

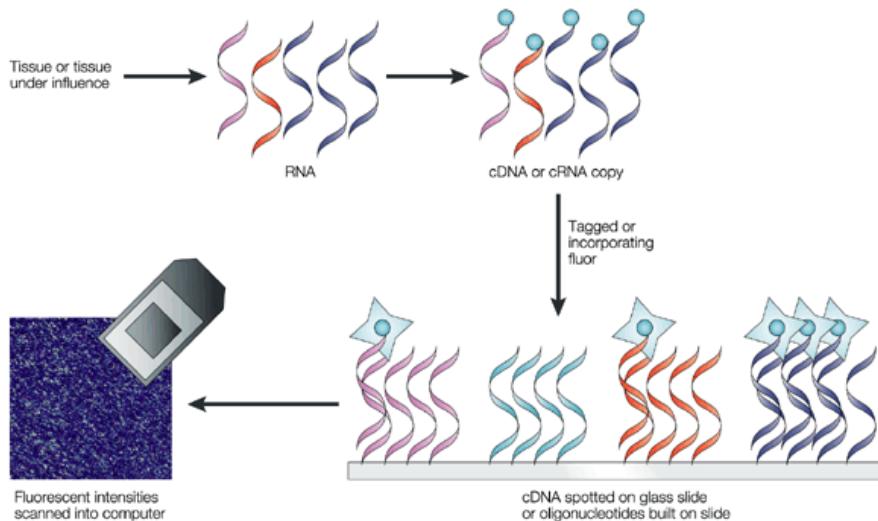
# Gene expression databases

Sean Eddy, PhD

# Outline

- How to measure gene expression?
  - Microarrays/RNA-seq
- What are gene expression databases?
- Which ones exist?
- How do they differ?
- How can they be used?
- What needs to be taken care of?
- What can I do with specialized databases?

# Microarrays: the beginning of high throughput gene expression

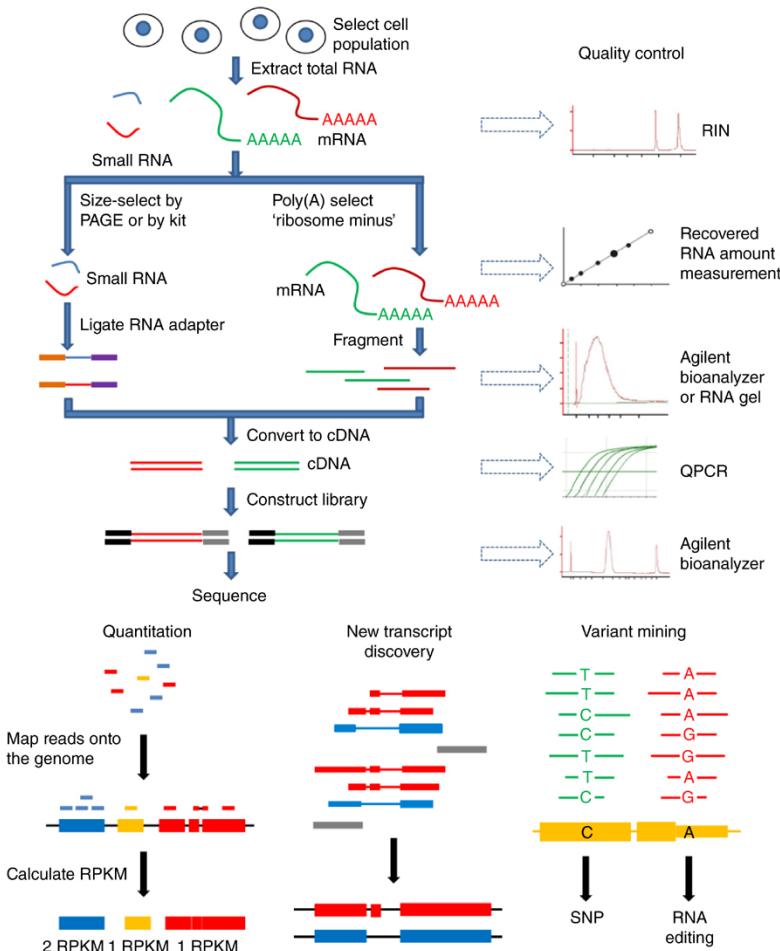


- Compartmentalized chips with sequences bound to the surface
- Sample is applied to surface
- Hybridizes to complementary sequence
- Hybridizations are quantified
  - No binding, “no” signal
  - Can only find what is specifically searched for
  - Usually represented as  $n \times m$  matrix

<http://www.nature.com/nrd/journal/v1/n12/images/nrd961-f1.gif>

Nature Reviews | Drug Discovery

# RNA-seq: the future of high throughput gene expression



- Samples (cDNA libraries) are applied to a sequencer
- Millions of sequences are generated and mapped onto a genome
- Sequence reads are quantified
  - No sequence = no expression
  - Can find novel transcripts, splice isoforms, fusion genes, etc.
- Quality of mapping depends largely on library prep and decisions made on how RNA is initially processed.

# Gene expression databases

- Repositories for gene expression data
  - Mostly microarray and now RNAseq
  - Primarily for storage
  - Curated or un-curated
  - Access to data on different levels:
    - Datasets
    - Individual levels
- Integrated databases
  - Contain array data and additional data of the samples
  - Array data tends to be more annotated
  - More analytical tools
  - Smaller (more QC and curation needed)
  - Often no direct data access

# Why do they exist

- Transparency/reproducibility of publications
  - Journals require data to be available for analysis
  - Nowadays raw data is required
  - Databases offer single resource and standardized access
- Data was generated for a specific purpose, but is not limited to that purpose
  - Can be reanalyzed in a different context
  - Can be combined with other datasets
  - Can be used as independent validation

# Gene expression repository examples

- Gene expression omnibus ([www.ncbi.nlm.nih/geo/](http://www.ncbi.nlm.nih/geo/))
  - [1,117,462](#) samples, [3848](#) datasets

The screenshot shows the GEO homepage with a blue header bar containing the NCBI logo, Resources, How To, Sign in to NCBI, GEO Home, Documentation, Query & Browse, and Email GEO links. Below the header is a search bar with 'Keyword or GEO Accession' and a 'Search' button. The main content area has a title 'Gene Expression Omnibus' and a brief description: 'GEO is a public functional genomics data repository supporting MIAME-compliant data submissions. Array- and sequence-based data are accepted. Tools are provided to help users query and download experiments and curated gene expression profiles.' To the right is the GEO logo and text.

**Gene Expression Omnibus**

GEO is a public functional genomics data repository supporting MIAME-compliant data submissions. Array- and sequence-based data are accepted. Tools are provided to help users query and download experiments and curated gene expression profiles.

**GEO**  
Gene Expression Omnibus

Keyword or GEO Accession

- Array express ([www.ebi.ac.uk/arrayexpress/](http://www.ebi.ac.uk/arrayexpress/))
  - [ArrayExpress – functional genomics data](#)
- Princeton University MicroArray database (PUMAdb)
  - 40084 experiments, 6598 made public
- NCBI SRA, ENA and Princeton HTseq for NGS data

# What is in a gene expression database?

- Gene expression data in different forms:
  - Resolution:
    - Gene level
    - Transcript level
    - Exon level
      - And / or raw data
  - Comprehensiveness
    - Targeted arrays
    - Whole genome arrays
  - Different platforms (microarrays, RNAseq)
- Generally only gene expression, may have limited sample information

# Where does the data come from?

- Expression profiles of
  - Patients
  - Model systems
  - Cell cultures
- Data used for publication
  - Most journals now require raw data submission
  - Very coarse quality control (peer review)
  - QC depends mostly on authors
- Datasets submitted without publication
  - Little or no QC
- Most datasets are tailored towards a specific question

# Example: GEO GSE32591

- Go to <http://www.ncbi.nlm.nih.gov/geo/>
- Enter GSE32591 into search box
- Click on “Analyze with GEO2R”
  - How would you set up the groups for analysis?
  - What do you get?
    - Does that make sense? How can results be verified?
- Go to “value distribution” tab
  - What do you see?
  - What are possible explanations?

GEO2R

Value distribution

Options

Profile graph

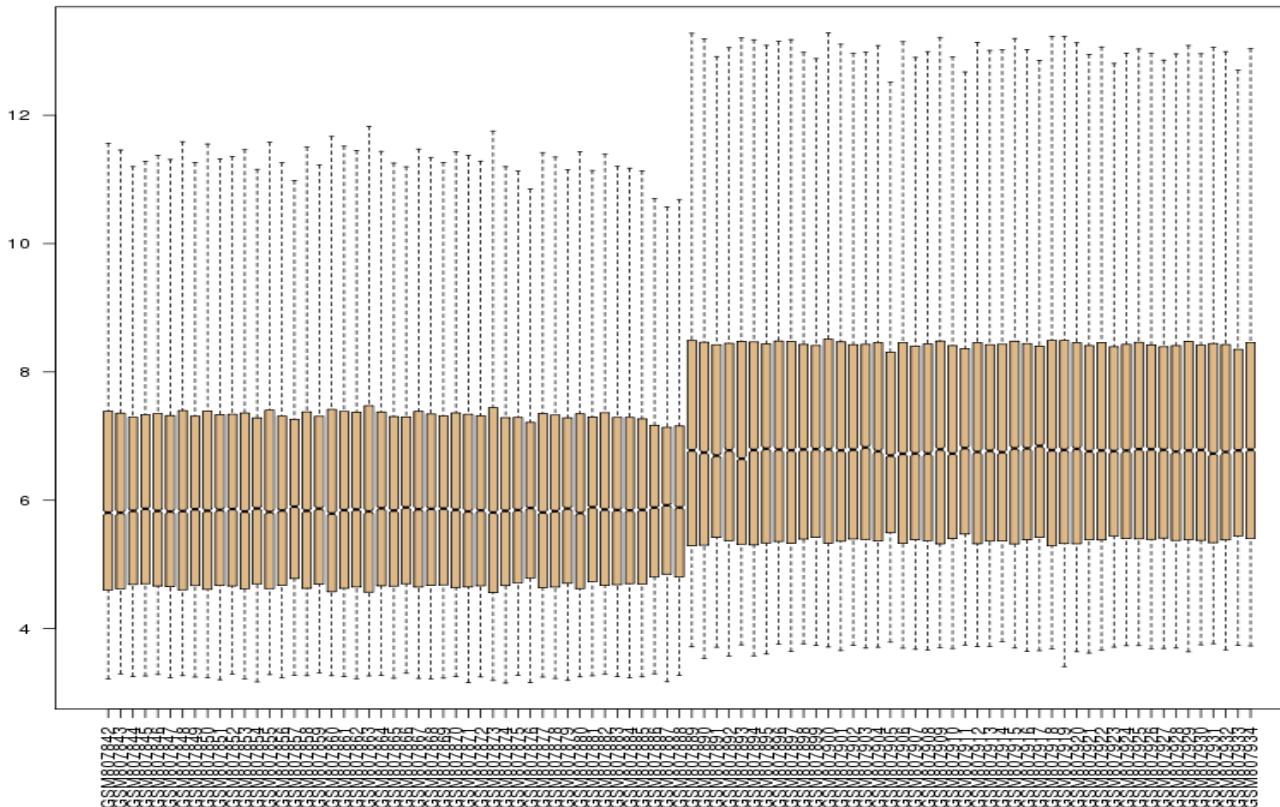
R script

Calculate the distribution of value data for the Samples you have selected. Distributions may be viewed graphically as a box plot or exported as a [number summary](#) table. The plot is useful for determining if value data are median-centered across Samples, and thus suitable for cross-comparison. [More...](#)

View

Export

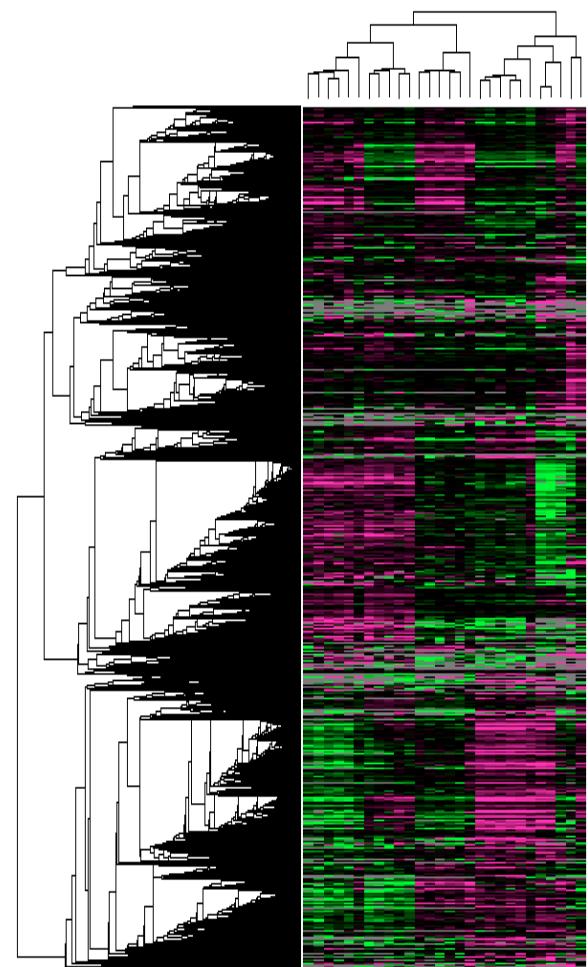
GSE32591/GPL14663, selected samples



# What can be done with GEO?

# What can be done with GEO?

- Programmatic access for data download
  - [http://www.ncbi.nlm.nih.gov/geo/info/geo\\_paccess.html](http://www.ncbi.nlm.nih.gov/geo/info/geo_paccess.html) (GEO)
  - [http://www.ebi.ac.uk/arrayexpress/help/programmatic\\_access.html](http://www.ebi.ac.uk/arrayexpress/help/programmatic_access.html) (ArrayExpress)
- Pre-computed analyses and on the fly analyses
  - Search by gene across all GEO experiments
  - Search by experiment to retrieve cluster analysis
  - Search by gene sequence for matching expression profiles
    - Described by Barret and Edgar, Methods Mol. Biol. 2006 “Mining Microarray Data at NCBI’s Gene Expression Omnibus (GEO)”
      - <http://www.ncbi.nlm.nih.gov/pmc/articles/PMC1619899/>



# What questions can be answered?

# What questions can be answered?

- If you download: anything
  - Only limited by your knowledge, skills, resources
- Pre-computed results
  - Preselected analysis methods/ sample groups
  - Generally within one dataset
- On-the-fly analyses
  - Sets of genes that cluster in under conditions given
  - Sample properties may not be entirely transparent.

What can be answered by doing it  
yourself?

# What can be answered by doing it yourself?

- The quality of the data
  - Is part of the data low quality?
  - Does some of the data not fit into the set (e.g. batch effect, outliers for other reasons)
  - Is it adequately processed?
- What is the relationship between expression data and non-expression variables?
  - How does my gene (of interest) associated with experimental treatments, clinical parameters?
- What are patterns across datasets?
  - Does my finding hold up across similar analyses in independent datasets?

# Why do you have to do it yourself?

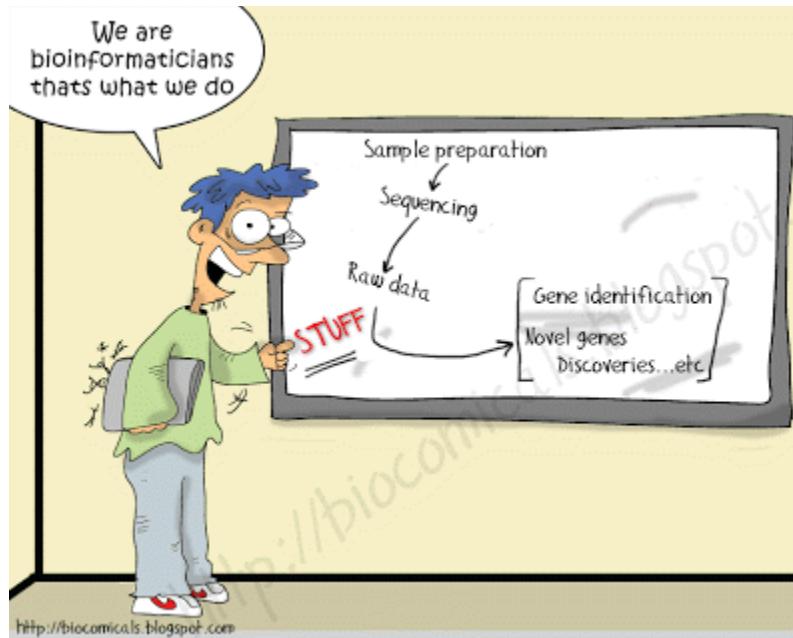
- Quality control:
  - QC parameters are often glossed over in papers and in microarray submissions
  - For Affymetrix QC modules are available, freely available and widely accepted in the bioinformatic community
  - Other array types have distinct, but also similar properties
    - <http://www.nature.com/nbt/focus/maqc/index.html>
- Relations to non-expression data variables
  - Data is often not standardized within fields

# Why not?

- Analysis across datasets:
  - Because:.... How?
  - Need to find a common standard for identification
  - Values need to be made comparable
    - If absolute expression values used, dynamic range can be a problem
    - If ratios used, information about expression level lost
  - Non-expression data even worse

# Who is the target group for doing it yourself?

- Users with experience in expression data
  - Crucial information (**STUFF**) is missing



# Why is this a problem?

- Excludes investigators with good hypotheses but lacking bioinformatic skills

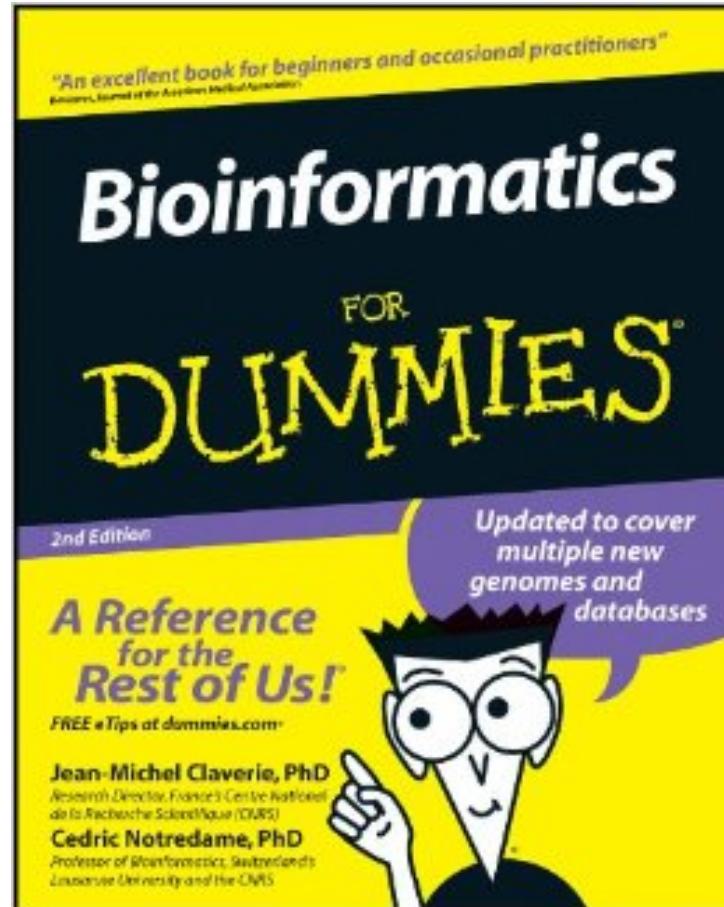


"Must be a clinical fellow."



"The computer says I need to upgrade my brain  
to be compatible with microarray data analysis."

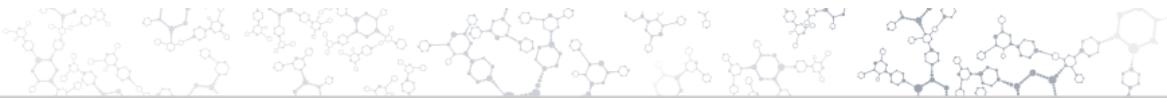
# How to fix that?



# How to fix that?

- Specialized databases
  - Datasets are easier to find
    - Datasets relevant to specific areas are collected in one place
      - NephroSeq for renal disease
      - Oncomine for cancer
  - Datasets are standardized and expertly curated
    - Controlled vocabulary is introduced for non-expression data
    - Curation of expression possible by introducing standardized references and data transformations across datasets
      - Gene IDs/Gene Symbols as references
      - Z-transformation or median centering of log transformed expression data

# Nephroseq ( [www.nephroseq.org](http://www.nephroseq.org) )



[Contact](#)

login

USER ID:

PASSWORD:

[Forgot password?](#)  
[Not a user? Register now!](#)

Login

about

Nephroseq is supported by the Applied Systems Biology Core as part of the University of Michigan O'Brien Renal Center. The primary goal of the Applied Systems Biology Core is to provide to the renal research community a platform for integrative data mining of comprehensive renal disease gene expression data sets, in order to:

1. Define molecular characteristics/features in the circulation or kidney and associate them with known disease phenotypes so as to obtain a better understanding of the pathophysiology of a specific renal disease

2. Identify markers of disease progression and treatment response (i.e., biomarkers)

## Welcome to Nephroseq

Developed for the renal research community, Nephroseq is a platform for integrative data mining of genotype/phenotype data, with optimized workflows that lead from search to visualization and from question to answer to next question:

- ▶ The expression of a gene is highly correlated with well-known podocyte genes. Is the gene functionally important in glomeruli?
- ▶ A gene is significantly differentially expressed in a subset of disease patients. Is the gene associated with a certain phenotype, severity or sub-category of the disease?
- ▶ A set of genes is significantly up-regulated in disease patients. Are the disease genes inversely related to the target profile of a compound/drug?

### ABOUT NEPHROSEQ

Originally a collaborative effort, Nephroseq is now solely developed and maintained by the Applied Systems Biology Core at the University of Michigan. This resource combines a wealth of publicly available renal gene expression profiles - gathered and curated by an experienced team of data scientists, bioinformaticians, and nephrologists - with a sophisticated analysis engine and powerful web application designed for data mining and visualization of gene expression data.

Nephroseq provides researchers with a rich set of publicly available renal gene expression data, packaged with the tools and interface necessary to analyze it, all aimed at seeking answers to questions and advancing a molecular understanding of kidney disease to ultimately improve clinical outcomes.

In particular, Nephroseq provides unique access to datasets from the Personalized Molecular Nephrology Research Laboratory incorporating clinical data which is often difficult to collect from public sources.



# Oncomine ([www.oncomine.com](http://www.oncomine.com))

# [www.oncomine.org](http://www.oncomine.org))

ONCOMINE™

upgrade @ contact

login

USER ID:

PASSWORD:

• Forgot password?  
• Not a user? Register now!

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Design better experiments.  
Gain more insights.  
Prepare to publish faster.

With Oncomine™ Research Premium Edition, you can:

**Design better experiments...** Answer more questions with fewer experiments, select the most promising gene or cell line, and test your hypothesis.

**Gain more biological insights...** Discover novel targets for therapeutic development, interrogate gene expression profiles, and identify drug and biological interactions.

**Prepare to publish faster...** Validate your results faster, visualize data easier and make connections to clinical significance.

The Oncomine™ Platform—from web applications to translational bioinformatics services—provides solutions for individual researchers and multinational companies, with robust, peer-reviewed analysis methods and a powerful set of analysis functions that compute gene expression signatures, clusters and gene-set modules, automatically extracting biological insights from the data. It has become an industry-standard tool cited in more than 1,100 peer-reviewed journal articles. The Oncomine Platform has been used as a foundation for ground-breaking discoveries with unique features that include:

- **Scalability** – with 700+ independent datasets
- **High quality** – with expertly curated data
- **Consistency** – with a rich, extensive and controlled ontology of terms
- **Standardized analysis** – with conventions that assure clear and consistent interpretations of results

Oncomine Research Edition remains free to the academic and nonprofit cancer research communities.

iontorrent  
by Thermo Fisher Scientific

The Origin of the Oncomine Platform  
The Oncomine Platform was conceived by physicians, scientists, and software engineers at the University of Michigan. It was commercialized by Arul Chinnaiyan and Dan Rhodes in February 2006 with the goal of building a version that would have a greater ability to impact drug development and clinical practice.

# NephroSeq and Oncomine

- Pros:
  - Each focus on one area of interest
  - Clinical data for many individual samples available
  - Advanced analysis using integrated systems biology tools in a pre-defined automated manner
  - Meta analysis possible
  - User friendly, free accessible for academic users
  - **Hypotheses-generating**
- Cons:
  - No raw data download
  - No programmatic access
  - Only predefined analyzes

# NephroSeq main Page

? HELP

The screenshot shows the NephroSeq main page with several annotations:

- A large red arrow points to the "Primary Filters" sidebar on the left.
- A green arrow points to the "Analysis type" section in the center.
- A pink box highlights the text "26 datasets (2000 samples)" in the center.
- A red box highlights the "Help" button in the top right corner.
- A green box highlights the "Analysis type" heading in the center.
- A blue box highlights the "Coexpression analysis", "Differential analysis", and "Outlier analysis" sections in the center.
- A yellow box highlights the "Welcome to NephroSeq!" text at the bottom left.
- A grey box highlights the "Analyses that are available include:" list at the bottom left.
- A light blue box highlights the "In addition, users have the ability to upload gene lists to use as filters and export data and visualizations directly to Excel, PowerPoint and SVG." text at the bottom left.
- A grey box highlights the "support" and "press releases" sections on the right.
- A grey box highlights the "events" section on the right.

**26 datasets (2000 samples)**

## Analysis type

- Coexpression analysis
- Differential analysis
- Outlier analysis

Welcome to NephroSeq!

This application is a web-based analysis engine for molecular biology researchers and clinician scientists who study renal disease and related disorders. NephroSeq gives access to renal genome-wide gene expression datasets generated by the renal research community. This tool is especially powerful because the data are already pre-analyzed and datasets include clinical data. The 3-paneled user interface moves users from left to right within the application to choose data, sort and prioritize analyses, and visualize and export results.

Analyses that are available include:

- Differential Expression: Identify over- or under-expression for a particular gene
- Coexpression: View genes that are coordinately expressed with your gene of interest across a dataset
- Outlier: Identify outlier patterns where a gene is highly over-expressed in a fraction of samples
- Concept Association: Identify significant overlap between gene sets that represent underlying biology

In addition, users have the ability to upload gene lists to use as filters and export data and visualizations directly to Excel, PowerPoint and SVG.

**support**  
scientific and password/account management  
support is available from: support@nphroseq.org

**press releases**  
JANUARY 2016  
**Identification of urinary protein biomarker for Chronic Kidney Disease**  
<http://www.ncbi.nlm.nih.gov/pubmed/26631632>

**events**  
**mark your calendar**  
**11th International Podocyte Conference**  
APRIL 3rd - 6th  
HAIFA AND JERUSALEM, ISRAEL  
<http://podocyte2016.org>

**ISN Nexus Symposium: Translational Immunology in Kidney Disease**  
APRIL 14th - 17th  
BERLIN, GERMANY  
<http://www.isnnexus.org/berlin>

# Two Search Options

- Gene specific search:
  - Gene
- Dataset search:
  - Specific conditions/diseases

NPHS2: encodes podocin,  
a podocyte specific protein

# Gene Search

## Gene summary view

Nephroseq

Welcome, Sean Eddy. Tools Help Logout

search NPHS2

filter selected 11 datasets (795 samples)

Gene: NPHS2 Threshold (P-value): 1E-4 Threshold (Fold Change): 2 Threshold (Gene Rank): Top 10% Data Type: All

Disease Summary Demographics

Dataset/Disease type

Primary Filters

- Analysis Type
  - Coexpression Analysis (11)
  - Differential Analysis (11)
    - + Demographics Analysis (10)
    - Group Analysis (6)
    - + Indices Analysis (1)
    - + Pathology Analysis (8)
    - Tissue Type Analysis (5)
    - + Treatment Analysis (1)
    - Outlier Analysis (11)
- + Group
- + Donor Type
- + Tissue Type
- + Dataset Type

Sample Filters

Dataset Filters

Concept Filters

Demographics

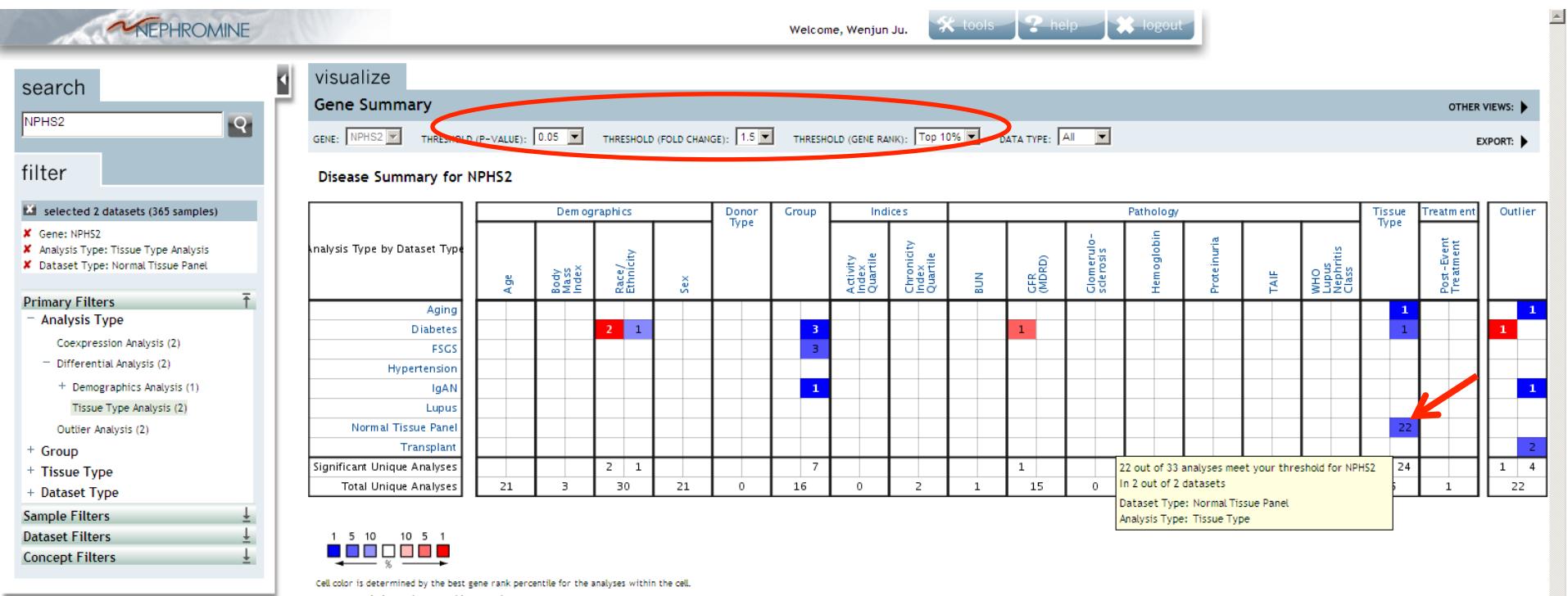
Age	Body Mass Index	Race/Ethnicity	Sex	Donor Type	Group	Indices		Pathology				Tissue Type	Treatment	Outlier			
						Activity Index Quartile	Chronicity Index Quartile	BUN	GFR (mRD)	Glomerulo-Sclerosis	Hemoglobin				Proteinuria	TAIF	WHO Lupus Nephritis Class
21	3	30	21	0	16	1	0	2	1	15	0	0	7	2	0	36	1

1 5 10 10 5 1

Cell color is determined by the best gene rank percentile for the analyses within the cell.  
NOTE: An analysis may be counted in more than one cancer type.

# Gene Search

## Gene summary view



22 out of 33 analysis meet your threshold for NPHS2 in 2 out of 2 datasets

# Four Basic Analysis Modes

- Differential expression
- Co-expression analysis
- Outlier analysis
  - Heterogeneity within predefined groups
- Concepts analysis
  - Gene set (Nephromine & third-party sources)

# Gene Search

## Differential expression (Box graph)

NEPHROMINE

Welcome, Wenjun Ju. tools help logout

search: NPHS2

filter: selected 2 datasets (365 samples)

- Gene: NPHS2
- Analysis Type: Tissue Type Analysis
- Dataset Type: Normal Tissue Panel

Primary Filters

- Analysis Type
  - Coexpression Analysis (2)
  - Differential Analysis (2)
    - + Demographics Analysis (1)
    - Tissue Type Analysis (2) →
  - Outlier Analysis (2)
- + Group
- + Tissue Type
- + Dataset Type

Sample Filters

Dataset Filters

Concept Filters

datasets concepts

ORDER BY: Under-expression: Gene Rank

ON: NPHS2

THRESHOLD BY:

P-VALUE	FOLD CHANGE	GENE RANK
0.05	1.5	Top 10%

visualize Differential Analysis

GROUP BY: Tissue Type

SHOW: Only Samples in Analysis

OTHER VIEWS: ▶

NPHS2 Expression in Lindenmeyer Normal Tissue Panel

Tissue Type: Glomeruli vs. Tubulointerstitium

Lindenmeyer Normal Tissue Panel Statistics

Under-expression Gene Rank: 694 (in top 6%)  
P-value: 5.68E-8  
Reporter: 220424\_at  
t-Test: -26.283  
Fold Change: -10.623

log2 median-centered intensity

Legend

1. Glomeruli (6)
2. Tubulointerstitium (6)

Lindenmeyer Normal Tissue Panel

PLOS One 2010/07/12 12 samples NPHS2 Information  
mRNA 12,624 measured genes Reporter Information  
Human Genome U133A Array

Reference

# THE HUMAN PROTEIN ATLAS



ABOUT & HELP

SEARCH ? »

NPHS2

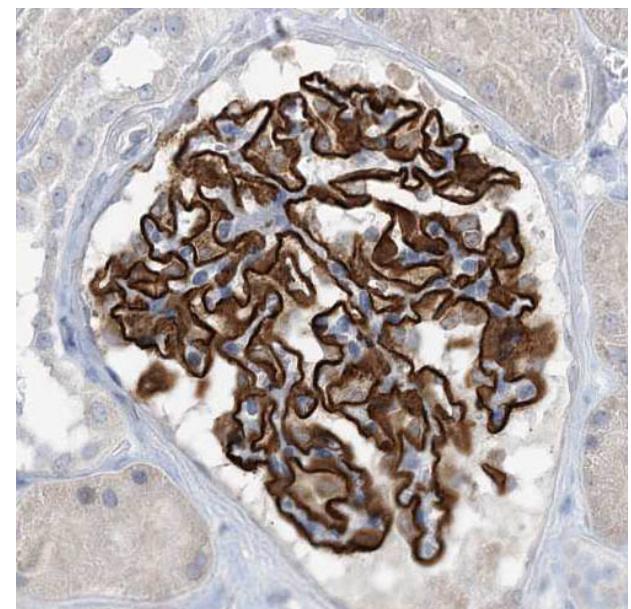
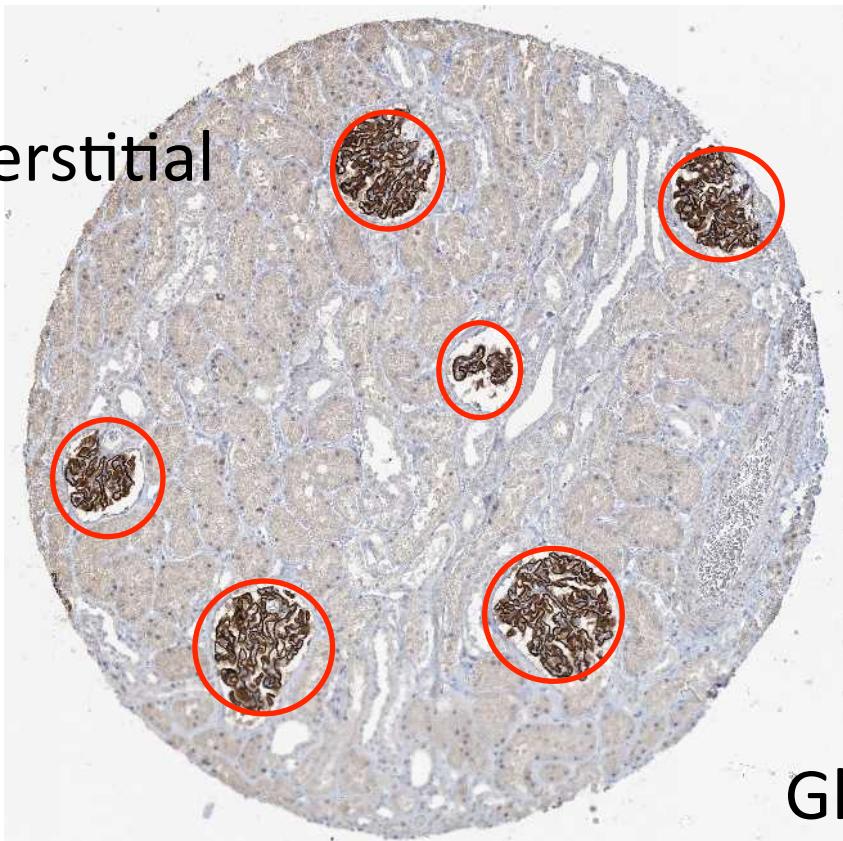
e.g. CD44, ELF3, KLK3, or use Fields to search specific fields such as  
protein\_class:Transcription factors or chromosome:X

Search

Clear

Fields »

Tubulointerstitial



Glomeruli

# Gene Search

## Correlation with clinical continuous variation

search  
VCAN  
filter

Gene: VCAN  
Analysis type: GFR  
Dataset Type: Diabetes

selected 2 datasets (66 samples)  
Gene: VCAN  
Analysis Type: GFR (MDRD) Analysis  
Dataset Type: Diabetes

Primary Filters  
Analysis Type  
Coexpression Analysis (2)  
Differential Analysis (2)  
Demographics Analysis (2)  
Age Analysis (2)  
Race/Ethnicity Analysis (1)  
Sex Analysis (2)  
Group Analysis (2)  
Pathology Analysis (2)  
GFR (MDRD) Analysis (2)  
Proteinuria Analysis (1)  
Tissue Type Analysis (1)  
Outlier Analysis (2)

+ Group  
+ Tissue Type  
+ Dataset Type

Sample Filters  
Dataset Filters  
Dataset Name  
Schmid Diabetes (1)  
Woronecka Diabetes (1)

Concept Filters

visualize  
Differential Analysis  
GROUP BY: GFR (MDRD) (Tubulointerstitium)  
SHOW: Only Samples in Analysis

VCAN Expression in Woronecka Diabetes  
Tubulointerstitium: GFR (MDRD)

Woronecka Diabetes Statistics  
Under-expression Gene Rank: 12 (in top 1%)  
Reporter: 221731\_x\_at  
P-value: 7.71E-7  
Correlation: -0.832

P=7.71E-7  
Correlation: -0.832

Legend  
1. < 15 ml/min/1.73m<sup>2</sup> (3)  
2. 15 - 29 ml/min/1.73m<sup>2</sup> (4)  
3. 30 - 59 ml/min/1.73m<sup>2</sup> (7)  
4. 60 - 89 ml/min/1.73m<sup>2</sup> (5)  
5. > 90 ml/min/1.73m<sup>2</sup> (3)

Legend  
1. Less Than 15 ml/min/1.73m<sup>2</sup> (3)  
2. 15 - 29 ml/min/1.73m<sup>2</sup> (4)  
3. 30 - 59 ml/min/1.73m<sup>2</sup> (7)  
4. 60 - 89 ml/min/1.73m<sup>2</sup> (5)  
5. More Than 90 ml/min/1.73m<sup>2</sup> (3)

Woronecka Diabetes  
Diabetes 2011/09/01 44 samples  
mRNA 12,603 measured genes  
Human Genome U133A 2.0 Array  
VCAN Information  
Reporter Information

# Gene Search

## Outlier analysis

Outlier analysis helps to identify an expression profile where differential pattern is only seen in a fraction of samples of all patients within a disease type.

**Why do we need it:** 25% of patients show over-expression of a gene. This gene may not generate a significant p-value in a t-test comparing DN relative to normal kidney.

**How to do it:** Transform all samples within a dataset, so that genes could be ranked by their expression from high to low. The data transformation is performed at certain percentile bins (75, 90 & 95%), and a line is drawn at the percentile of that analysis to define outliers.

For example, in an outlier analysis at the 75th percentile, the system draws a line at the point at which only the top 25th percentile samples extend above it.

# Gene Search

## Outlier analysis

search

filter

selected 2 datasets (66 samples)

Gene: VCAN

Analysis Type: Outlier Analysis

Dataset Type: Diabetes

Primary Filters

- Analysis Type
  - Coexpression Analysis (2)
  - Differential Analysis (2)
    - + Demographics Analysis (2)
    - Group Analysis (2)
    - + Pathology Analysis (2)
    - Tissue Type Analysis (1)
  - Outlier Analysis (2)
- + Group
- + Donor Type
- + Tissue Type
- + Dataset Type

Sample Filters

Dataset Filters

Concept Filters

datasets concepts

ORDER BY: Outlier: Over-expression

ON: VCAN

THRESHOLD BY:

P-VALUE 1E-4	FOLD CHANGE 2	GENE RANK Top 10%
-----------------	------------------	----------------------

Compare | Clear All

Schmid Diabetes (22)

Outlier 75th%	7
Outlier 90th%	33
Outlier 95th%	89

Woroniecka Diabetes (44)

Outlier 90th%	32
Outlier 95th%	34
Outlier 75th%	65

visualize

### Outlier Analysis

GROUP BY: Group

SHOW: All Samples in Dataset

#### VCAN Expression in Schmid Diabetes

Outlier Analysis at 75th Percentile, Grouped by Group

#### Schmid Diabetes Statistics

Over-expression Gene Rank: 7 (in top 1%) COPA: 4.228

Reporter: 204620\_s\_at

log<sub>2</sub> median-centered intensity

75%

Controls

Diabetic

Legend

- 1. Cadaveric Donor Control (4)
- 2. Healthy Living Donor (3)
- 3. Minimal Change Disease (4)
- 4. Diabetic Nephropathy (11)

#### Schmid Diabetes

Diabetes 2006/11/01 22 samples VCAN Information

mRNA 12,624 measured genes Reporter Information

Human Genome U133A Array

# Differential expression – Dataset search

search

filter  selected 16 datasets (1121 samples)

Analysis Type: Differential Analysis

**Primary Filters**

- Analysis Type
  - Coexpression Analysis (16)
  - Differential Analysis (16)** (highlighted with red circle)
  - + Demographics Analysis (14)
  - Donor Type Analysis (1)
  - Group Analysis (9)
  - + Indices Analysis (3)
  - + Pathology Analysis (11)
  - Tissue Type Analysis (8)
  - + Treatment Analysis (1)
  - Outlier Analysis (16)
- + Group
- + Donor Type
- + Tissue Type
- + Dataset Type

Sample Filters

Dataset Filters

Concept Filters

datasets concepts

ORDER BY: Dataset Name

ON:

Compare | Clear All

**Flechner Transplant (62)**

- Cadaveric Donor Kidney Specimen: Acute Rejection vs. No Rejection (highlighted with red box)
- Cadaveric Donor Kidney Specimen: Age
- Cadaveric Donor Kidney Specimen: GFR (MDRD)
- Cadaveric Donor Kidney Specimen: Sex
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: Acute Rejection vs. No Rejection
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: Age
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: GFR (MDRD)
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: Renal Dysfunction vs. No Rejection
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: Sex
- Cadaveric Donor Tissue Type: Kidney vs. Peripheral Blood Lymphocyte
- Kidney Specimen Donor Type: Living vs. Cadaveric
- Living Donor Kidney Specimen: Age
- Living Donor Kidney Specimen: GFR (MDRD)
- Living Donor Kidney Specimen: Renal Dysfunction vs. No Rejection
- Living Donor Kidney Specimen: Sex
- Living Donor Peripheral Blood Lymphocyte Specimen: Age
- Living Donor Peripheral Blood Lymphocyte Specimen: GFR (MDRD)
- Living Donor Peripheral Blood Lymphocyte Specimen: Renal Dysfunction vs. No Rejection
- Living Donor Peripheral Blood Lymphocyte Specimen: Sex
- Living Donor Tissue Type: Kidney vs. Peripheral

visualize

**Differential Analysis**

GROUP BY: Group (Cadaveric Donor Kidney Specimen)

SHOW: Only Samples in Analysis

OTHER VIEWS: + PRIMARY CONCEPT: EXPORT:

Over-expression

**Comparison of All Genes in Flechner Transplant**

Over-expression in Cadaveric Donor Kidney Specimen: Acute Rejection vs. No Rejection (log2 median-centered intensity)

Rank	P-value	Fold Change	Gene	Reporter	Gene
1	2.26E-8	1.51	NIPAL3	37850_at	NIPAL3
2	2.20E-7	1.41	STXBP5L	34130_at	STXBP5L
3	3.33E-7	1.51	KSR1	1716_at	KSR1
4	3.53E-7	1.51	SRC	1938_at	SRC
5	4.04E-7	1.68	LLGL1	804_s_at	LLGL1
6	4.45E-7	1.73	FSTL4	34518_at	FSTL4
7	6.21E-7	1.43	ARID3A	35913_at	ARID3A
8	7.71E-7	2.02	CYP1A1	1024_at	CYP1A1
9	7.93E-7	1.44	RIN1	1777_at	RIN1
10	9.63E-7	1.63	PDGFB	1573_at	PDGFB
11	1.10E-6	2.14	SKI	1918_at	SKI
12	1.28E-6	1.52	COL11A2	1027_at	COL11A2
13	1.36E-6	1.81	PTCH1	836_at	PTCH1
14	1.93E-6	1.63	ZNF646	39863_at	ZNF646
15	2.13E-6	1.55	COL2A1	598_at	COL2A1
16	2.22E-6	1.42	KISS1	1645_at	KISS1
17	2.29E-6	1.39	TBL3	41603_at	TBL3
18	2.31E-6	1.49	CYP2C18	1477_s_at	CYP2C18
19	2.44E-6	1.47	MAST1	35962_at	MAST1
20	2.52E-6	1.34	PCDHGC3	657_at	PCDHGC3
21	3.14E-6	1.40	BRF1	141_s_at	BRF1

Legend

- 1. No Rejection (5)
- 2. Acute Rejection (6)

Least Expressed

Most Expressed

Not measured

Note: Colors are z-score normalized to depict relative values within rows. They cannot be used to compare values between rows.

**Flechner Transplant**

Am J Transplant 2004/09/01 62 samples

mRNA 8,603 measured genes

Human Genome U95A-Av2 Array

Export

# Differential expression – dataset search – compare analysis

- Compare different analyzes
- Data is standardized on upload (centered to 0 and standardized by variance)
- all features are mapped to common identifier (EntrezGeneID)

**search**

**datasets**    **concepts**

ORDER BY: Dataset Name

ON:

**Compare | Close All**

**Flechner Transplant (62)**

- Cadaveric Donor Kidney Specimen: Acute Rejection vs. No Rejection
- Cadaveric Donor Kidney Specimen: Age
- Cadaveric Donor Kidney Specimen: GFR (MDRD)
- Cadaveric Donor Kidney Specimen: Sex
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: Acute Rejection vs. No Rejection
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: Age
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: GFR (MDRD)
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: Renal Dysfunction vs. No Rejection
- Cadaveric Donor Peripheral Blood Lymphocyte Specimen: Sex
- Cadaveric Donor Tissue Type: Kidney vs. Peripheral Blood Lymphocyte
- Kidney Specimen Donor Type: Living vs. Cadaveric
- Living Donor Kidney Specimen: Age
- Living Donor Kidney Specimen: GFR (MDRD)
- Living Donor Kidney Specimen: Renal Dysfunction vs. No Rejection
- Living Donor Kidney Specimen: Sex
- Living Donor Peripheral Blood Lymphocyte Specimen: Age
- Living Donor Peripheral Blood Lymphocyte Specimen: GFR (MDRD)
- Living Donor Peripheral Blood Lymphocyte Specimen: Renal Dysfunction vs. No Rejection
- Living Donor Peripheral Blood Lymphocyte Specimen: Sex
- Living Donor Tissue Type: Kidney vs. Peripheral

**filter**

selected 16 datasets (1121 samples)

Analysis Type: Differential Analysis

**Primary Filters**

- Analysis Type
  - Coexpression Analysis (16)
  - Differential Analysis (16)
    - + Demographics Analysis (14)
    - Donor Type Analysis (1)
    - Group Analysis (9)
    - + Indices Analysis (3)
    - + Pathology Analysis (11)
    - Tissue Type Analysis (8)
    - + Treatment Analysis (1)
  - Outlier Analysis (16)
- + Group
- + Donor Type
- + Tissue Type
- + Dataset Type

**Sample Filters**

**Dataset Filters**

**Concept Filters**

**visualize**

**Differential Analysis**

GROUP BY: Group (Cadaveric Donor Kidney Specimen)

SHOW: Only Samples in Analysis

+ PRIMAR

1 | 2 | 3 | 4 | 5 | <

**Comparison of All Genes in Flechner Transplant**

Over-expression in Cadaveric Donor Kidney Specimen: Acute Rejection vs. No Rejection (log2 median-centered intensity)

Rank	P-value	Fold Change	Gene	Reporter	Gene
1	2.26E-8	1.51	NIPAL3	37850_at	NIPAL3
2	2.20E-7	1.41	STXBP5L	34130_at	STXBP5L
3	3.33E-7	1.51	KSR1	1716_at	KSR1
4	3.53E-7	1.51	SRC	1938_at	SRC
5	4.04E-7	1.68	LLGL1	804_s_at	LLGL1
6	4.45E-7	1.73	FSTL4	34518_at	FSTL4
7	6.21E-7	1.43	ARID3A	35913_at	ARID3A
8	7.71E-7	2.02	CYP1A1	1024_at	CYP1A1
9	7.93E-7	1.44	RIN1	1777_at	RIN1
10	9.63E-7	1.63	PDGF8	1573_at	PDGF8
11	1.10E-6	2.14	SKI	1918_at	SKI
12	1.28E-6	1.52	COL11A2	1027_at	COL11A2
13	1.36E-6	1.81	PTCH1	836_at	PTCH1
14	1.93E-6	1.63	ZNF646	39863_at	ZNF646
15	2.13E-6	1.55	COL2A1	598_at	COL2A1
16	2.22E-6	1.42	KISS1	1645_at	KISS1
17	2.29E-6	1.39	TBL3	41603_at	TBL3
18	2.31E-6	1.49	CYP2C18	1477_s_at	CYP2C18
19	2.44E-6	1.47	MAST1	35962_at	MAST1
20	2.52E-6	1.34	PCDHGC3	657_at	PCDHGC3
21	3.14E-6	1.40	BRF1	141_s_at	BRF1

Legend

1. No Rejection (5)

2. Acute Rejection (6)

# Meta analysis

- Find out which genes are significantly more expressed in glomeruli compared to tubulointerstitium
- Can you verify that with another dataset?
- Or with more than one other dataset?
- Does it matter if the datasets are different?
- Can you imagine a use of this functionality for an exclusive filter (NOT)

# Example

Firefox heise online - IT-News, Nachrichten ... Millionärfacher Datenklau: Provider s... Re: Ich versteh's nicht - erklärt's mir b... Home - SRA - NCBI Princeton University High Throughpu... Nephromine Main shortReadArchive

www.nephromine.org/resource/main.html#ac%3A1N7515%2C1N9853%2C1N5129%3Bdso%3AdatasetName%3Bec%3A[1N2)%3Bpv%3A1N1.1N%2C1N41%3Bet%3Anone%3Bpg%3A1%3Bpvf%3A1N800044980%2C1N800091432? C

Welcome, Felix Eichinger. tools help logout

NEPHROMINE

search

filter selected 4 datasets (835 samples)  
Dataset Type: Normal Tissue Panel  
Dataset Type: Podocyte

Primary Filters

- Analysis Type
  - Coexpression Analysis (4)
  - Differential Analysis (4)
    - + Demographics Analysis (1)
    - Group Analysis (1)
    - Tissue Type Analysis (3)
    - Outlier Analysis (4)
- + Group
- + Tissue Type
- Dataset Type
  - Aging (2)
  - Diabetes (2)
  - Diabetes Mouse (1)
  - FGS (1)
  - Hypertension (1)
  - IgAN (1)
  - Lupus (2)
  - Lupus Mouse (1)
  - Normal Tissue Panel (3)
  - Podocyte (1)
  - Transplant (5)

Sample Filters

Dataset Filters

Concept Filters

datasets concepts

visualize

Analysis Comparison

OTHER VIEWS: ▶

EXPORT: Over-expression

Compare | Clear All

Higgins Normal Tissue Panel (34)

Tissue Type: Glomeruli vs. All Others  
Tissue Type: Papillary Tips vs. All Others  
Tissue Type: Renal Cortex vs. All Others  
Tissue Type: Renal Medulla vs. All Others  
Tissue Type: Renal Pelvis vs. All Others  
Outlier 5th%  
Outlier 10th%  
Outlier 25th%  
Outlier 75th%  
Outlier 90th%  
Outlier 95th%

Ju Podocyte (436)

Glomeruli: Arterial Hypertension vs. Healthy Living Donor  
Glomeruli: Diabetic Nephropathy vs. Healthy Living Donor  
Glomeruli: Focal Segmental Glomerulosclerosis vs. Healthy Living Donor  
Glomeruli: IgA Nephropathy vs. Healthy Living Donor  
Glomeruli: Lupus Nephritis vs. Healthy Living Donor  
Glomeruli: Membranous Glomerulonephritis vs. Healthy Living Donor  
Glomeruli: Minimal Change Disease vs. Healthy Living Donor

Comparison of All Genes Across 3 Analyses

Over-expression

Median Rank	p-Value	Gene
108.5	3.13E-8	KCNH3
124.0	1.89E-6	GPR17
133.0	6.86E-30	MEG3
177.0	1.22E-8	CDH19
245.0	1.18E-22	GNAL
245.0	2.76E-8	NRNN1
276.5	6.55E-10	TCERG1L
294.0	1.74E-7	CD99L2
295.0	4.21E-8	RRBP1
295.0	1.12E-4	FKBP15
298.0	1.14E-4	EZR
298.5	1.20E-7	SPOCK3
311.0	4.73E-8	QKI
328.0	1.65E-4	P2RX7
329.0	4.11E-20	STXBP1
342.5	1.31E-8	UBASH3B
361.0	2.57E-4	HTR2A
366.0	2.93E-4	GLE1
370.0	7.78E-8	NEB
372.0	3.17E-4	SEZ6L2

RRBP1 Rank: 295  
p-Value: 4.21E-8

Legend

1. Tissue Type: Glomeruli vs. All Others Higgins Normal Tissue Panel, *Mol Biol Cell*, 2004

2. Tissue Type: Glomeruli vs. Tubulointerstitium *Lindenmeyer Normal Tissue Panel, PLoS One*, 2010

3. Tissue Type: Brain vs. Kidney Roth Normal Tissue Panel, *Neurogenetics*, 2006

1 5 10 25 25 10 5 1 Not measured

The rank for a gene is the median rank for that gene across each of the analyses.  
The p-value for a gene is its p-value for the median-ranked analysis.

# Concepts Analysis

**Concepts** are sets of genes representing some aspect of biology.

Concepts are derived from both **Nephromine gene expression signatures** as well as **third-party sources** such as Gene Ontology, KEGG Pathways, Human Protein Reference Database, etc.

User can upload a self-defined custom concept (a set of genes) to Nephromine to explore it's association with Nephromine and third-party concepts.

# Concepts Analysis

Logout  
OTHER VIEWS: ▶  
Upload Custom Concept  
Manage My Concepts  
Change password

## Upload My Concept

Concept Name: Podo-50-symbol

Gene Set (Text File):

Category: HUGO Gene Symbol

Null Set(s):  
Affymetrix Human Genome HT U133 Plus 2.0 PM Array  
Agilent-016436 Human miRNA Microarray 1.0 (CBI v1)  
Agilent-019116 Human miRNA Microarray 2.0 G4470B (CBI v1)  
**All Entrez Gene IDs**  
CodeLink Human Whole Genome Bioarray

Description (Optional):

(Up to 500 characters)

Download list from C-tools  
to the desktop, then upload

The press "validate"

# Concepts Analysis

## Upload

Upload My Concept

Concept Name:

Gene Set (Text File):    
podocyte-50\_gene\_symbol.txt

Category: HUGO Gene Symbol

Null Set(s):

Description (Optional):  
  
(Up to 500 characters)

Concept [Podo-50-symbol] validated successfully.  
50 terms were recognized as distinct HUGO gene symbols and will be uploaded.

Concept (Podo-50-symbol) validated successfully.

Then press “Upload”

# Concepts Analysis

## Upload

Upload My Concept

Concept Name: Podo-50-symbol  
Gene Set (Text File): podocyte-50\_gene\_symbol.txt  
Category: HUGO Gene Symbol  
Null Set(s): All Entrez Gene IDs  
Description (Optional):

Your custom concept [Podo-50-symbol] was successfully uploaded and can now be viewed in My Concepts.  
[Select \[Podo-50-symbol\] as primary concept now.](#)

**Close**

Concept (Podo-50-symbol) was successfully uploaded and can be viewed in My Concepts

Select (Podo-50-symbol) as primary concept now.

# Concepts Summary View

## Nephromine Concept Summary

Associated Concept Summary for "Podo-50-symbol - My Concepts"  
Nephromine Concept Summary

Concept Type by Dataset Type	Demographics				Donor Type	Group	Indices		Pathology					Treatment	Nephromine Clusters
	Age	Body Mass Index	Race/Ethnicity	Sex			Hematuria Index Quartile	Chronicity Index Quartile	BUN	GFR (MDRD)	Glomerulosclerosis	Hemoglobin	Proteinuria		
Aging					1				2					1	2
Diabetes					4										1
FSGS															1
Hypertension															1
IgAN															3
Lupus															1
Normal Tissue Panel															9
Transplant															6
Significant Unique Concepts															1

Red: Over-expression    Blue: Under-expression

Other (Non-Nephromine) Concept Summary

Biological Annotations	Pathway Concepts	Regulatory Concepts	Connectivity Map v2 Drug Signatures	Literature-defined Concepts	Mutation Concepts	My Concepts	shRNA Concepts
7	7		3	17		3	

4 concepts meet your threshold and are associated with the primary concept

# Concepts Analysis

search

filter

**datasets** **associated concepts**

Primary Concept: Podo-50-symbol - My Concepts

ORDER BY: p-Value

THRESHOLD BY:

ODDS RATIO: 2.0 P-VALUE: 1E-4

**visualize**

**Concept Association Results**

**Primary Concept:**

- Name: Podo-50-symbol
- Concept Type: My Concepts
- Size: 50 genes
- Null List: All Entrez Gene IDs (42490)

**Associated Concept:**

- Name: Group: Collapsing Focal Segmental Glomerulosclerosis vs. Normal Kidney - Top 5% Under-expressed (Hodgkin FSGS)
- Concept Type: Nephromine Gene Expression Signatures
- Size: 956 genes
- Null List: Affymetrix Human X3P Array (19139)

**Interaction:**

- p-value: 1.54E-18 Q-value: 1.15E-14 Odds Ratio: 18.0 Size: 24 genes

**Coll FSGS vs. Normal kidney**  
**Nephromine Gene Expression Signatures**  
 $P=1.54E-18, q=1.15E-14, Odds=18$   
Top 5% Under-expressed  
Hodgkin FSGS

+ Pathology Analysis (1)  
Outlier Analysis (1)

+ Group  
+ Dataset Type

Sample Filters  
Dataset Filters  
Concept Filters

Group: Focal Segmental Glomerulosclerosis vs. Normal Kidney  
Nephromine Gene Expression Signatures  
 $p=2.21E-8 \quad q=2.56E-5 \quad Odds=6.1$   
Top 10% Under-expressed  
Hodgkin FSGS

Group: Focal Segmental Glomerulosclerosis vs. Minimal Change Disease and Normal Kidney  
Nephromine Gene Expression Signatures  
 $p=7.15E-5 \quad q=0.032 \quad Odds=3.9$   
Top 10% Under-expressed  
Hodgkin FSGS

TN4 actinin, alpha 4  
AGRN agrin  
CD2AP CD2-associated protein  
CD80 CD80 molecule  
CLIC5 chloride intracellular channel 5  
DAG1 dystroglycan 1 (dystrophin-associated glycoprotein 1)  
EZR ezrin  
FYN FYN oncogene related to SRC, FGR, YES  
LRRC7 leucine rich repeat containing 7  
MAFB v-maf musculoaponeurotic fibrosarcoma oncogene homolog B (avian)  
MAGI2 membrane associated guanylate kinase, WW and PDZ domain containing 2  
MME membrane metallo-endopeptidase  
NES nestin  
NPHS1 nephrosis 1, congenital, Finnish type (nephrin)  
NPHS2 nephrosis 2, idiopathic, steroid-resistant (podocin)  
PLCE1 phospholipase C, epsilon 1  
PODXL podocalyxin-like  
PTPRO protein tyrosine phosphatase, receptor type, O  
SCEL sciellin  
SULF1 sulfatase 1  
SYNPO synaptopodin  
TCF21 transcription factor 21  
TJP1 tight junction protein 1 (zona occludens 1)  
WT1 Wilms tumor 1

# Concepts Analysis

**datasets**      **associated concepts**

ORDER BY: Dataset Name ▾  
ON: ▾

Compare | Clear All

Hodgin FSGS (30)

- Group: Collapsing Focal Segmental Glomerulosclerosis vs. Focal Segmental Glomerulosclerosis
- Group: Collapsing Focal Segmental Glomerulosclerosis vs. Minimal Change Disease and Normal Kidney
- Group: Collapsing Focal Segmental Glomerulosclerosis vs. Normal Kidney
- Group: Focal Segmental Glomerulosclerosis vs. Minimal Change Disease and Normal Kidney
- Group: Focal Segmental Glomerulosclerosis vs. Normal Kidney
- Group: Minimal Change Disease vs. Normal Kidney

**visualize**

Differential Analysis

GROUP BY: Group  
SHOW: Only Samples in Analysis ▾

1 | 2 | 3 ▾

Comparison of Concept: "Podo-50-symbol - My Conc  
Under-expression in Group: Collapsing Focal Segmental Glomerul (log2 median-centered intensity)

Rank	P-value	Fold Change	Gene	Reporter
99	0.001	-2.02	ACTN4	g3157975_3p_at
100	0.001	-3.74	SYNPO	g6005797_3p_at
142	0.002	-1.75	MAGI2	Hs.229355.0.A1_3p_at
156	0.002	-2.07	TJP1	g4507516_3p_at
171	0.002	-2.56	PODXL	g4885556_3p_at
194	0.002	-2.63	CLIC5	g8393146_3p_at
234	0.003	-2.78	NES	g13375818_3p_at
272	0.003	-2.19	SULF1	Hs.70823.0.S3_3p_at
278	0.003	-2.18	NPHS1	207673_3p_at
359	0.004	-1.25	LRRC7	Hs2.157325.1.S1_3p_s_at
385	0.005	-2.67	TCF21	g4507394_3p_at
504	0.006	-2.46	NPHS2	g7657614_3p_at
580	0.008	-1.58	DAG1	g4758115_3p_a_at
590	0.008	-3.31	PLCE1	g7705940_3p_s_at
594	0.008	-1.57	FYN	g181171_3p_a_at
658	0.009	-1.97	WT1	g13386509_3p_a_at
696	0.009	-2.99	EZR	g340216_3p_a_at
707	0.009	-1.13	CD80	Hs.2.838.3.S1_3p_at
717	0.009	-2.39	MAFB	Hs.169487.0.S1_3p_a_at
747	0.010	-2.40	CD2AP	g11321633_3p_at

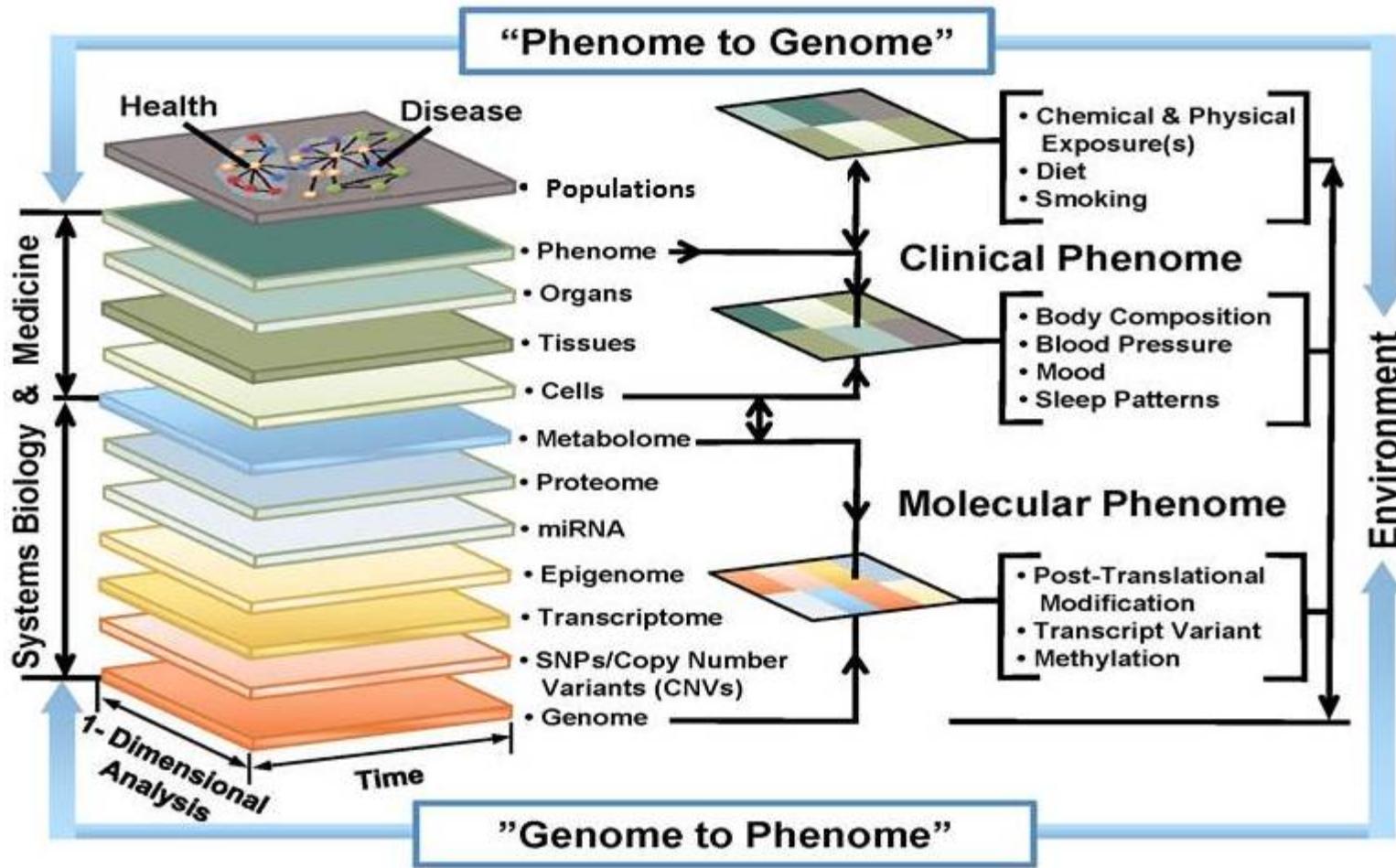
Legend

1. Normal Kidney (9)  
2. Collapsing Focal Segmental Glomerulosclerosis (6)

**PowerPoint**  
**Publication-quality graphic (SVG)**  
**Excel - Analysis Comparison**  
**Excel - Analysis Gene List**  
**Excel - Dataset Detail**

tranSMART

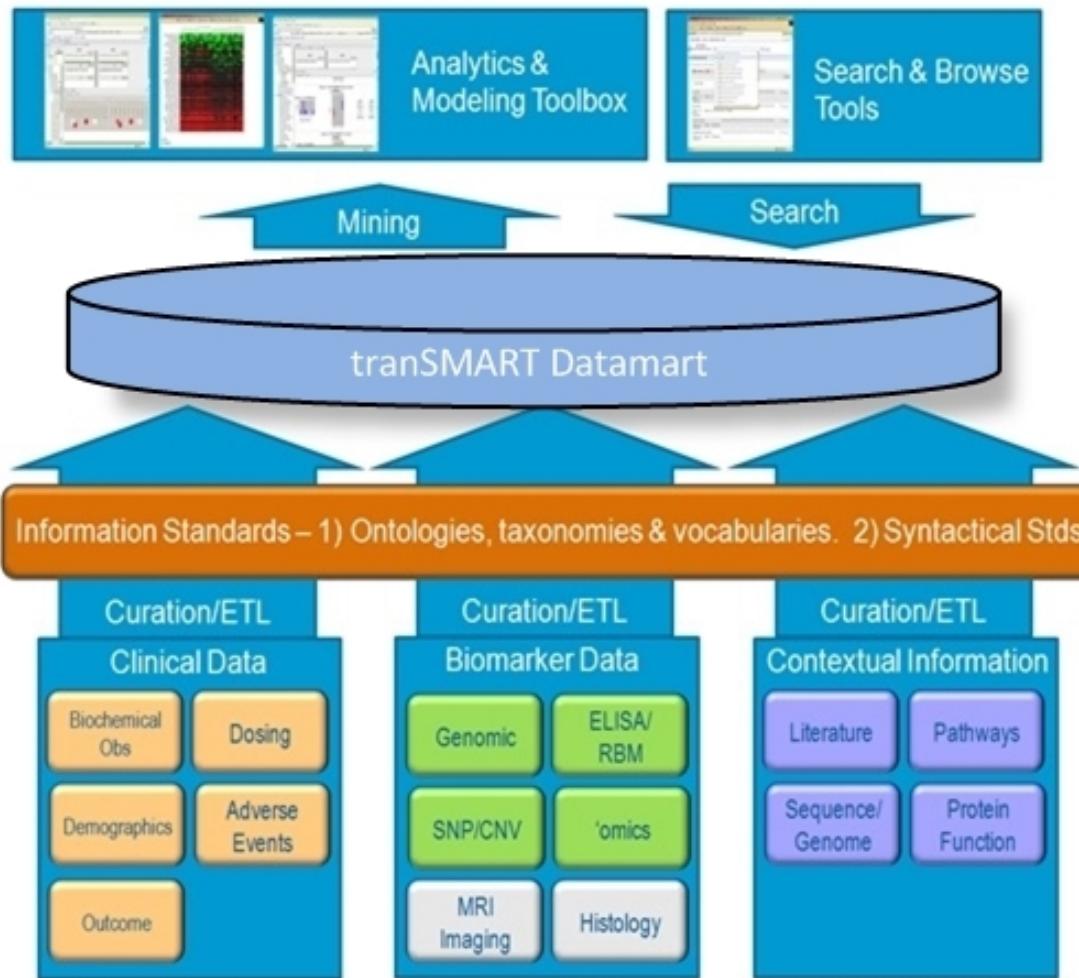
# The Translational Challenge: Data Integration & Analysis



Athey and Omenn, 2009

# tranSMART Platform:

## Enabling Translational research



### **tranSMART – A platform and community**

- Open-source and open-data translational biomedical research community
- Biomedical Researchers, Developers, Service Providers
- Clinician Researchers

# tranSMART Platform: Academics and industry

2009  
Johnson  
and  
Johnson

2010  
Thomson  
Reuters

2012  
One  
Mind for  
Research

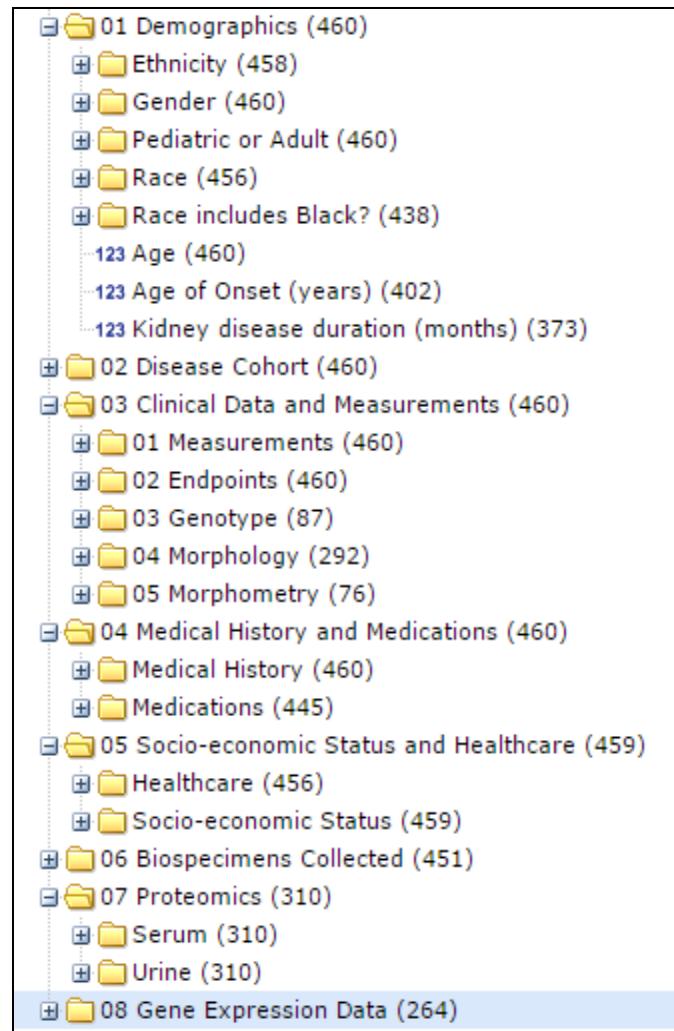
2012 St.  
Jude,  
Harvard,  
Johns  
Hopkins  
Univ.

2010  
Sage  
Bionetw  
orks

2012  
FDA

2012  
Pfizer

# tranSMART: controlled vocabulary



# Subset selection

The screenshot shows the Transmart Dataset Explorer interface at the URL [transmart-nephro.med.umich.edu:7070/transmart/datasetExplorer/index](https://transmart-nephro.med.umich.edu:7070/transmart/datasetExplorer/index). The top navigation bar includes tabs for Search, Dataset Explorer (selected), Gene Signature/Lists, Cross-Database Exploration, and Admin. Below the navigation is a toolbar with buttons for Generate Summary Statistics, Summary, Clear, and Save.

The left sidebar displays a tree view of "Private Studies" under "NeptunePOC2" and "Neptune\_POC2". A green callout box highlights the "eGFR" folder under "Neptune\_POC2" with the text: "Can further specify with AND or exclusion".

The main area contains two subset definitions:

- Subset 1:** Contains the path "...\\FSGS\\" and "...\\eGFR v2\\". These are connected by an "AND" operator. An arrow points from the "eGFR v2" path to the "AND" operator.
- Subset 2:** Contains the path "...\\IMCD\\" and "...\\eGFR v2\\". These are connected by an "AND" operator. An arrow points from the "eGFR v2" path to the "AND" operator.

Each subset entry includes "Exclude" and "X" buttons.

Below the subsets, two boxes labeled "Subset 1" and "Subset 2" have arrows pointing upwards to their respective subset definitions.

# Summary statistics 1

Generate Summary Statistics | Summary | Clear | Save | Export | Print

Comparison Advanced Workflow Results/Analysis Grid View Data Export Export Jobs

### Summary Statistics

**Query Summary for Subset 1**  
(\!Private Studies\Private Studies\Neptune\_POC2\Subjects\Medical History\DX\FSGS1 )  
AND  
(\!Private Studies\Private Studies\Neptune\_POC2\Clinical Measurements\Observations\cGFR\cGFR v21 )

**Query Summary for Subset 2**  
(\!Private Studies\Private Studies\Neptune\_POC2\Subjects\Medical History\DX\MCD )  
AND  
(\!Private Studies\Private Studies\Neptune\_POC2\Clinical Measurements\Observations\cGFR\cGFR v21 )

	Subset 1	Both	Subset 2
11		0	5

**Histogram of Age**

Count

Subset 1: Mean: 43.82, Median: 36, IQR: 41, SD: 21.84, Data Points: 11

Subset 2: Mean: 29.4, Median: 29, IQR: 43, SD: 23.67, Data Points: 5

**Comparison of Age**

Subset 1: Mean: 43.82, Median: 36, IQR: 41, SD: 21.84, Data Points: 11

Subset 2: Mean: 29.4, Median: 29, IQR: 43, SD: 23.67, Data Points: 5

**Sex**

Category	Subset 1 (n)	Subset 1 (%)
F	0	0%
FEMALE	3	27.3%
M	0	0%
MALE	8	72.7%
UNKNOWN	0	0%
null	0	0%
Total	11	100%

**Race**

Category	Subset 1 (n)	Subset 1 (%)
F	0	0%
FEMALE	3	27.3%
M	0	0%
MALE	8	72.7%
UNKNOWN	0	0%
null	0	0%
Total	11	100%

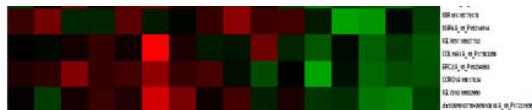
**Sex**

Category	Subset 2 (n)	Subset 2 (%)
F	0	0%
FEMALE	3	60%
M	0	0%
MALE	2	40%
UNKNOWN	0	0%
null	0	0%
Total	5	100%

**Race**

Category	Subset 2 (n)	Subset 2 (%)
F	0	0%
FEMALE	3	60%
M	0	0%
MALE	2	40%
UNKNOWN	0	0%
null	0	0%
Total	5	100%

# Differentially expressed genes



Gene symbols

P-values

Fold change

Table of top Markers

Gene Symbol	Probe ID	Raw p-value	Bonferroni	Holm	Hochberg	SidakS*	SidakS#	BH	BY	t	t (permutation)	Raw P (permutation)	Adjusted P (permutation)	Rank	S1 Mean	S2 Mean	S1 SD	S2 SD	Fold Change
CENPE	A_16_P16795940	0.00000	0.00037	0.00037	0.00037	0.00037	0.00037	0.00018	0.00209	-5.789798	-5.789798	0.0004578755	0.2426740	1	0.38883375	-0.585704500	0.4883929	0.1823082	-0.66387358
LTF	16952883	0.00000	0.00037	0.00037	0.00037	0.00037	0.00037	0.00018	0.00209	-5.789798	-5.789798	0.0004578755	0.2426740	2	0.38883375	-0.585704500	0.4883929	0.1823082	-0.66387358
CORO1A	16817824	0.00001	0.40718	0.40716	0.40715	0.33447	0.33446	0.09503	1.00000	-4.469327	-4.469327	0.0011446886	0.8759158	3	0.63605159	-0.384358980	0.4674378	0.4016459	-1.65483734
ERC2	A_16_P16234993	0.00001	0.40718	0.40716	0.40715	0.33447	0.33446	0.09503	1.00000	-4.469327	-4.469327	0.0011446886	0.8759158	4	0.63605159	-0.384358980	0.4674378	0.4016459	-1.65483734
chr5:095910759-095910818	A_16_P17221956	0.00001	0.57016	0.57011	0.57010	0.43456	0.43454	0.09503	1.00000	-4.396770	-4.396770	0.0006868132	0.9063645	5	0.72159773	-0.020050546	0.4414995	0.2316556	-35.98893154
IGLV310	16932960	0.00001	0.57016	0.57011	0.57010	0.43456	0.43454	0.09503	1.00000	-4.396770	-4.396770	0.0006868132	0.9063645	6	0.72159773	-0.020050546	0.4414995	0.2316556	-35.98893154
MAPT	A_16_P17737416	0.00002	1.00000	1.00000	1.00000	0.68246	0.68242	0.14339	1.00000	4.242481	4.242481	0.0052655678	0.9553571	7	-0.05458949	0.881133820	0.5509338	0.3244491	-0.06195369
RNSS473	16866280	0.00002	1.00000	1.00000	1.00000	0.68246	0.68242	0.14339	1.00000	4.242481	4.242481	0.0052655678	0.9553571	8	-0.05458949	0.881133820	0.5509338	0.3244491	-0.06195369
AK092155	A_16_P17061075	0.00004	1.00000	1.00000	1.00000	0.90285	0.90282	0.23315	1.00000	4.080551	4.080551	0.0050366300	0.9809982	9	-0.06729611	0.450180580	0.3635915	0.1425495	-0.14948691

Table of top Markers

Enlarged:

Gene Symbol	Probe ID	Raw p-value	Bonferroni
CENPE	A_16_P16795940	0.00000	0.00037
LTF	16952883	0.00000	0.00037
CORO1A	16817824	0.00001	0.40718
ERC2	A_16_P16234993	0.00001	0.40718
chr5:095910759-095910818	A_16_P17221956	0.00001	0.57016
IGLV310	16932960	0.00001	0.57016

# Comparisons can be saved/mailed

The screenshot shows the Transmart dataset explorer interface at the URL [transmart-nephro.med.umich.edu:7070/transmart/datasetExplorer/index](http://transmart-nephro.med.umich.edu:7070/transmart/datasetExplorer/index). The top navigation bar includes tabs for Search, Dataset Explorer (selected), Gene Signature/Lists, Cross-Database Exploration, and Admin. Below the navigation bar is a toolbar with buttons for Generate Summary Statistics, Summary, Clear, and Save.

The left sidebar displays a tree view of study datasets:

- Private Studies
  - NeptunePOC2 (55)
    - Biomarker Data (55)
      - ST2 1 (55)
        - kidney tub (55)
    - Clinical Measurements (55)
      - Endpoints (52)
        - 123 evercr (52)
        - 123 evercrpr (52)
      - Observations (55)
        - eGFR (51)
          - 123 eGFR Slope (43)
          - 123 eGFR v2 (50)
          - 123 eGFR v4 (42)
          - 123 eGFR v5 (43)
          - 123 eGFR v6 (38)
          - 123 eGFR v7 (35)
          - 123 eGFR v8 (20)
          - 123 eGFR v9 (7)
    - Serum Creatinine (53)
      - 123 Screat v2 (53)
      - 123 Screat v4 (42)
      - 123 Screat v5 (43)

# tranSMART – why do we care?

- Enables data exploration with low hurdles
- Integrates many different data types
- Has interfaces to real analysis tools
- Provides a consistent data set
- Can be run locally/ institutional etc
- Can possibly be “shared” across institutions
  - McMurry et al, PLOS one: *Shrine: enabling nationally scalable Multi-site disease studies*
- Go to: <http://transmartfoundation.org/>

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Rachel Dull

# Homework for fun

- Connectivity map
  - Use Diabetes vs. control (tubulointerstitium dataset)
  - Select top 1% overexpressed as primary concept
  - Compare to significantly overlapping concepts with Connectivity map
  - Can you find potential drug candidates? Are there any drugs that work for both glom. and tub?
  - What could be optimized? How will you plan further experiments to test your hypothesis?

