

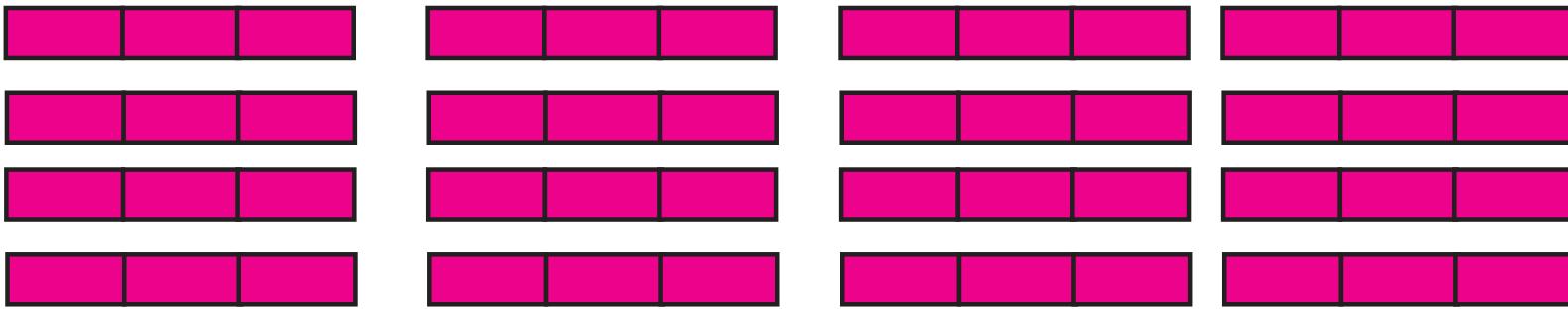
LIBD

# **Reproducible RNA-seq analysis with recount2**

Leonardo Collado-Torres  
@fellgernon  
#bioc2017

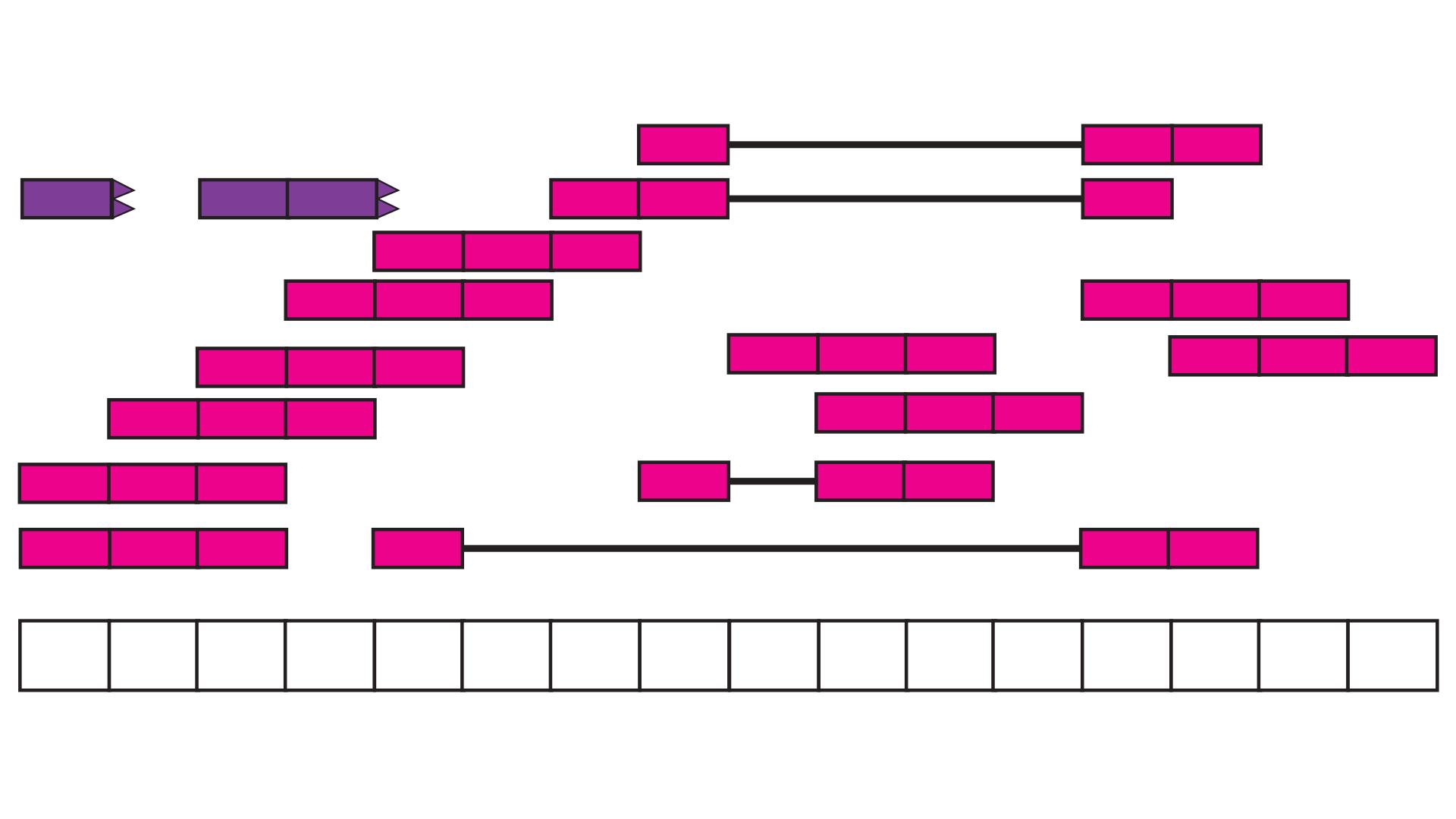
LIEBER INSTITUTE for  
BRAIN DEVELOPMENT  
MALTZ RESEARCH LABORATORIES

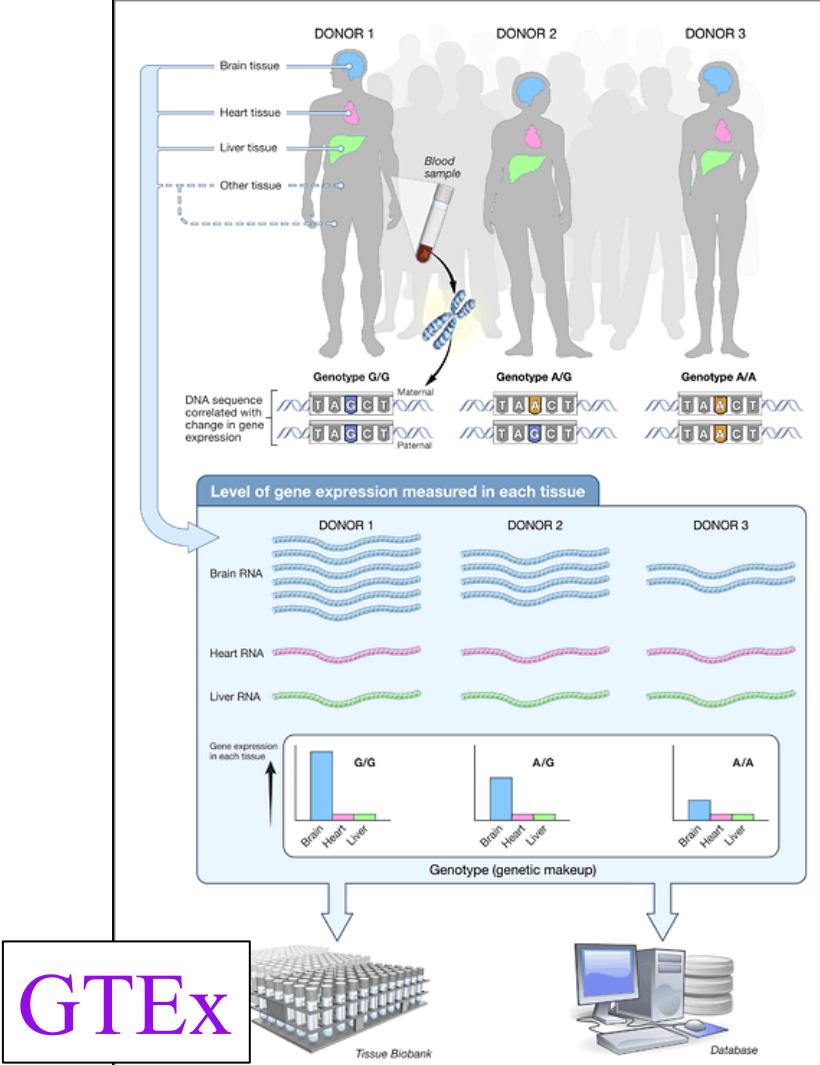
# Reads



# Reference genome







## NATIONAL CANCER INSTITUTE THE CANCER GENOME ATLAS

### TCGA BY THE NUMBERS

TCGA produced over

**2.5 PETABYTES**  
of data

To put this into perspective, 1 petabyte of data is equal to

**212,000 DVDs**



TCGA data describes

**33 DIFFERENT TUMOR TYPES**  
**10 RARE CANCERS**

...based on paired tumor and normal tissue sets collected from

**11,000 PATIENTS**

...using  
**7 DIFFERENT DATA TYPES**



### TCGA RESULTS & FINDINGS

MOLECULAR BASIS OF CANCER Improved our understanding of the genomic underpinnings of cancer

TUMOR SUBTYPES Revolutionized how cancer is classified

THERAPEUTIC TARGETS Identified genomic characteristics of tumors that can be targeted with currently available therapies or used to help with drug development

For example, a TCGA study found the basal-like subtype of breast cancer to be similar to the serous subtype of ovarian cancer on a molecular level, suggesting that despite arising from different tissues in the body, these subtypes may share a common path of development and respond to similar therapeutic strategies.

TCGA revolutionized how cancer is classified by identifying tumor subtypes with distinct sets of genomic alterations.\*

TCGA's identification of targetable genomic alterations in lung squamous cell carcinoma led to NCI's Lung-MAP Trial, which will treat patients based on the specific genomic changes in their tumor.

### THE TEAM

**20 COLLABORATING INSTITUTIONS** across the United States and Canada

### WHAT'S NEXT?

The Genomic Data Commons (GDC) houses TCGA and other NCI-generated data sets for scientists to access from anywhere. The GDC also has many expanded capabilities that will allow researchers to answer more clinically relevant questions with



slide adapted from Shannon Ellis

SRA

SRA

  
Advanced

Search

Help



## SRA

Sequence Read Archive (SRA) makes biological sequence data available to the research community to enhance reproducibility and allow for new discoveries by comparing data sets. The SRA stores raw sequencing data and alignment information from high-throughput sequencing platforms, including Roche 454 GS System®, Illumina Genome Analyzer®, Applied Biosystems SOLiD System®, Helicos Heliscope®, Complete Genomics®, and Pacific Biosciences SMRT®.

### Getting Started

[Understanding and Using SRA](#)[How to Submit](#)[Login to Submit](#)[Download Guide](#)

### Tools and Software

[Download SRA Toolkit](#)[SRA Toolkit Documentation](#)[SRA-BLAST](#)[SRA Run Browser](#)[SRA Run Selector](#)

### Related Resources

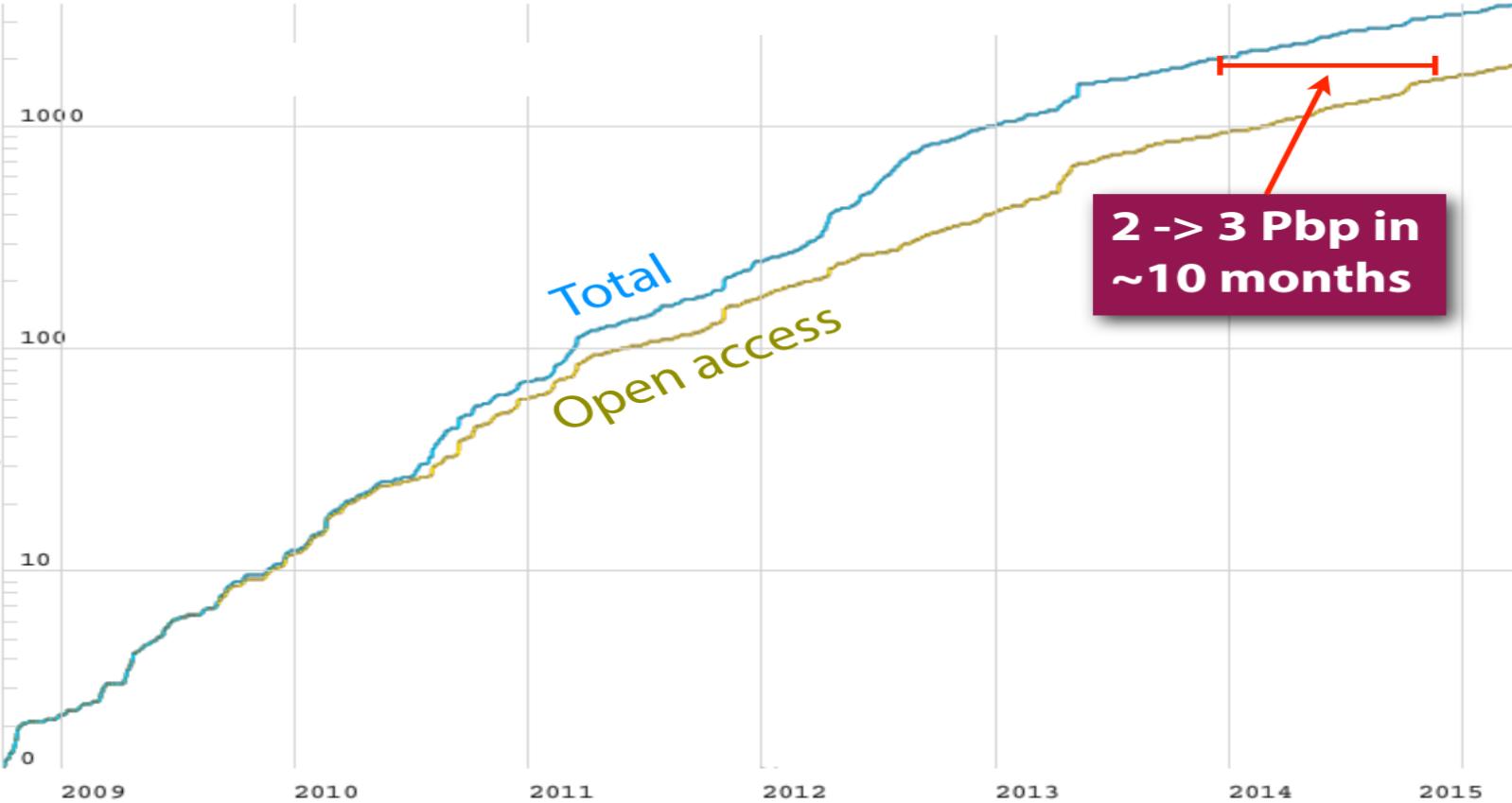
[dbGaP Home](#)[Trace Archive Home](#)[BioSample](#)[GenBank Home](#)

# SRA

# Sequence Read Archive (SRA) growth

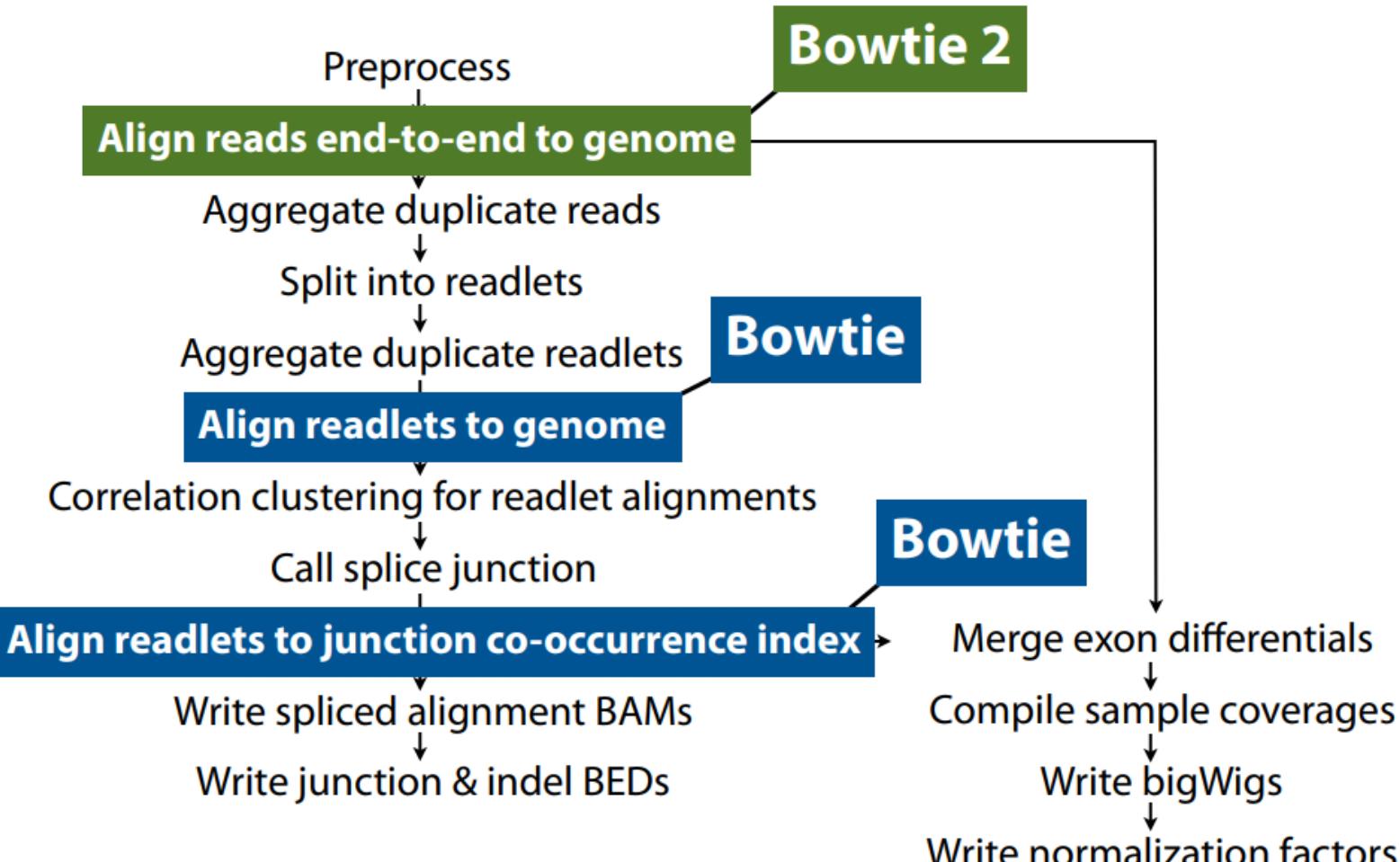
1 Pbp

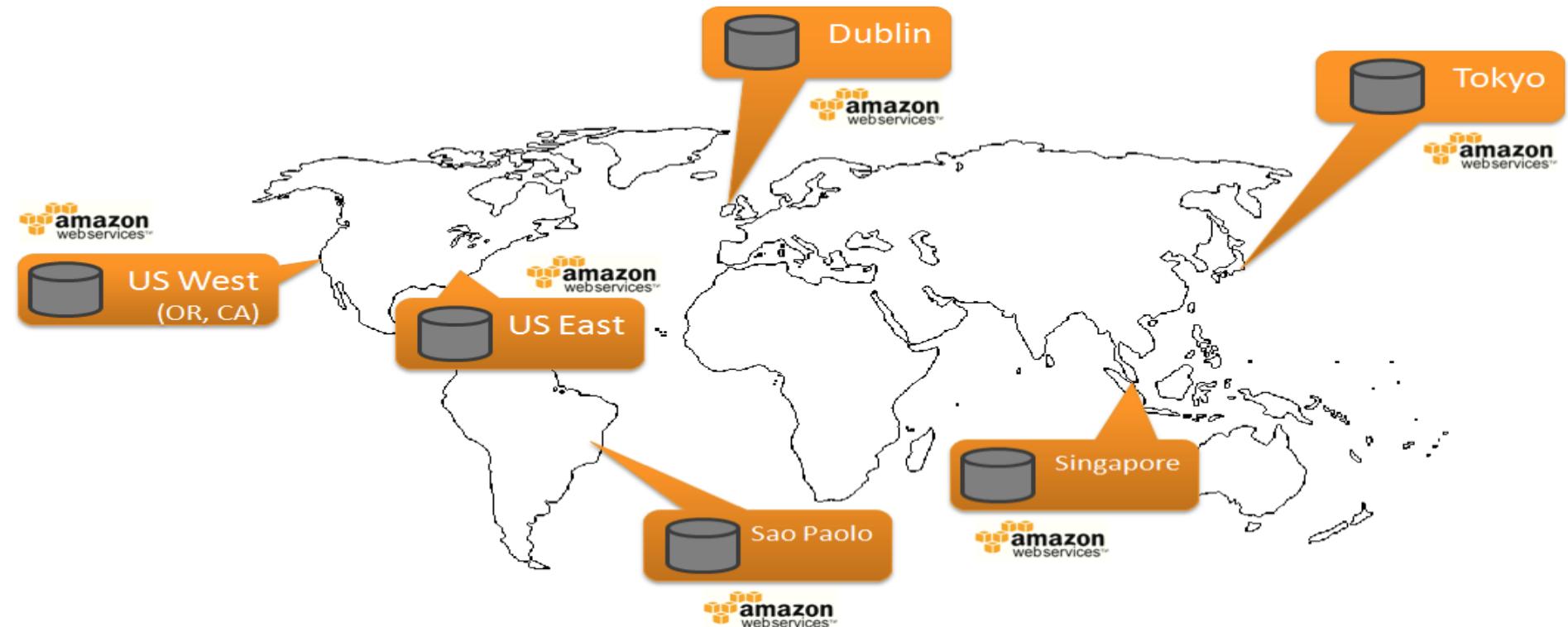
Terabases



03/26/2015 12:07pm

Slide adapted from Ben Langmead







# A multi-experiment resource of analysis-ready RNA-seq gene and exon count datasets

recount2 is an online resource consisting of RNA-seq gene and exon counts as well as coverage bigWig files for 2041 different studies. It is the second generation of the [ReCount project](#). The raw sequencing data were processed with [Rail-RNA](#) as described in the recount2 paper and at [Nellore et al, Genome Biology, 2016](#) which created the coverage bigWig files. For ease of statistical analysis, for each study we created count tables at the gene and exon levels and extracted phenotype data, which we provide in their raw formats as well as in RangedSummarizedExperiment R objects (described in the [SummarizedExperiment](#) Bioconductor package). We also computed the mean coverage per study and provide it in a bigWig file, which can be used with the [derfinder](#) Bioconductor package to perform annotation-agnostic differential expression analysis at the expressed regions-level as described at [Collado-Torres et al, Genome Research, 2017](#). The count tables, RangedSummarizedExperiment objects, phenotype tables, sample bigWigs, mean bigWigs, and file information tables are ready to use and freely available here. We also created the [recount](#) Bioconductor package which allows you to search and download the data for a specific study . By taking care of several preprocessing steps and combining many datasets into one easily-accessible website, we make finding and analyzing RNA-seq data considerably more straightforward.

## Related publications

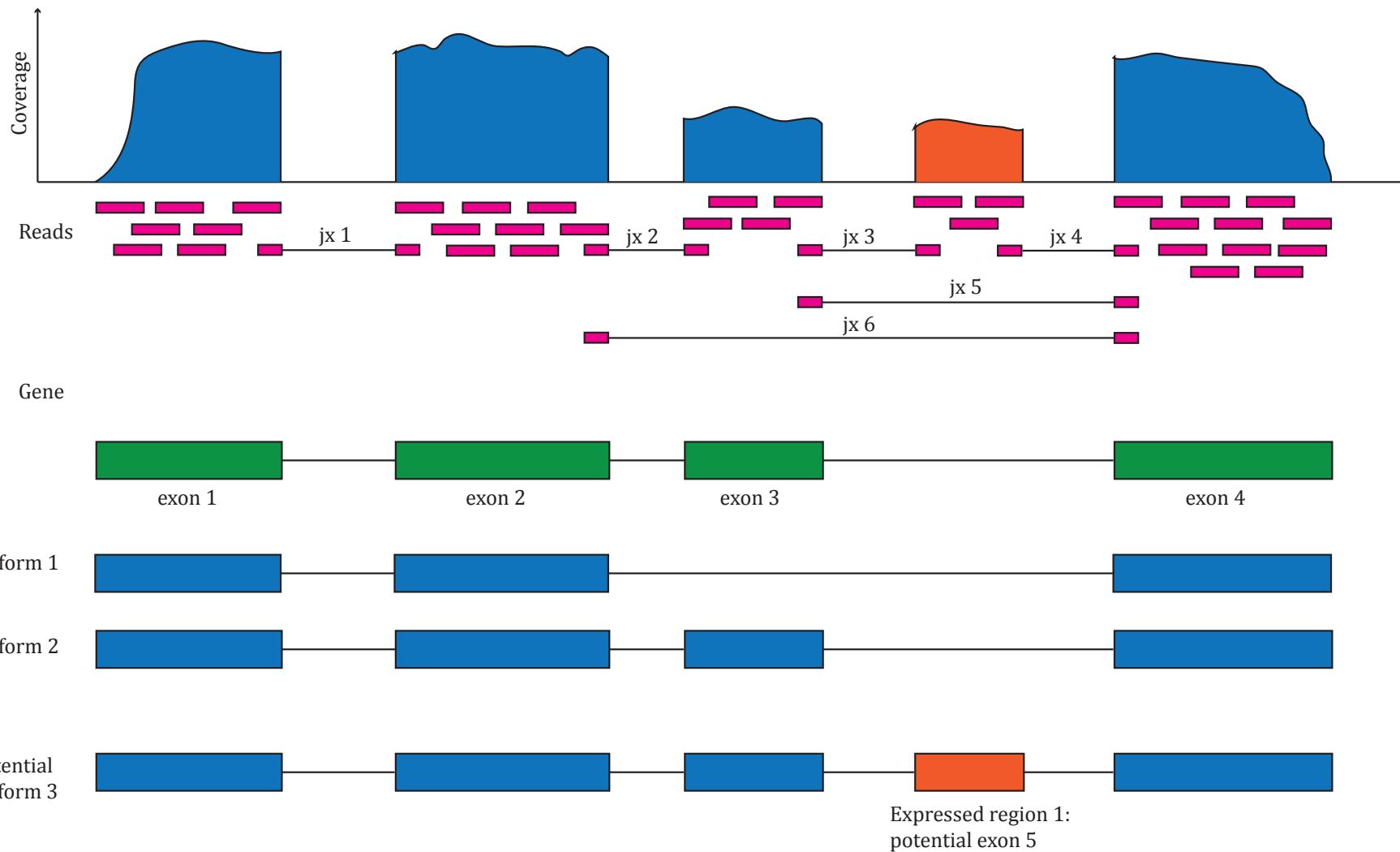
Collado-Torres L, Nellore A, Kammers K, Ellis SE, Taub MA, Hansen KD, Jaffe AE, Langmead B, Leek JT. Reproducible RNA-seq analysis using [recount2](#). *Nature Biotechnology*, 2017. doi: 10.1038/nbt.3838.

## The Datasets

Show  entries

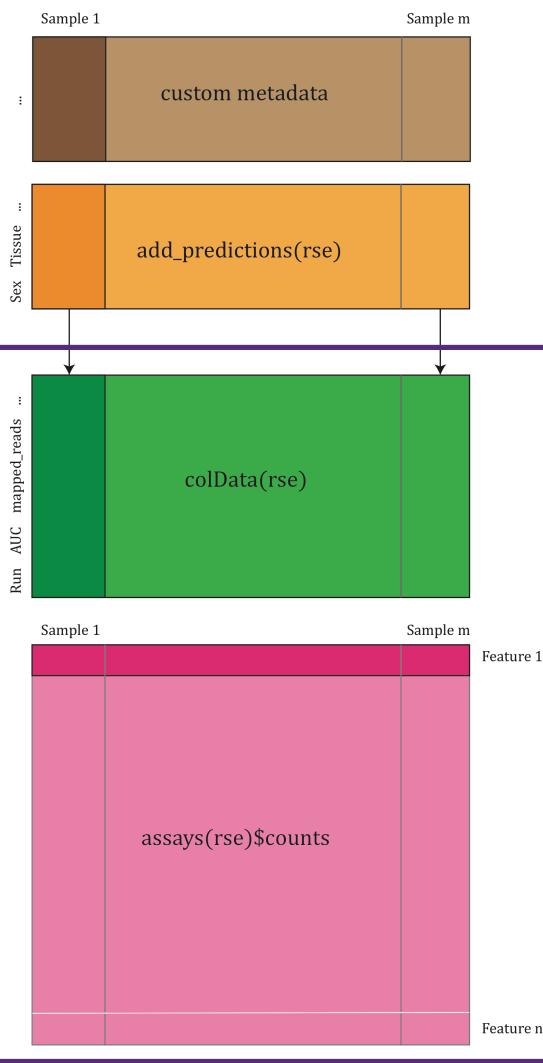
<https://jhubiostatistics.shinyapps.io/recount/>

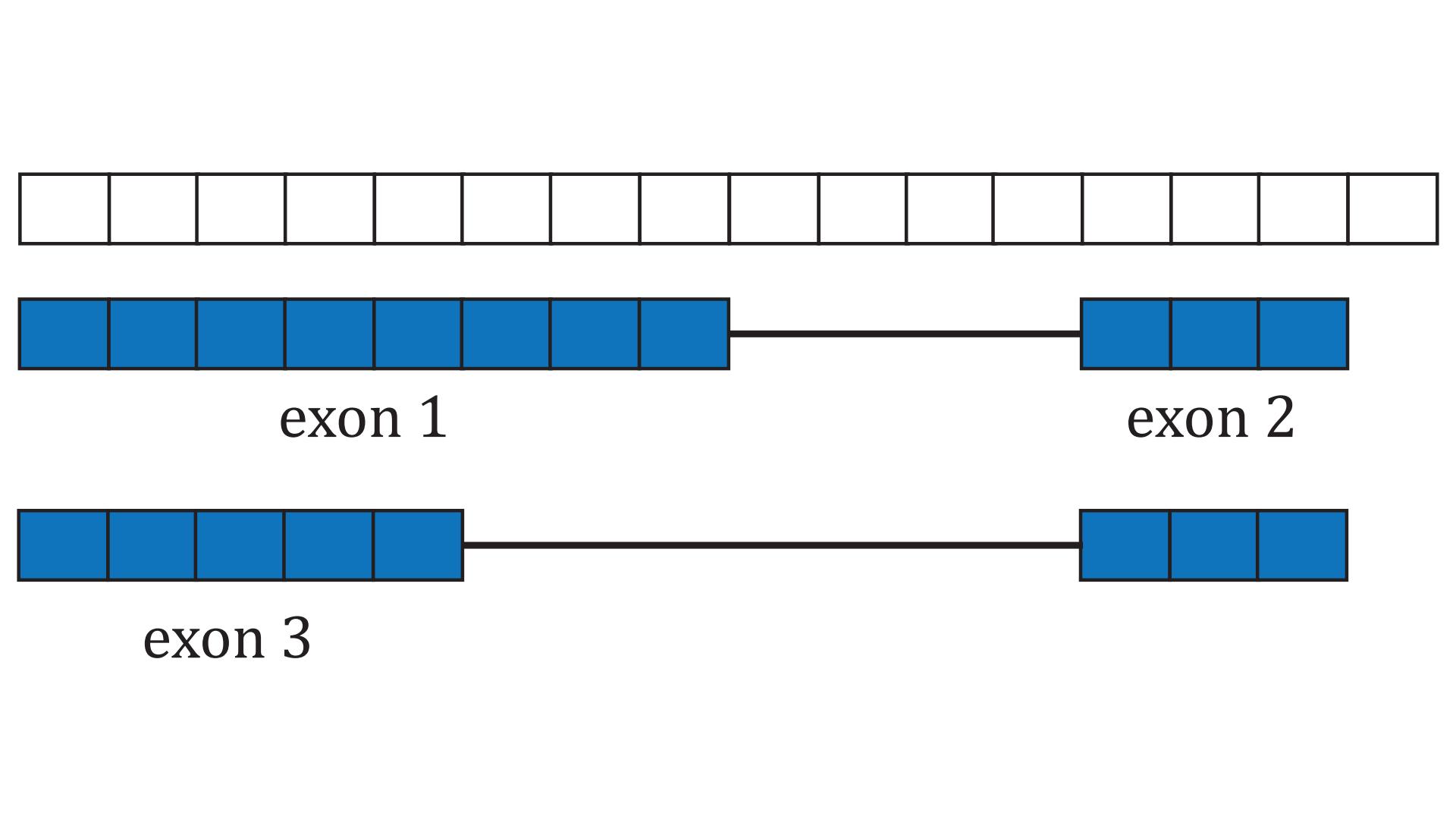
accession	number of samples	species	abstract	gene	exon	junctions	phenotype	files info
All	All	All	<input type="text" value="libd"/> <input type="button" value="⊗"/>	All	All	All	All	All
SRP045638	72	human	RNAseq data of 36 samples across human brain development by age group from LIBD	<a href="#">RSE counts</a>	<a href="#">RSE counts</a>	<a href="#">RSE jx_bed</a> <a href="#">jx_cov counts</a>	<a href="#">link</a>	<a href="#">link</a>





download\_study()  
load()







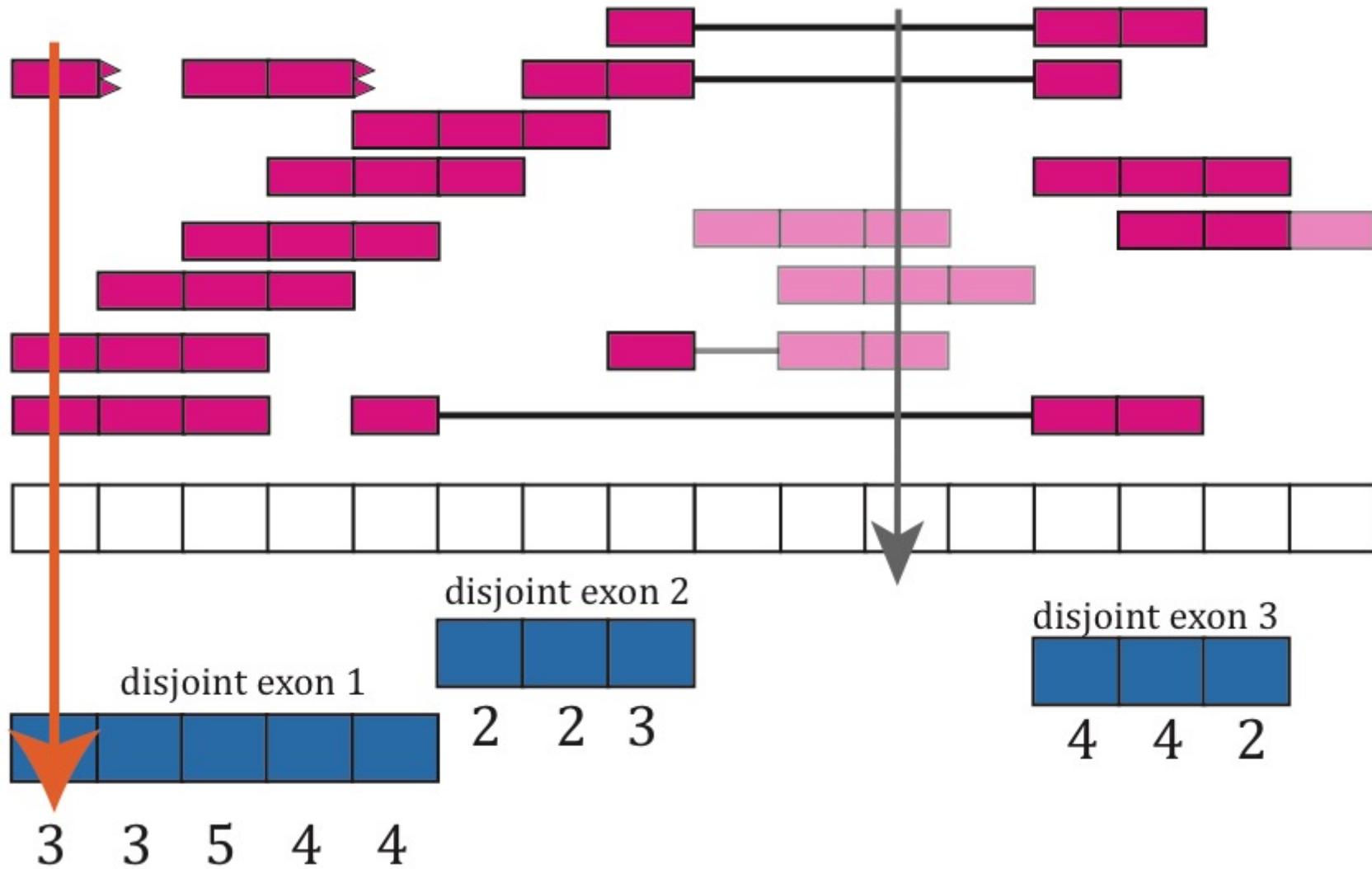
disjoint exon 2

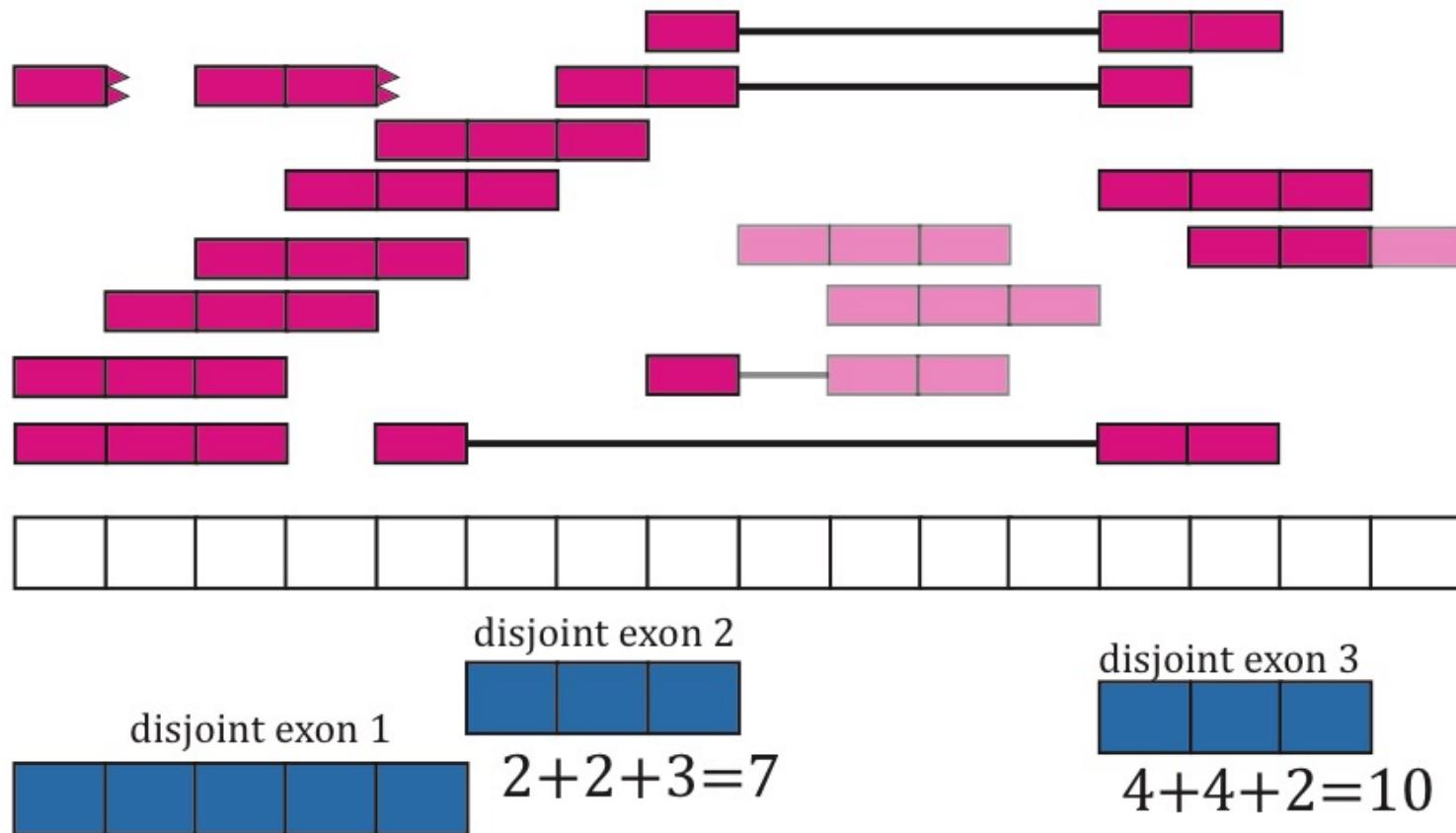


disjoint exon 3



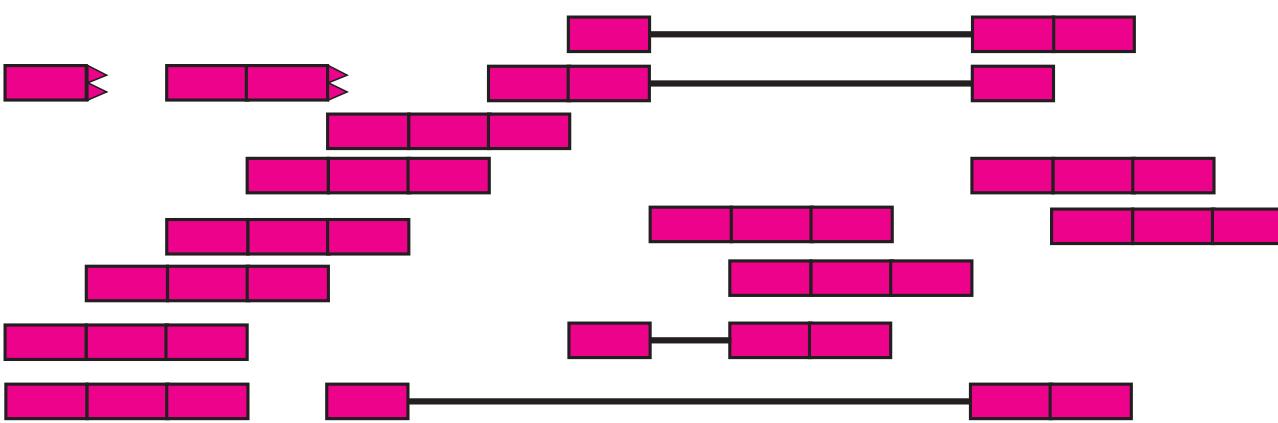
disjoint exon 1



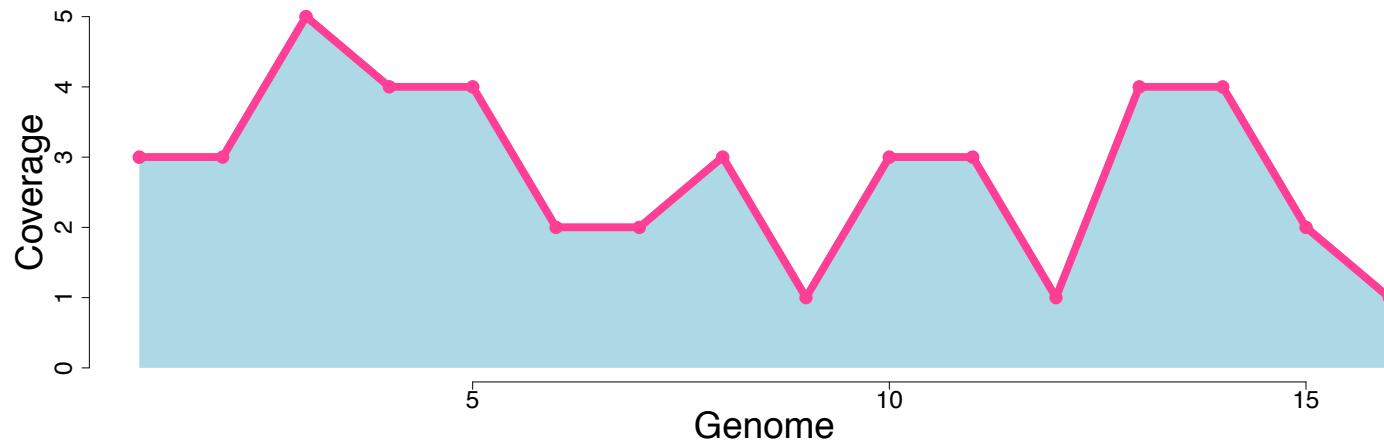


$$\text{Gene} = 19 + 7 + 10 = 36$$

$$\frac{\sum_i^n \text{coverage}_i}{\text{Read Length}} * \frac{\text{target}}{\text{mapped}} = \text{scaled read counts}$$



3	3	5	4	4	2	2	3	1	3	3	1	4	4	2	1
---	---	---	---	---	---	---	---	---	---	---	---	---	---	---	---



AUC = area under coverage = 45

$$\frac{\sum_i^n \text{coverage}_i}{\text{Read Length}} * \frac{\text{target}}{\text{mapped}} = \text{scaled read counts}$$

$$\frac{\sum_i^n \text{coverage}_i}{\text{AUC}} * \text{target} = \text{scaled read counts}$$

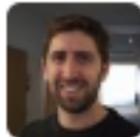
```
> library('recount')

> download_study( 'ERP001942', type='rse-gene')

> load(file.path('ERP001942 ', 'rse_gene.Rdata'))

> rse <- scale_counts(rse_gene)
```

<https://github.com/leekgroup/recount-analyses/>



**Mike Love**

@mikelove

Following

Replying to [@jtleek](#)

Recount has been very useful for me over the years in developing and testing methods

RETWEETS

4

LIKES

5



10:17 AM - 11 Apr 2017



4

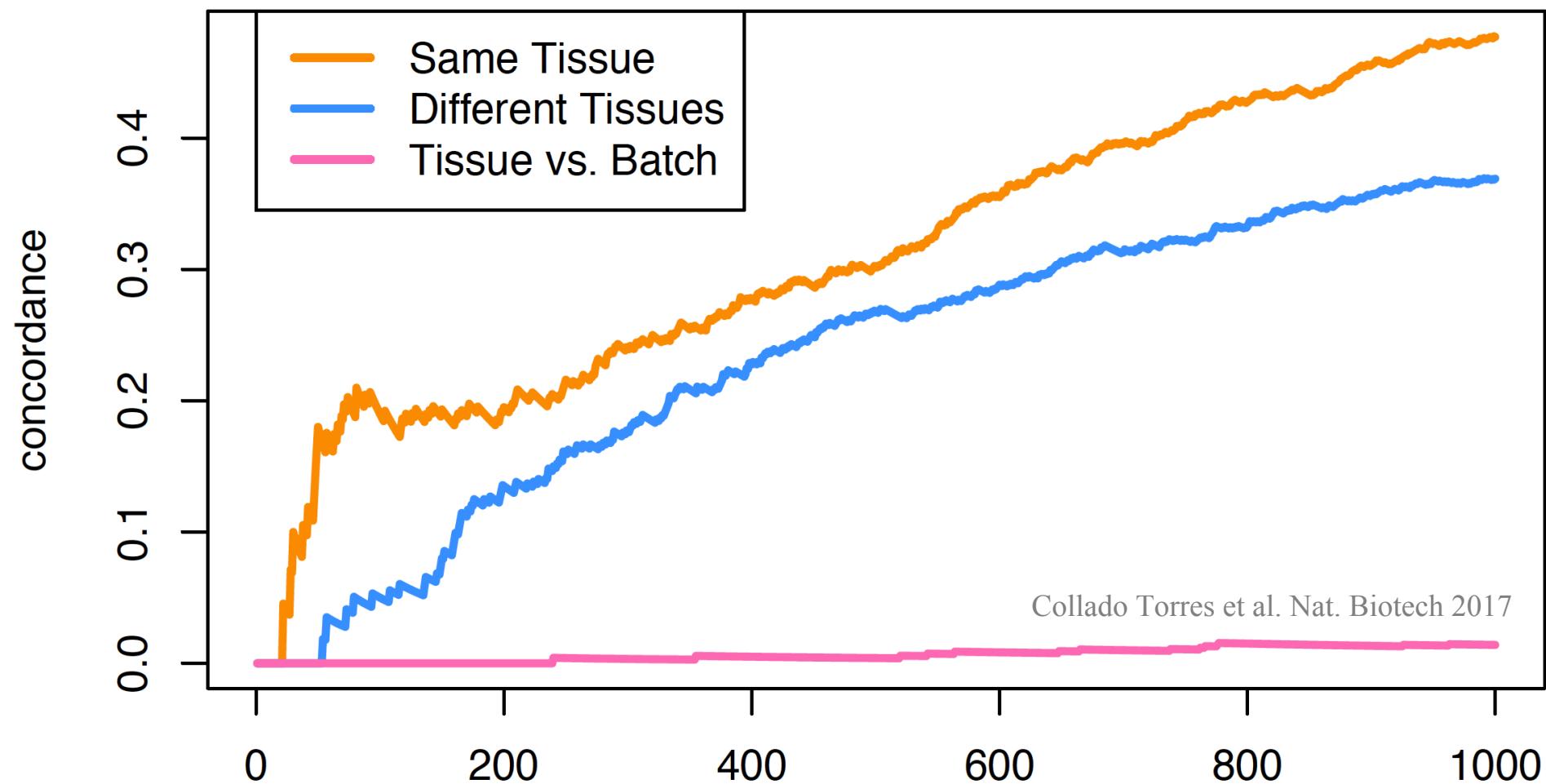


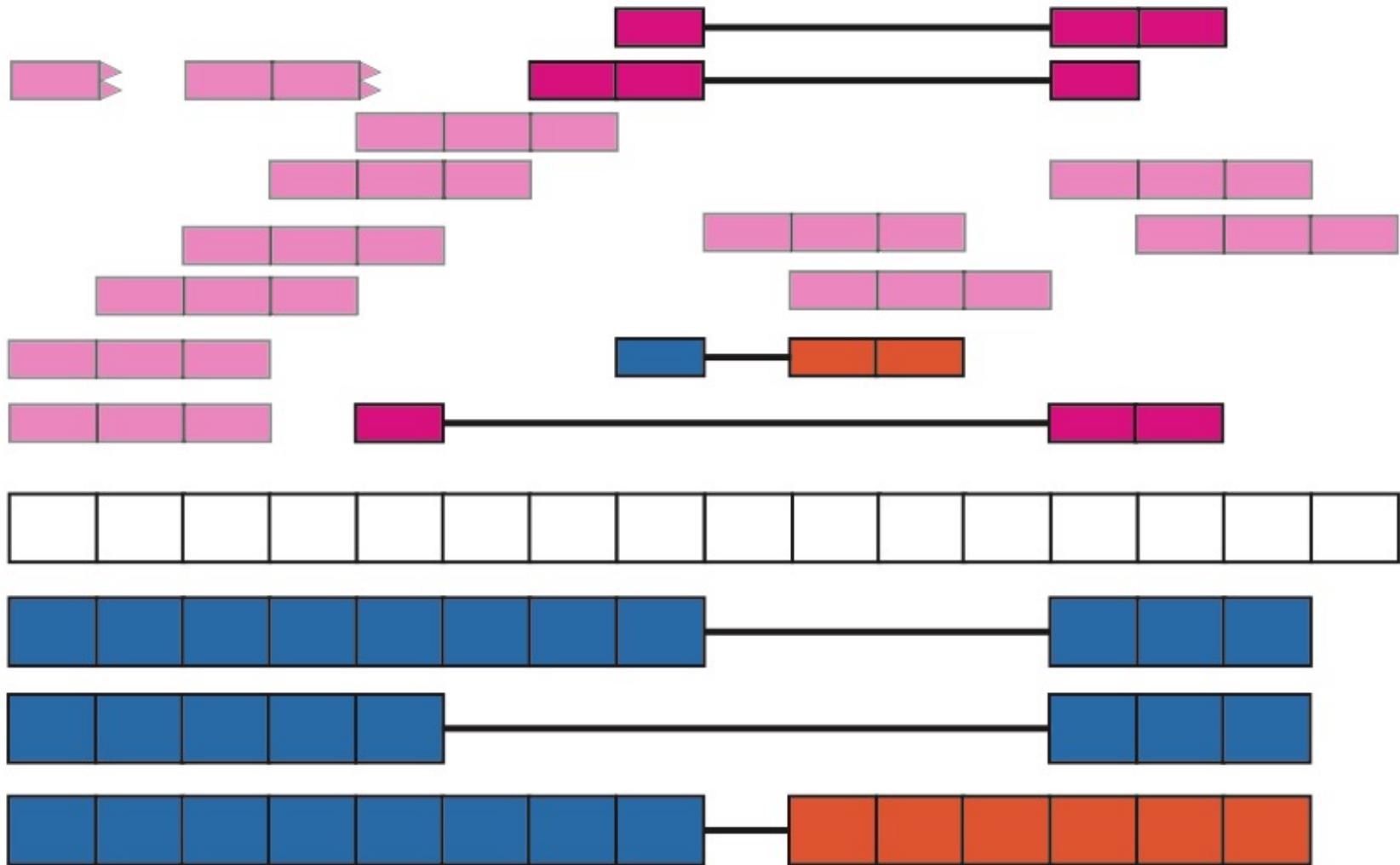
5

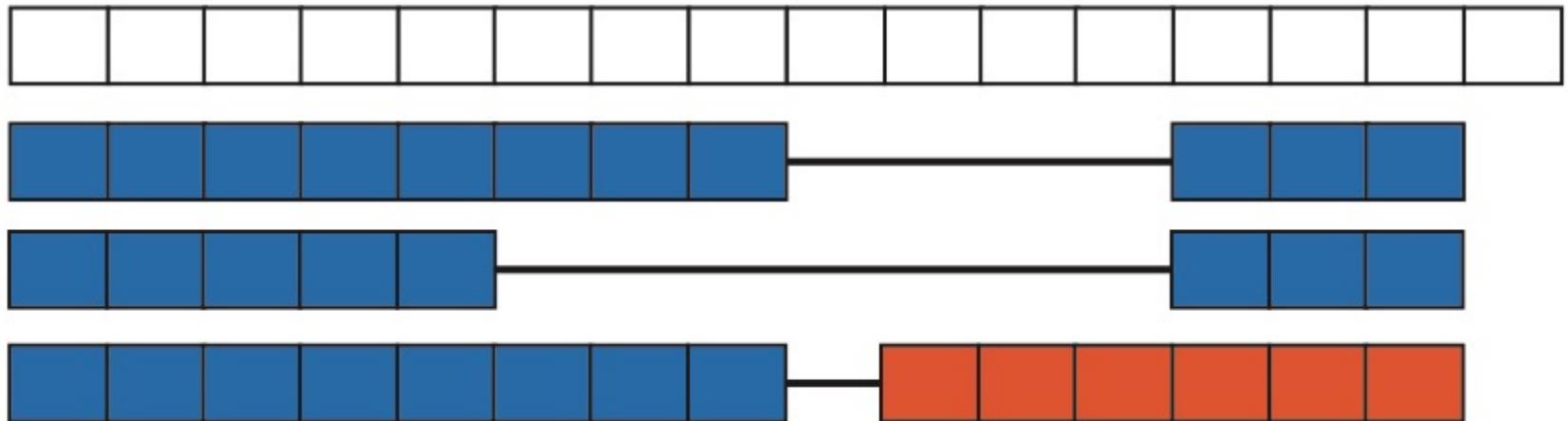
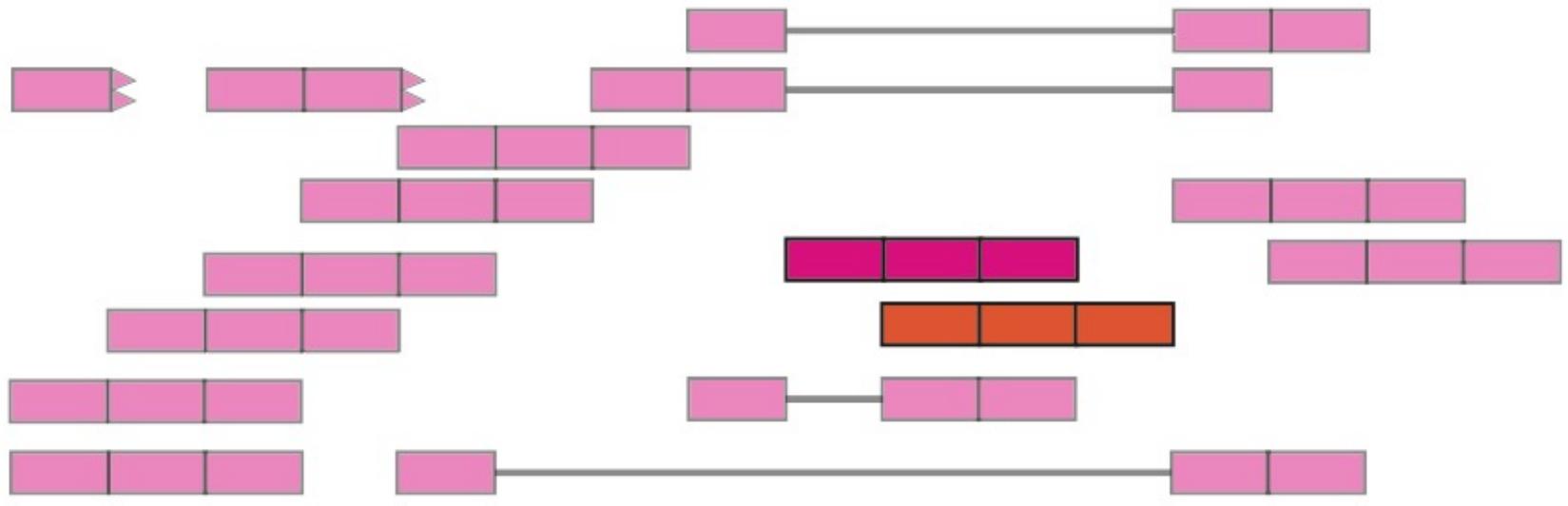
```
> library('recount')

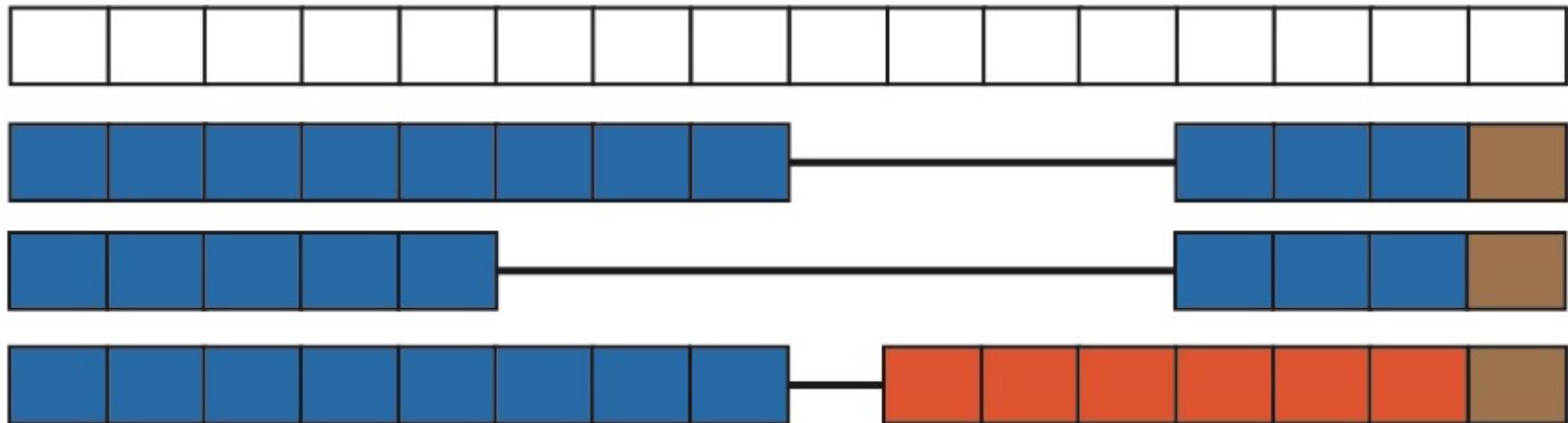
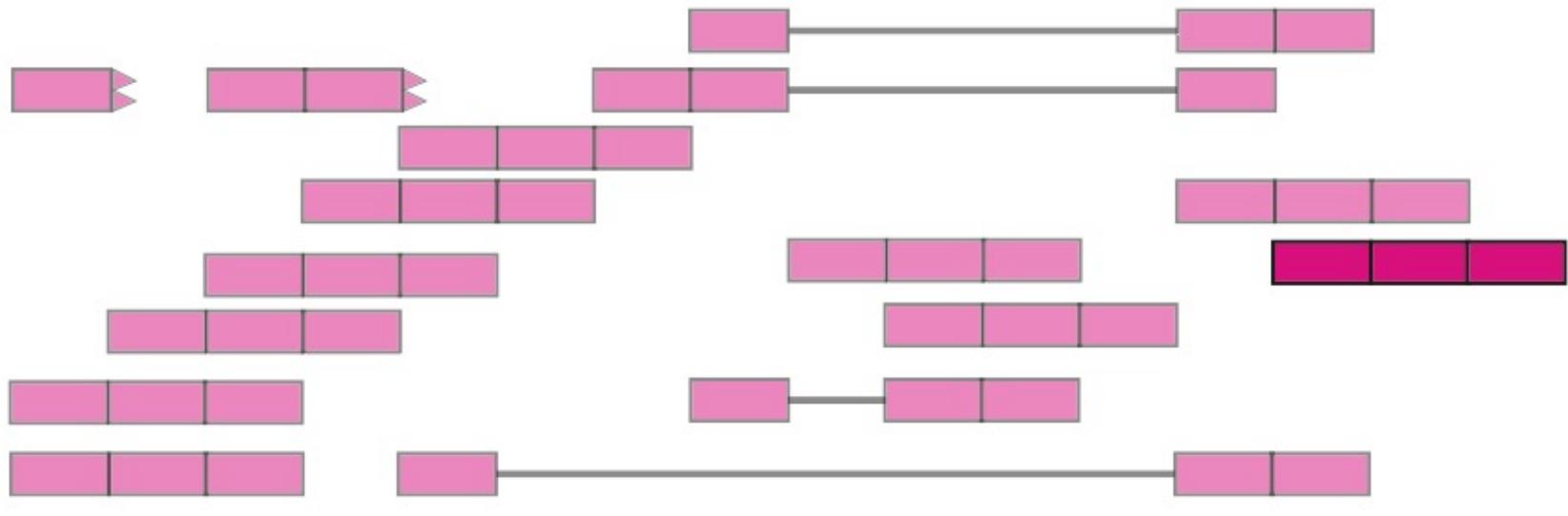
> download_study('SRP029880', type='rse-gene')
> download_study('SRP059039', type='rse-gene')
> load(file.path('SRP029880 ', 'rse_gene.Rdata'))
> load(file.path('SRP059039', 'rse_gene.Rdata'))
> mdat <- do.call(cbind, dat)
```

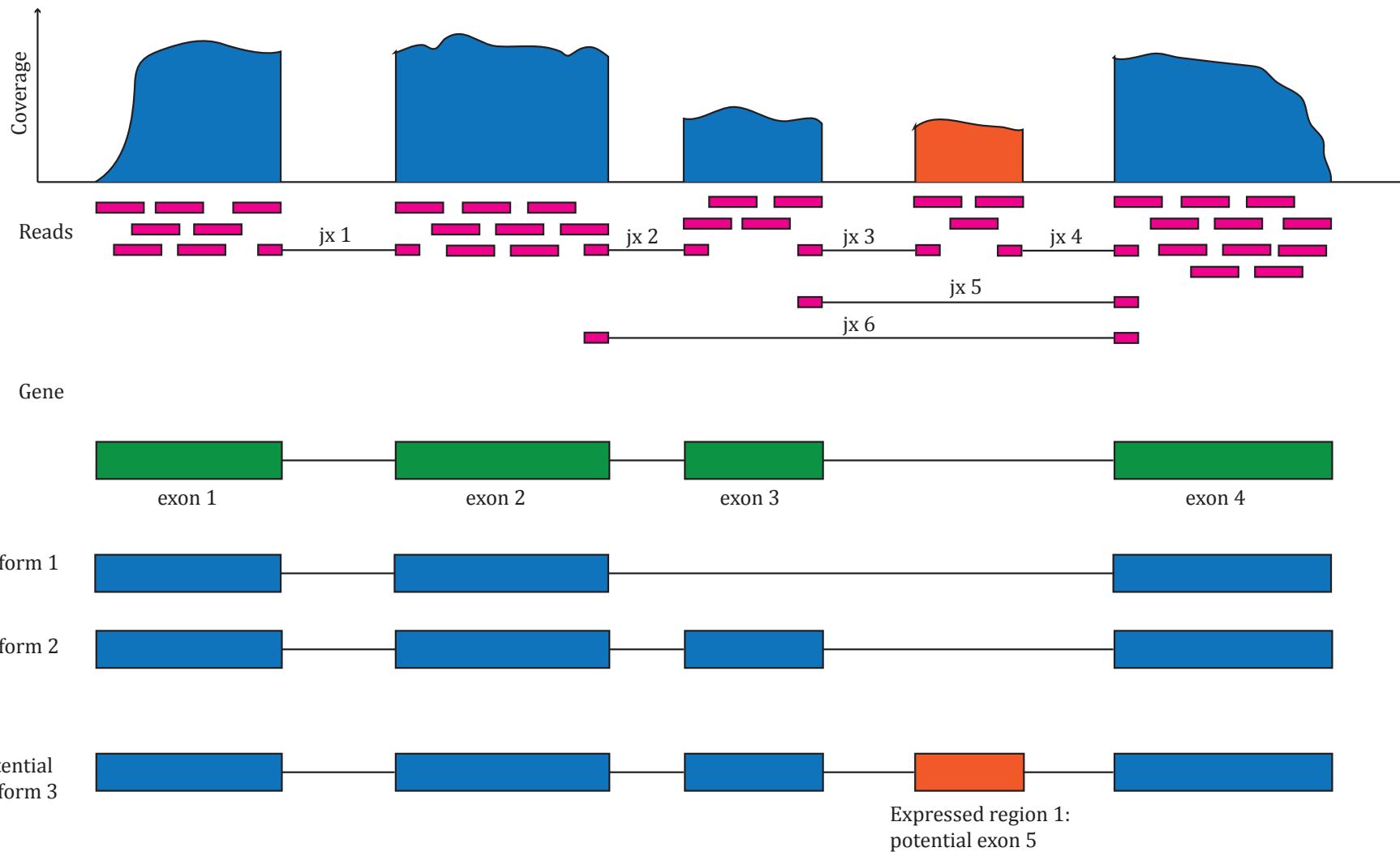
# Average Log2 Fold Change

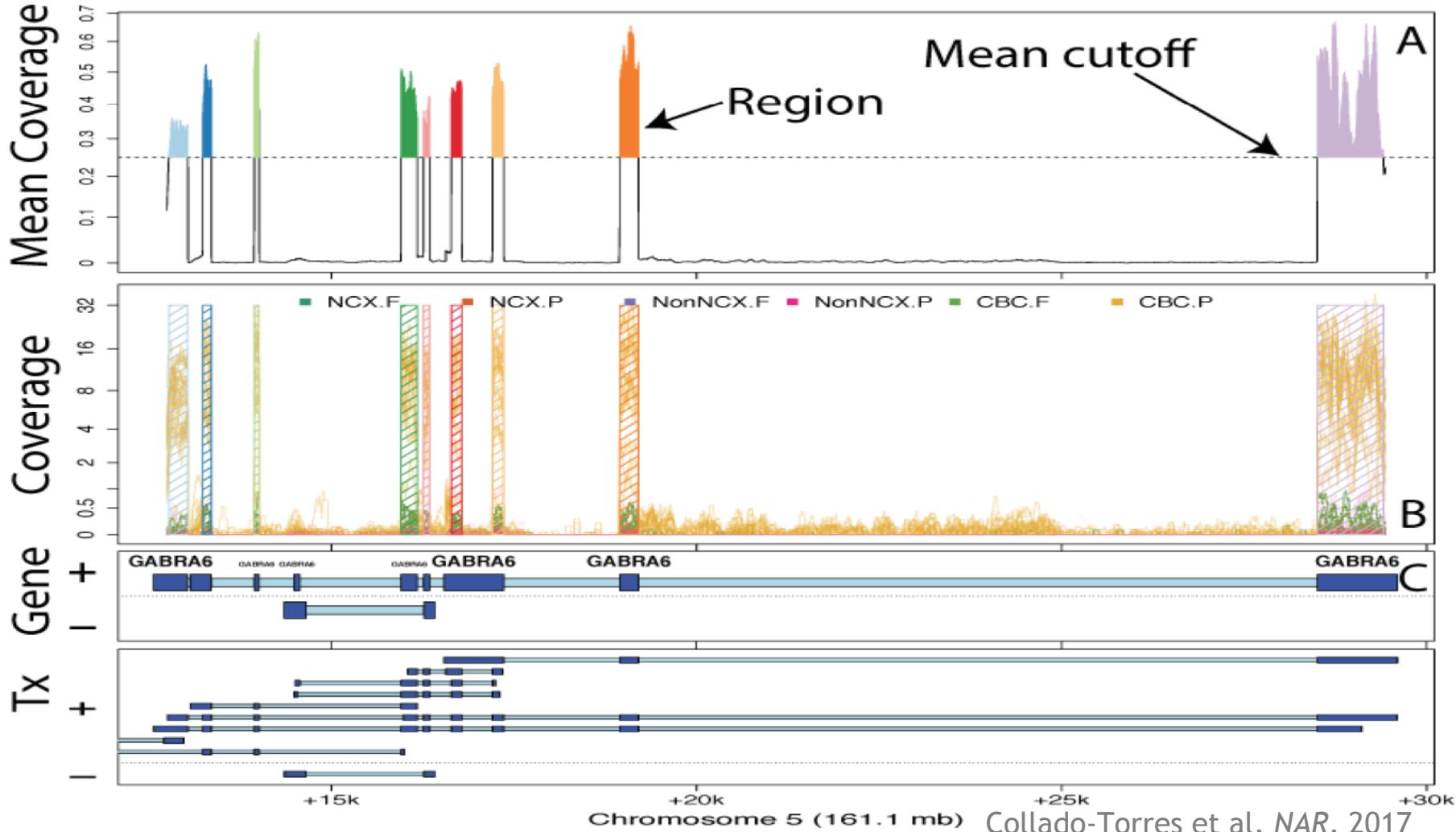












# Postmortem Human Brain Samples

Discovery data



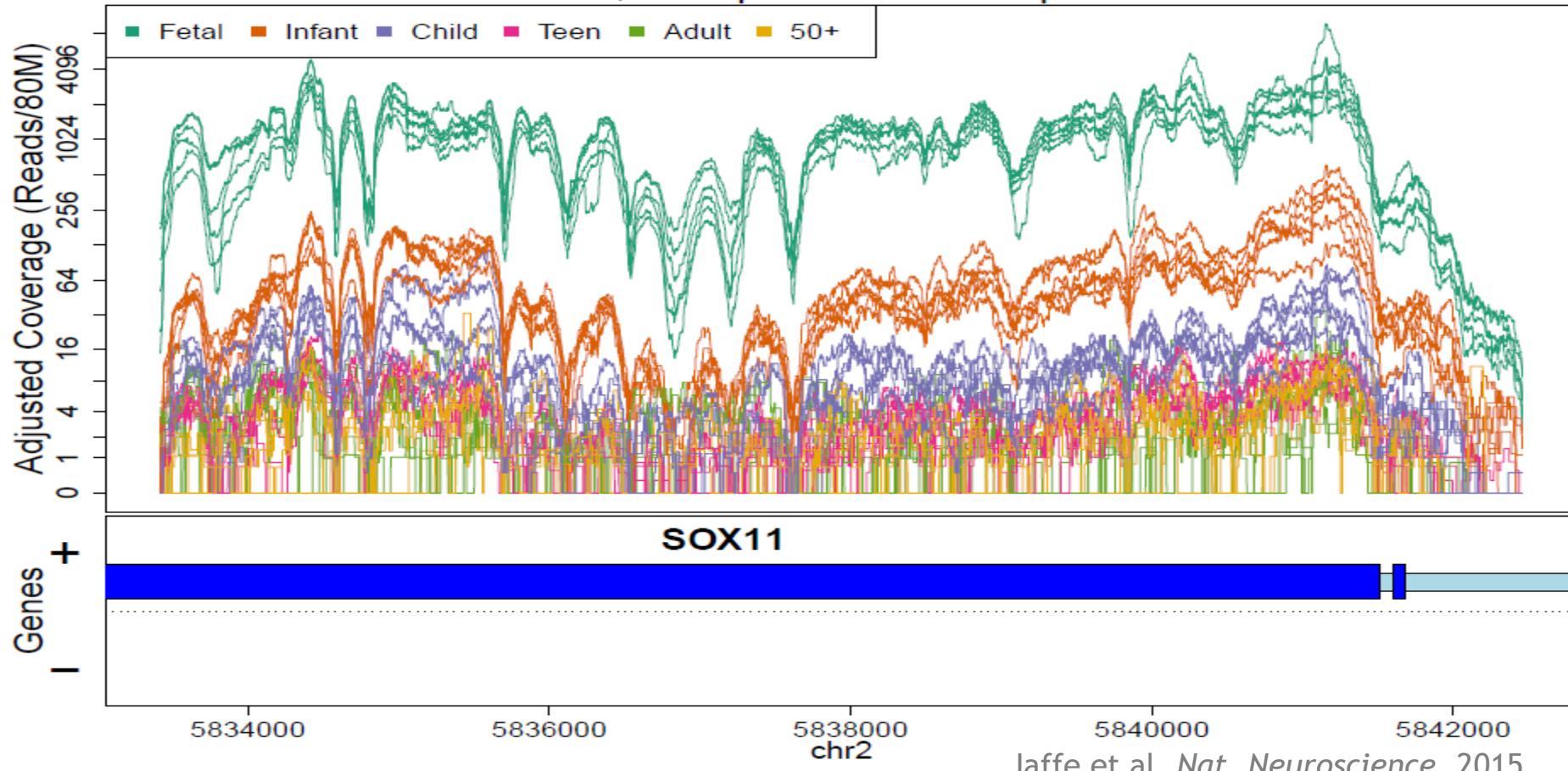
6 / group, N = 36

Replication data

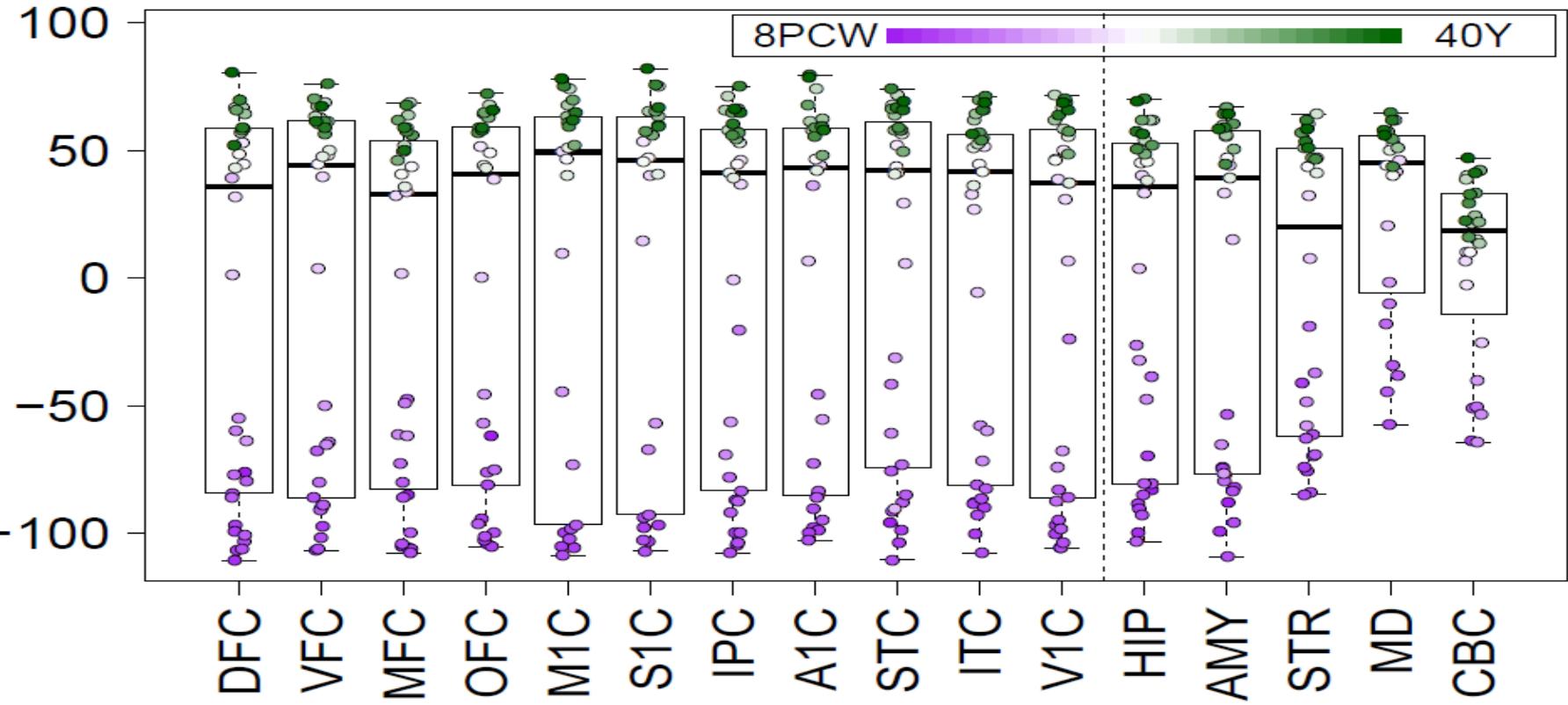


6 / group, N = 36

### SOX11 , 619 bp from tss: overlaps 3'



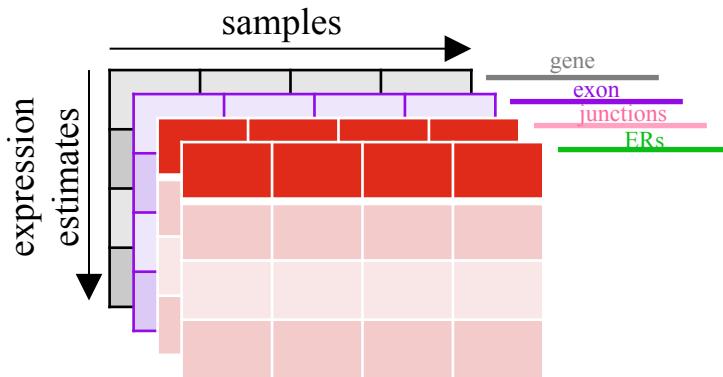
PC1: 59% of Var Explained



BrainSpan data  
Jaffe et al, *Nat. Neuroscience*, 2015

The logo for recount2 features three colored bars at the top: red, green, and blue. To the right of the bars, the word "recount" is written in a lowercase, sans-serif font, with the letter "t" being significantly larger than the others. A large, bold, black number "2" is positioned to the right of "recount".

*expression data for ~70,000 human samples*

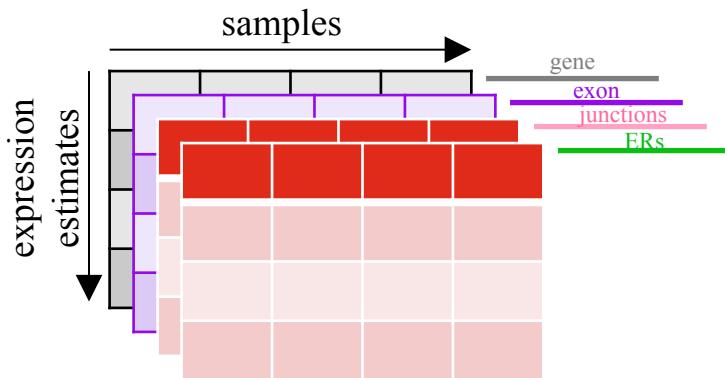


GTEX	SRA	TCGA
N=9,962	N=49,848	N=11,284



# recount2

*expression data for ~70,000 human samples*



GTEX	SRA	TCGA
N=9,962	N=49,848	N=11,284

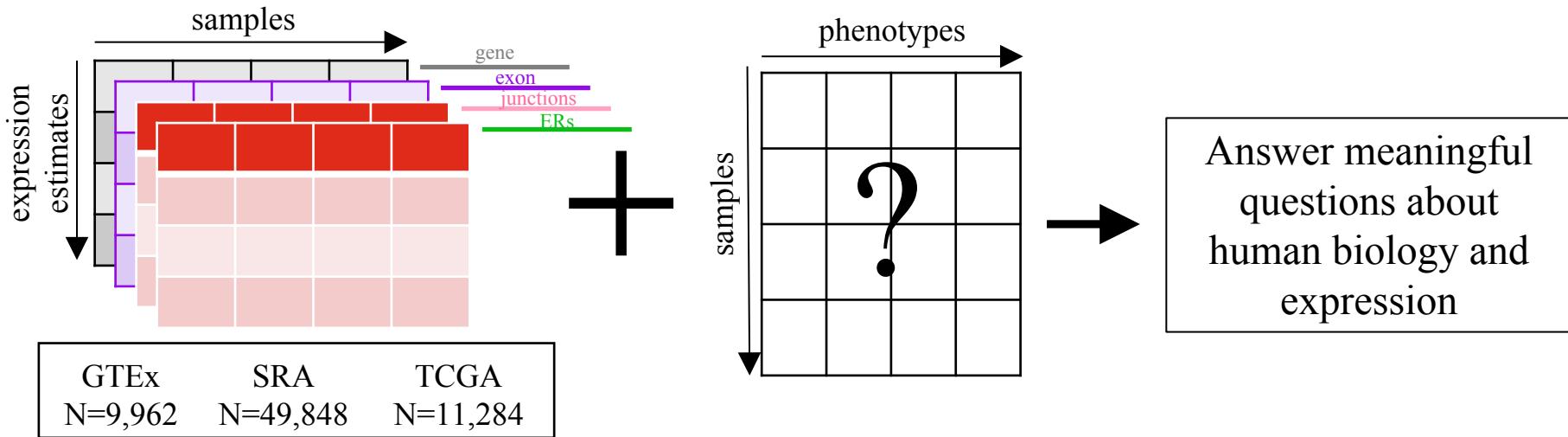


Answer meaningful  
questions about  
human biology and  
expression



# recount2

*expression data for ~70,000 human samples*



Even when information *is* provided, it's not always clear...

**sra\_meta\$Se**

<b>Category</b>	<b>Frequency</b>
F	95
female	2036
Female	51
M	77
male	1240
Male	141
<b>Total</b>	<b>3640</b>

“1 Male, 2 Female”, “2 Male, 1 Female”, “3 Female”, “DK”, “male and female” “Male (note: ....)”, “missing”, “mixed”, “mixture”, “N/A”, “Not available”, “not applicable”, “not collected”, “not determined”, “pooled male and female”, “U”, “unknown”, “Unknown”

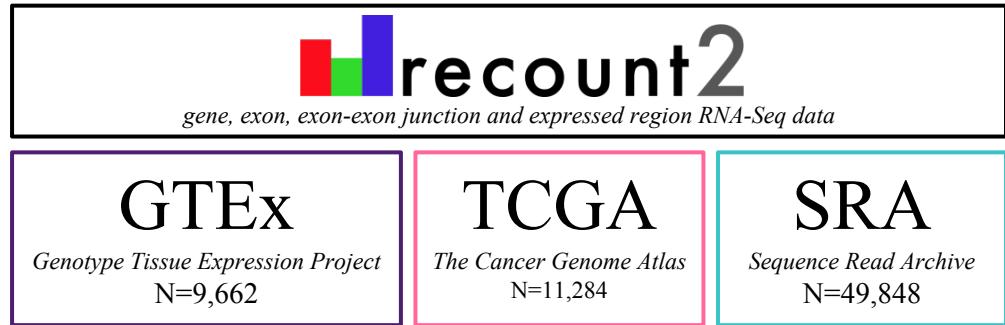
# SRA phenotype information is far from complete

SubjectID	Sex	Tissue	Race	Age
6620	NA	female	liver	NA
6621	NA	female	liver	NA
6622	NA	female	liver	NA
6623	NA	female	liver	NA
6624	NA	female	liver	NA
6625	NA	male	liver	NA
6626	NA	male	liver	NA
6627	NA	male	liver	NA
6628	NA	male	liver	NA
6629	NA	male	liver	NA
6630	male	liver	NA	NA
6631	NA	blood	NA	NA
6632	NA	blood	NA	NA
6633	NA	blood	NA	NA
6634	NA	blood	NA	NA
6635	NA	blood	NA	NA
6636	NA	blood	NA	NA

slide adapted from Shannon Ellis

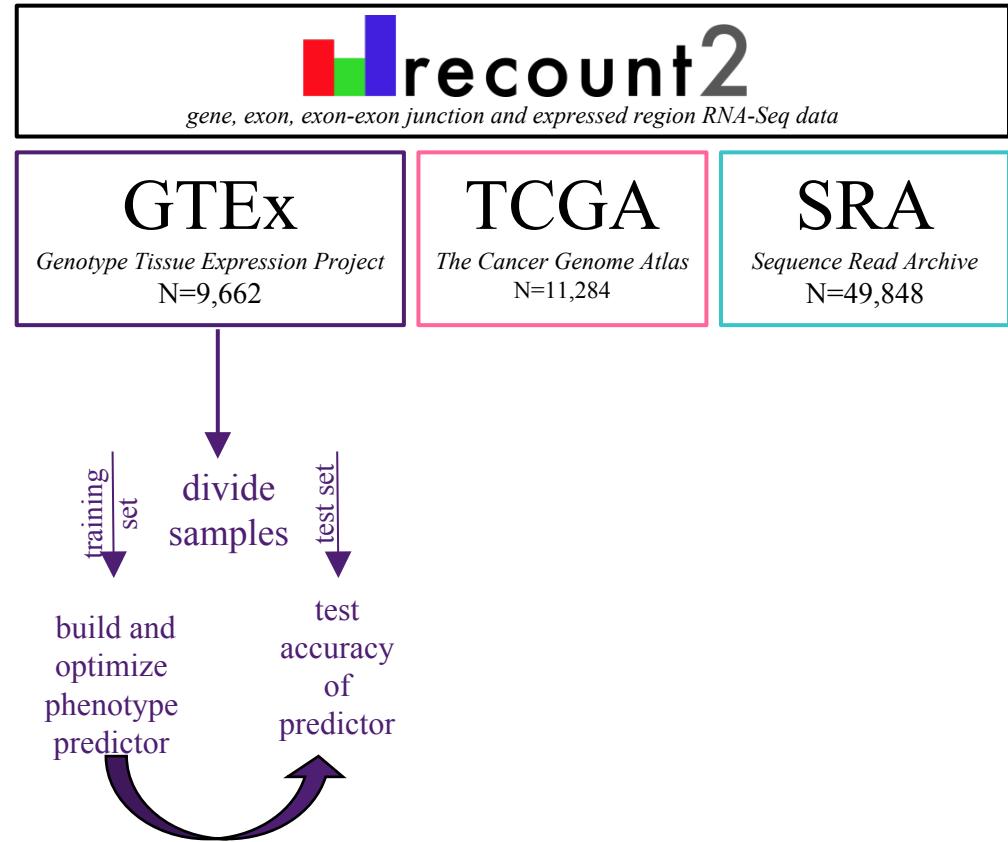
**Goal :**

to accurately  
predict critical  
phenotype  
information for  
all samples in  
*recount*



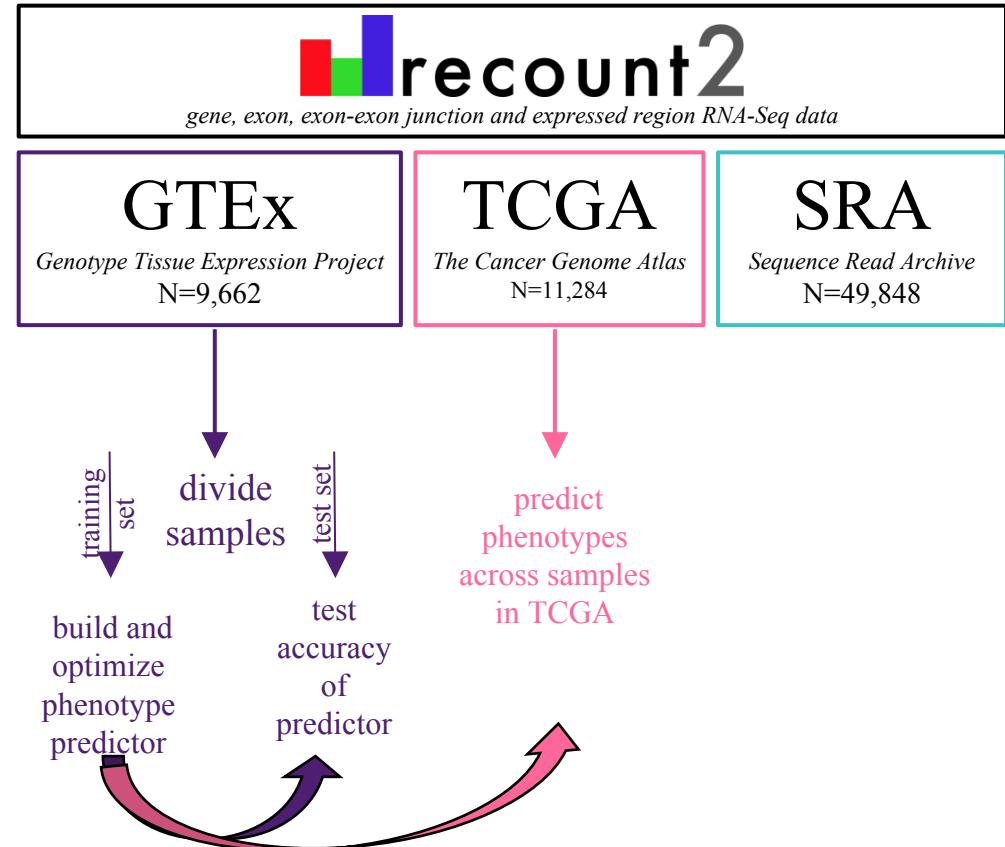
**Goal :**

to accurately predict critical phenotype information for all samples in *recount*



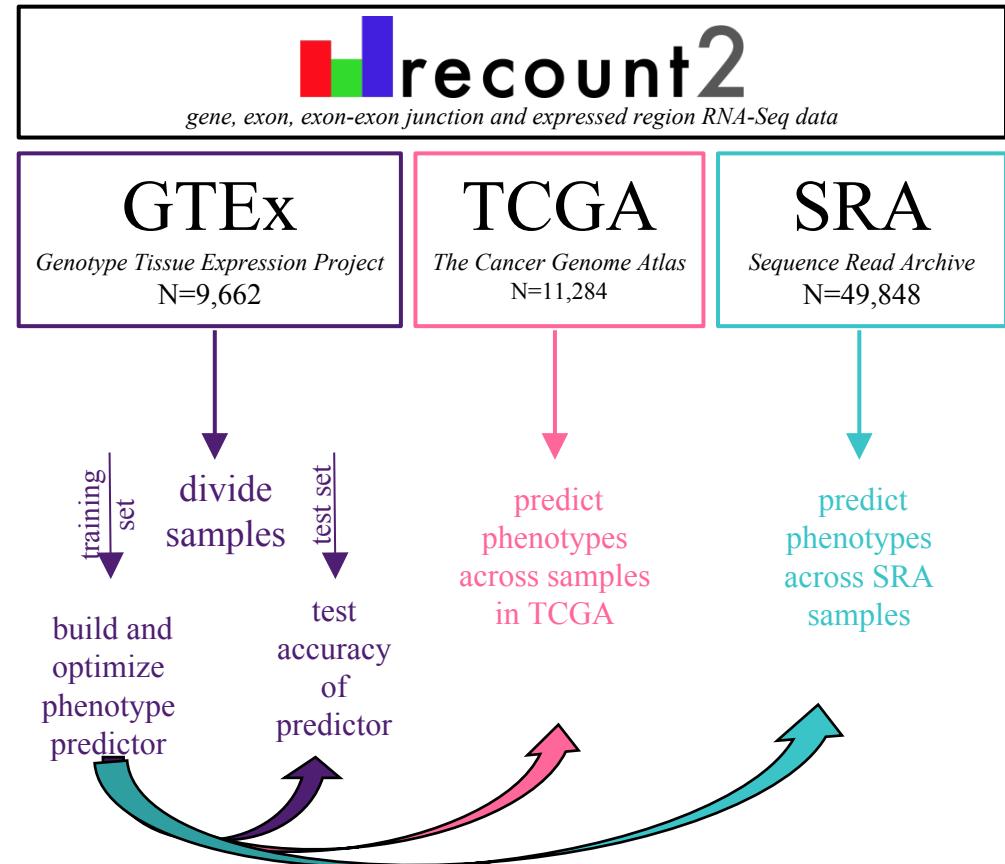
**Goal :**

to accurately predict critical phenotype information for all samples in *recount*

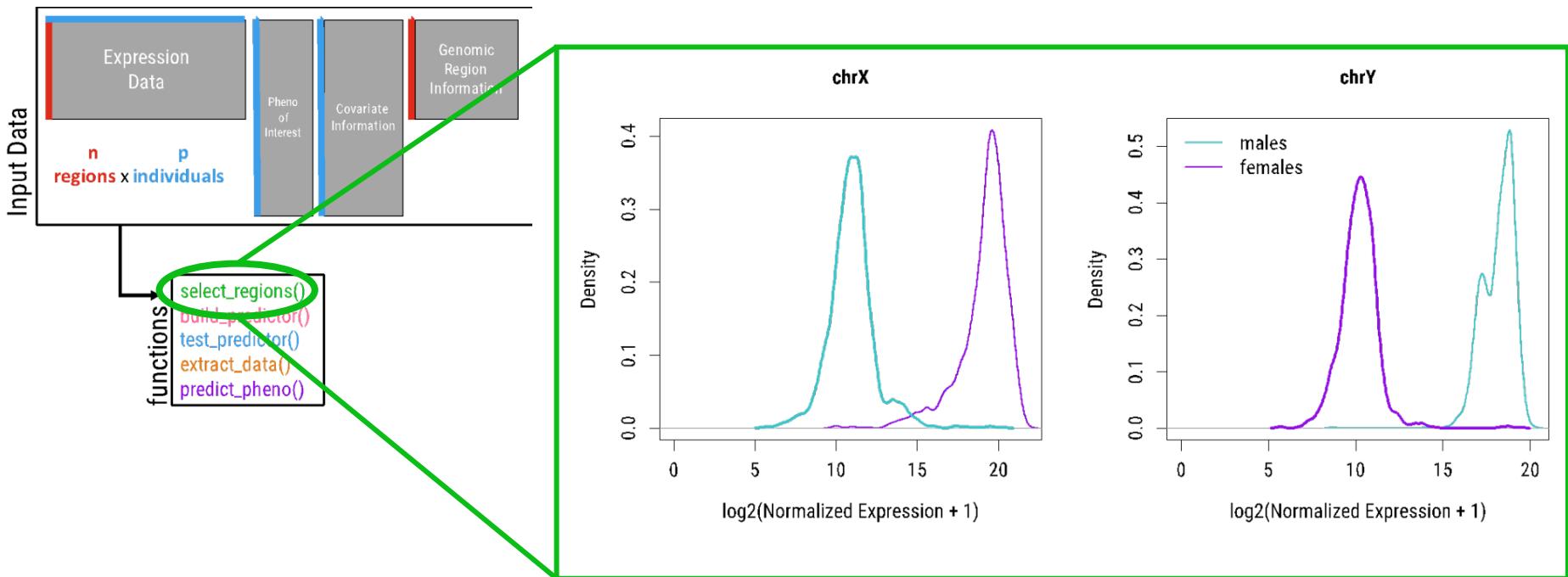


**Goal :**

to accurately predict critical phenotype information for all samples in *recount*

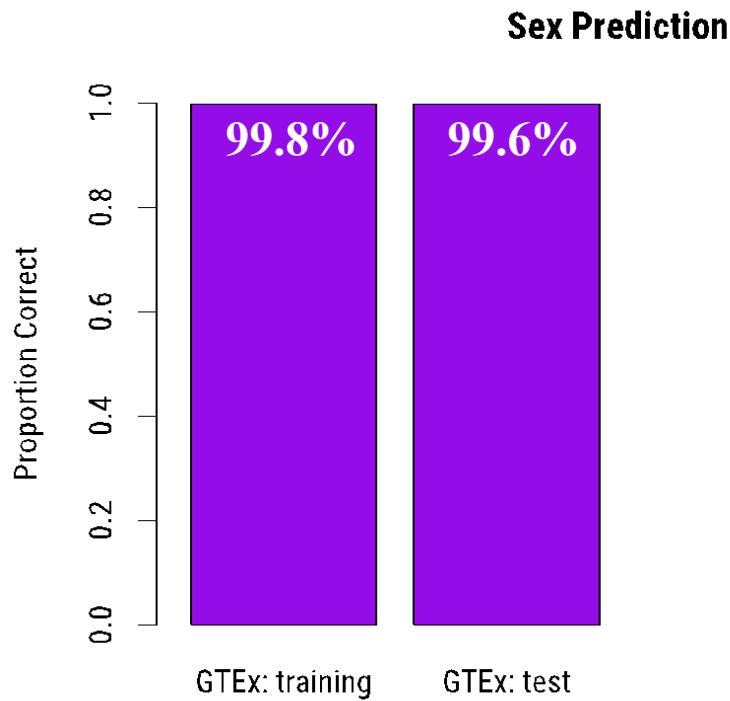


# select\_regions()



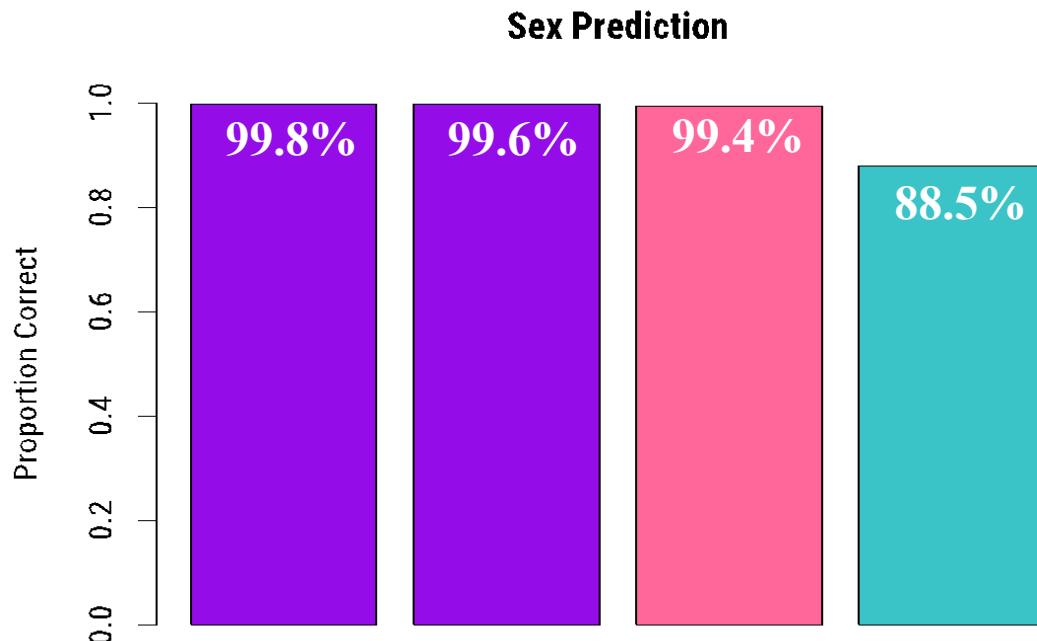
**Output:**  
Coverage matrix (data.frame)  
Region information (GRanges)

Sex  
prediction is  
accurate  
across data  
sets

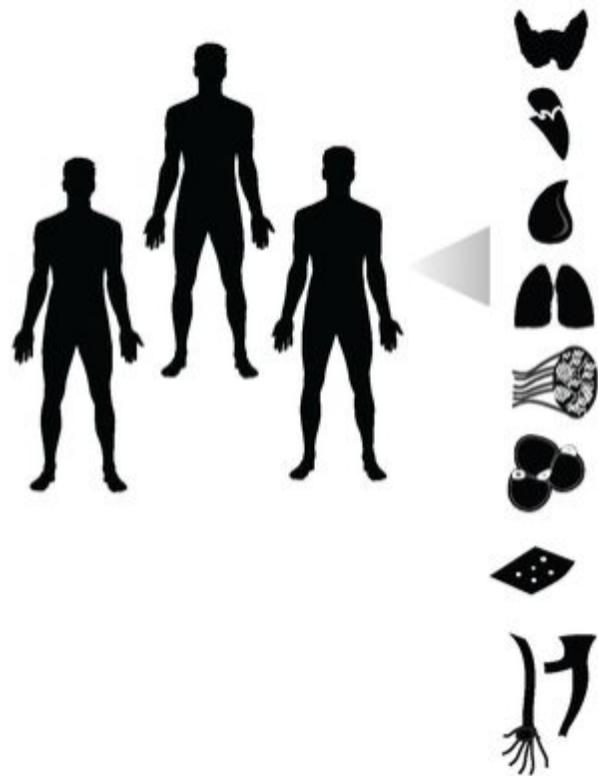


<b>Number of Regions</b>	20	20
<b>Number of Samples (N)</b>	4,769	4,769

Sex  
prediction is  
accurate  
across data  
sets

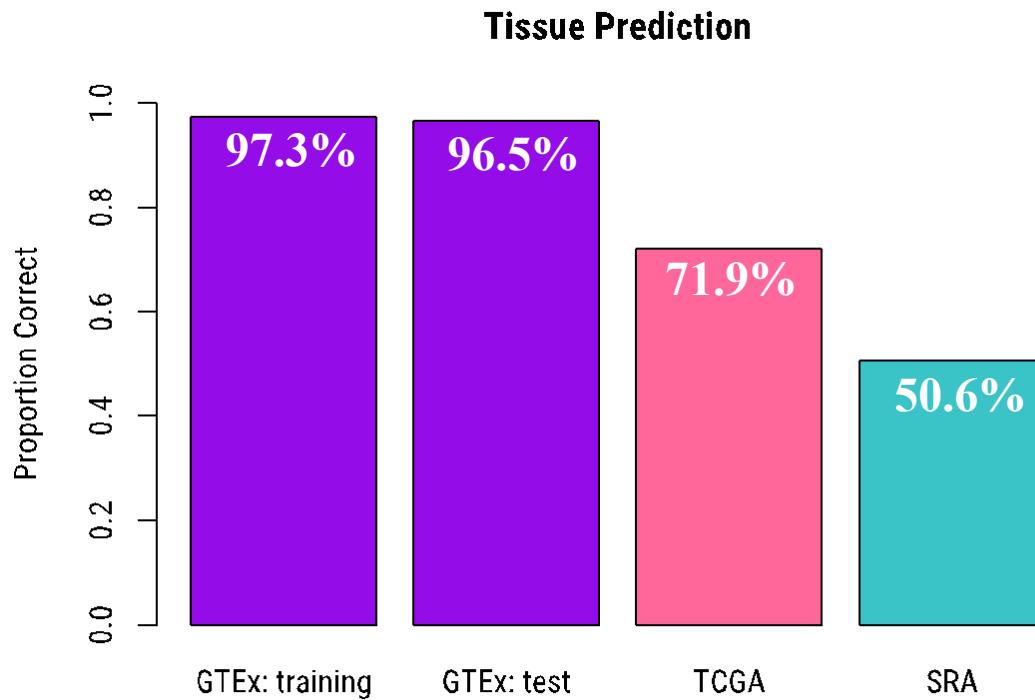


<b>Number of Regions</b>	20	20	20	20
<b>Number of Samples (N)</b>	4,769	4,769	11,245	3,640



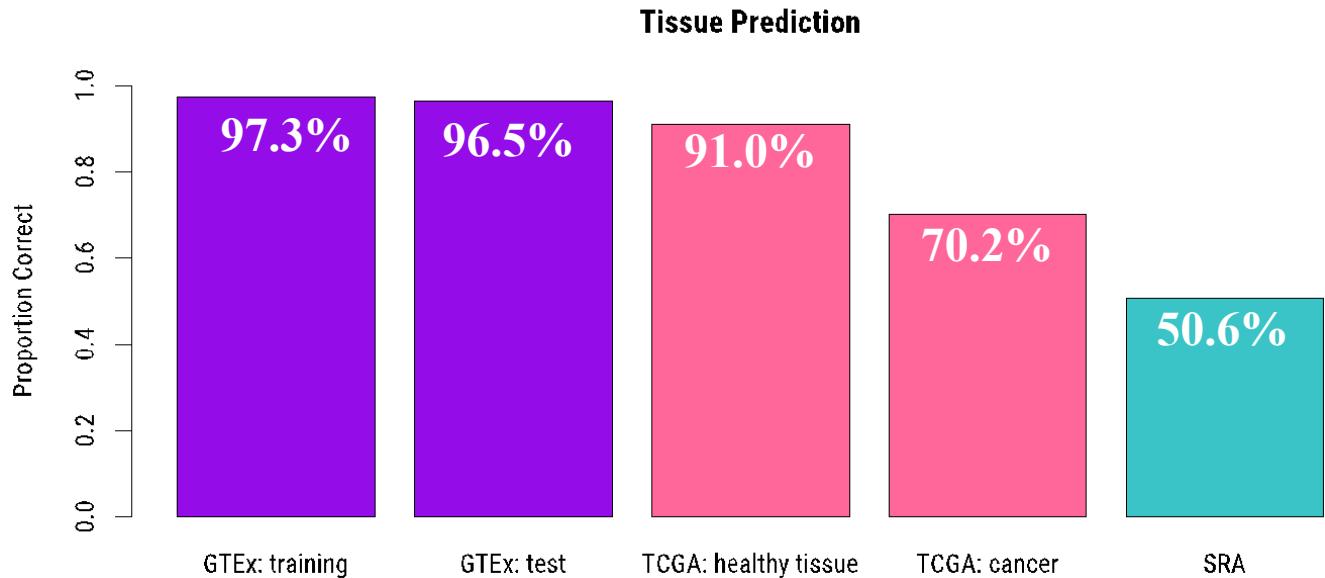
Can we use  
expression data  
to predict  
tissue?

Tissue  
prediction is  
accurate  
across data  
sets



<b>Number of Regions</b>	589	589	589	589
<b>Number of Samples (N)</b>	4,769	4,769	7,193	8,951

Prediction is  
more  
accurate in  
healthy  
tissue



Number of Regions	589	589	589	589	589
Number of Samples (N)	4,769	4,769	613	6,579	8,951

```
> library('recount')

> download_study( 'ERP001942', type='rse-gene')

> load(file.path('ERP001942 ', 'rse_gene.Rdata'))

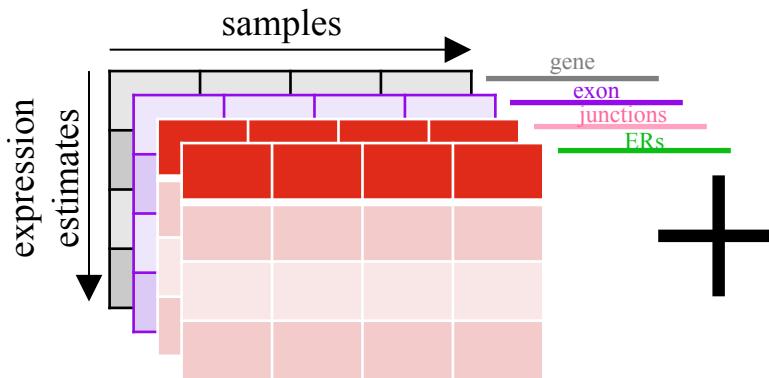
> rse <- scale_counts(rse_gene)

> rse_with_pred <- add_predictions(rse_gene)
```

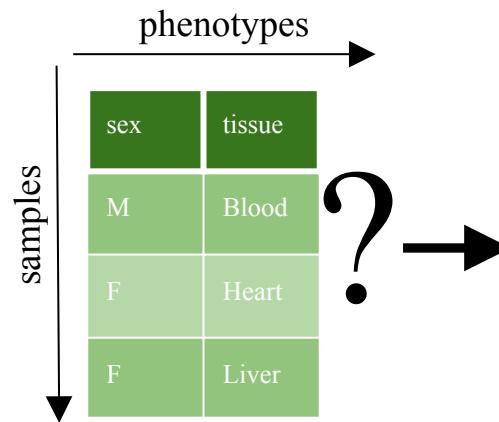
<https://github.com/leekgroup/recount-analyses/>

The logo for recount2 features the word "recount" in a bold, black, sans-serif font. The letter "r" is composed of three vertical bars in red, green, and blue. The letter "c" is also composed of these three colors. The letters "e", "o", "u", and "n" are in black. A large, stylized number "2" is positioned to the right of "recount".

*expression data for ~70,000 human samples*



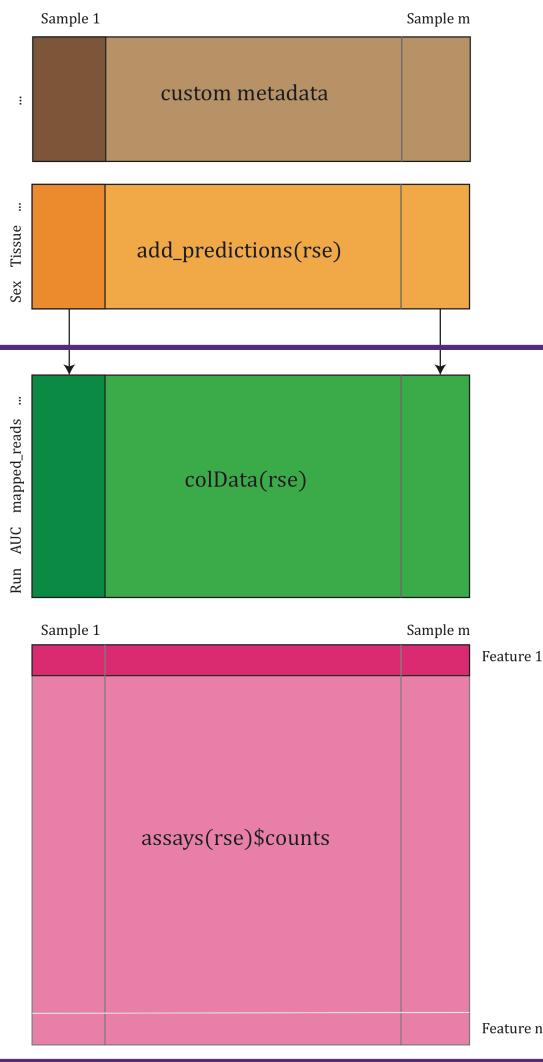
GTEX	SRA	TCGA
N=9,962	N=49,848	N=11,284



Answer meaningful  
questions about  
human biology and  
expression



download\_study()  
load()



# LIBD

## Collaborators

### The Leek Group

*Jeff Leek*

*Shannon Ellis*

### Hopkins

*Ben Langmead*

*Chris Wilks*

*Kai Kammers*

*Kasper Hansen*

*Margaret Taub*

### OHSU

*Abhinav Nellore*

### LIBD

*Andrew Jaffe*

*Emily Burke*

*Stephen Semick*

*Carrie Wright*

*Amanda Price*

*Nina Rajpurohit*

## Funding

NIH R01 GM105705

NIH IR21MH109956

CONACyT 351535

AWS in Education

Seven Bridges

IDIES SciServer

LIEBER INSTITUTE for  
BRAIN DEVELOPMENT  
MALTZ RESEARCH LABORATORIES



**`http://research.libd.org/recountWorkshop/`**

**`help(package = recountWorkshop)`**

**`file.edit(`**

**`system.file('doc/recount-workshop.Rmd', package = 'recountWorkshop')`**

**`)`**

Leonardo Collado-Torres  
@fellgernon  
#bioc2017

LIEBER INSTITUTE *for*  
BRAIN DEVELOPMENT  
MALTZ RESEARCH LABORATORIES



*expression data for ~70,000 human samples*

(Multiple) Postdoc positions available to

- develop methods to process and analyze data from recount2
- use recount2 to address specific biological questions

This project involves the Hansen, Leek, Langmead and Battle labs at JHU

Contact: Kasper D. Hansen ([khansen@jhsph.edu](mailto:khansen@jhsph.edu) | [www.hansenlab.org](http://www.hansenlab.org))