,830 gene sets )	Manual curation of targets from the Therapeutics Targets Database (TTD).	1,389	10,830	GMT File

#### **Download Collections**

Download	
DSigDB provides several	options for downloading the data.
Current Release The current data release	of DSigDB is Release 1 (released May 2015).
DSigDB Release 1 (Upda	ited)
<ul> <li>DSigDBv1.0.gmt</li> <li>DSigDBv1.0.txt</li> <li>DSigDBv1.0 Detail</li> </ul>	led.txt
	Drug Gene Type Source
¥	GefitinibEGFRKd=40.0(nM)FDAGefitinibEGFRKd=0.54(nM)FDA
Compound : Gefitinib	Gefitinib EGFR Kd=0.98(nM) FDA
EPHA6	Gefitinib ABL1 Kd=460.0(nM) FDA
	Gefitinib CDK7 Kd=610.0(nM) FDA
STK10	Gefitinib EGFR Kd=140.0(nM) FDA
MKNK1	Gefitinib ABL1 Kd=680.0(nM) FDA
EGFR	GefitinibABL1Kd=360.0(nM)FDAGefitinibLCKKd=630.0(nM)FDA
RIPK2	Gefitinib ABL1 Kd=480.0(nM) FDA
	Gefitinib MKNK1 Kd=290.0(nM) FDA
MAP2K5	Gefitinib SBK1 Kd=560.0(nM) FDA
HIPK4	Gefitinib SLK Kd=920.0(nM) FDA
ABL1	GefitinibEGFRKd=1.1(nM)FDAGefitinibABL1Kd=230.0(nM)FDA
FLT3	Gefitinib IRAK4 Kd=540.0(nM) FDA
CSNK1E	Gefitinib ERBB3 Kd=790.0(nM) FDA
	Gefitinib GAK Kd=13.0(nM) FDA
GAK	Gefitinib ABL1 Kd=780.0(nM) FDA
LYN	GefitinibLYNKd=990.0(nM)FDAGefitinibIRAK1Kd=69.0(nM)FDA
IRAK1	GefitinibIRAK1Kd=69.0(nM)FDAGefitinibCHEK2Kd=800.0(nM)FDA

# Use Case Example: EGFRwt NSCLC

				Gefitinib IC50
Cell line	Histology	EGFR	KRAS	(µmol/L)
Sensitive				
H358	BAC	Wild-type	Mutant	0.18
H322	BAC	Wild-type	Wild-type	0.25
Calu-3	Adenocarcinoma	Wild-type	Wild-type	0.3
H1334	Large	Wild-type	Wild-type	0.3
H1648	Adenocarcinoma	Wild-type	Wild-type	0.38
HCC78	Adenocarcinoma	Wild-type	Wild-type	0.4
H2126	Large	Wild-type	Wild-type	1
HCC193	Adenocarcinoma	Wild-type	Wild-type	1.5
HCC95	Adenocarcinoma	Wild-type	Wild-type	1.9
Resistant				
H125	Adenosquamous	Wild-type	Wild-type	4.8
HCC44	Adenocarcinoma	Wild-type	Mutant	7.9
H1703	Squamous	Wild-type	Wild-type	8
HCC15	Squamous	Wild-type	Wild-type	9.4
A549	Adenocarcinoma	Wild-type	Wild-type	9.6
H157	Squamous	Wild-type	Mutant	12.8
H460	Large	Wild-type	Mutant	12.9
H520	Squamous	Wild-type	Wild-type	13.6
H1299	Large	Wild-type	Wild-type	14.7

(Adapted from Coldren et al MCR 2006)

# **Enriched Gene Sets**

Inhibiting EGER/ERBR2/ERBR3

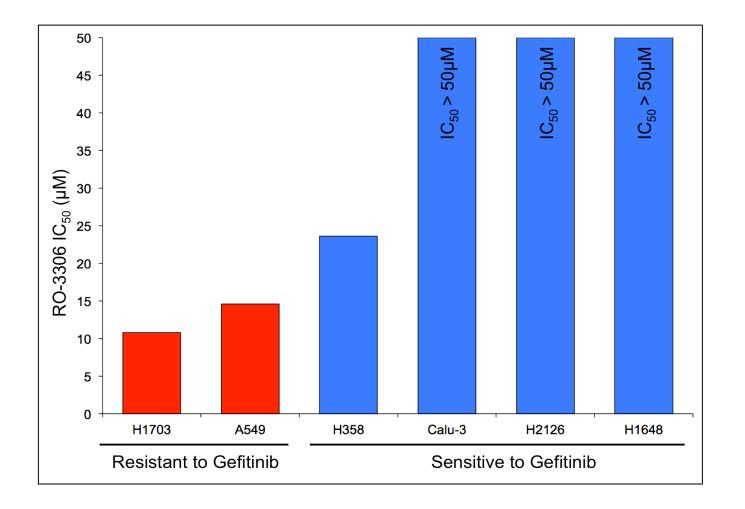
#### Enriched in Sensitive Group (p < 0.05)

	Normalized GENE Enrichment Nominal				based on Kinase Inhibition		
GENE SET NAME	GENE SET SIZE	Enrichment Score	Nominal p-val	Intended targets	EGFR	Assays ERBB2	ERBB3
CI-1033 KINOME SCAN	28	1.78	0.0000	EGFR/ERBB2	Yes	Yes	Yes
AZD-9291 LINCS	43	1.63	0.0000	EGFR	Yes	Yes	No
ZM-447439_LINCS	41	1.55	0.0285	AURKA	Yes	Yes	Yes
AZD-2171 KINOME SCAN	42	1.52	0.0271	VEGFR2/PDGFRA/PDGFRB	Yes	No	Yes
SB-203580_KINOME SCAN	18	1.52	0.0496	p38-alpha	Yes	No	No
WH-4-023 LINCS	124	1.48	0.0101	LCK	Yes	Yes	Yes
PP-242_KINOME SCAN	111	1.48	0.0116	MTOR/PIK3CA	Yes	Yes	No
CABOZANTINIB_FDA	45	1.47	0.0313	VEGFR2,MET	No	No	No
VANDETANIB_FDA	51	1.47	0.0355	RET/VEGFR2/EGFR	Yes	No	Yes
HG-9-91-01_LINCS	137	1.46	0.0179	SIK1	Yes	Yes	Yes
AZ-628_LINCS	51	1.46	0.0265	BRAF	Yes	No	No
VANDETANIB_KINOME SCAN	51	1.45	0.0448	RET/VEGFR2/EGFR	Yes	No	Yes
EXEL-2880/GSK-1363089_KINOME SCAN	131	1.45	0.0147	MET/AXL/VEGFR2	Yes	No	Yes
PD-173955_KINOME SCAN	105	1.40	0.0305	ABL1/SRC	Yes	No	Yes
BOSUTINIB_LINCS	69	1.40	0.0490	ABL1/SRC	Yes	Yes	Yes
R406_KINOME SCAN	183	1.39	0.0123	SYK,FLT3	Yes	No	No

#### Enriched in Resistant Group (p < 0.05)

	Normalized				Inhibiting EGFR/ERBB2/ERBB3 based on Kinase Inhibition		
GENE SET NAME	GENE SET SIZE	Enrichment Score	Nominal p-val	Intended targets	EGFR	Assays ERBB2	ERBB3
KINOME 858 ROCHE	17	-1.81	0.0041	NA	No	No	No
KINOME_1901_ROCHE	17	-1.67	0.0149	NA	No	No	No
CHEMBL2062936 ROCHE	16	-1.66	0.0042	NA	No	No	No
KINOME 1242 ROCHE	16	-1.62	0.0198	NA	No	No	No
KINOME 1221 ROCHE	19	-1.62	0.0194	NA	No	No	No
KINOME 866 ROCHE	15	-1.56	0.0395	NA	No	No	No
RO-3306_MRC	29	-1.45	0.0455	CDK1	No	No	No

## RO-3306 Sensitivity (From GDSC)



# Take home message

- Genes don't act alone to drive biological processes
- Gene set analysis such as GSEA can identify set of coordinately and subtle expressed genes participated in a functional group compared to Candidate Gene Analysis
- Biology trumps statistics if you can validate the gene sets