LIBD

Using the recount2 resource and related tools

# recount2

Leonardo Collado-Torres @fellgernon @LieberInstitute #CONABIO2019

LIEBER INSTITUTE for BRAIN DEVELOPMENT MALTZ RESEARCH LABORATORIES

## History

Undergrad in Genomic Sciences



Division Leader

WINTER GENOMICS

Turning Data Into Forefront Knowledge

**Data Science** 

2009-2011

Ph.D. Biostatistics

JOHNS HOPKINS

BLOOMBERG SCHOOL

of PUBLIC HEALTH

2011-2016

Pls:

Jeff Leek: 2012+

Andrew Jaffe: 2013+

Staff Scientist I → II

Data Science Team I

LIBD

August 2016+

PI: Andrew Jaffe

#### Interests







RopenSci



@LIBDrstats 2018+



@CDSBMexico 2018+

- BioC 2008-2011, 2014, 2017
- useR!2013
- rOpenSci unconf 2018
- RStudio::conf 2019





@RLadiesBmore

Blog: <a href="http://lcolladotor.github.io">http://lcolladotor.github.io</a>
2011+



FB: 75k, Tw: 66k





Defunct: BmoreBiostats, Biostats Cultural Mixers

# recount2 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, RangedSummarizeExperiment 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

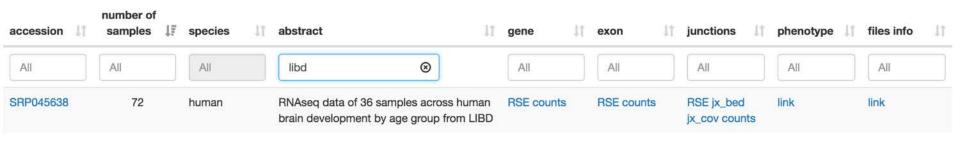
entries

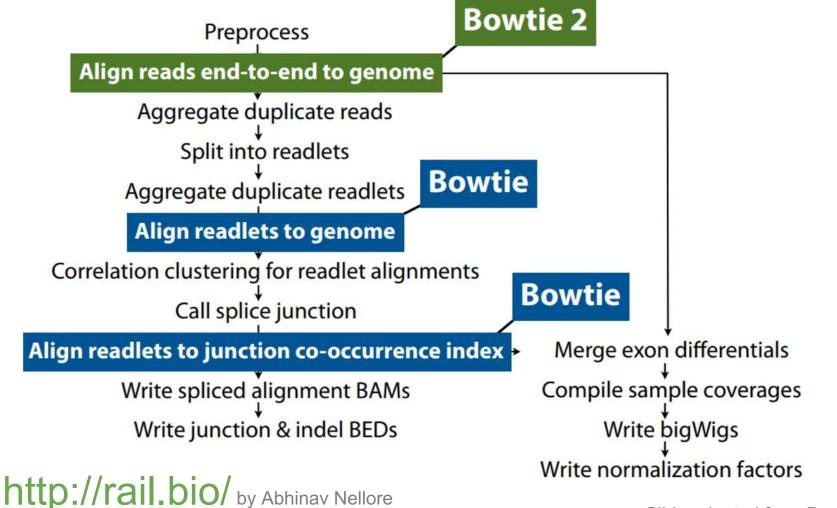
Collado-Torres L, Nellore A, Kammers K, Ellis SE, Taub MA, Hansen KD, Jaffe AE, Langmead B, Leek JT. Reproducible RNA-seq analysis using recount2. Nature Biotechnology, 2017. doi: 10.1038/nbt.3838.

#### The Datasets

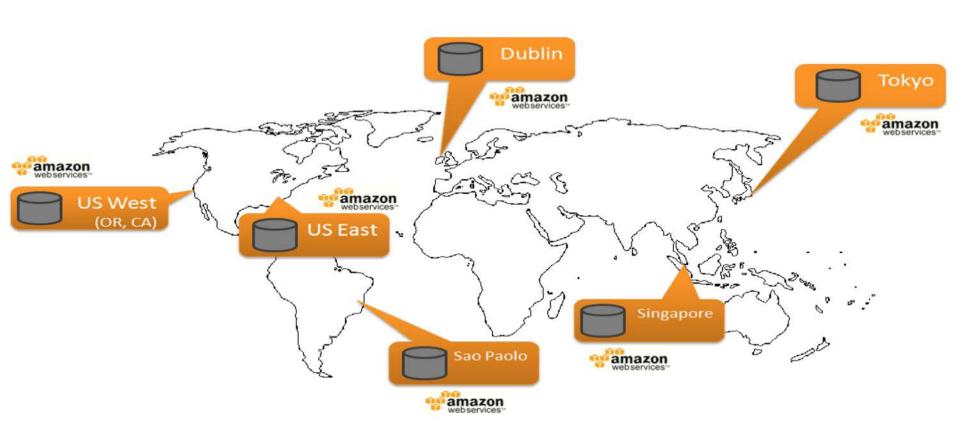
Show 10

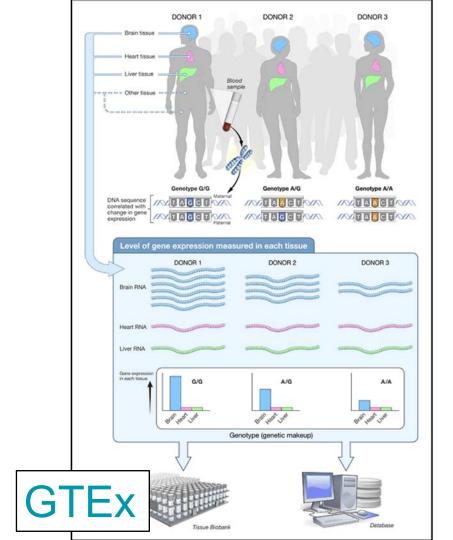
## https://jhubiostatistics.shinyapps.io/recount/





Slide adapted from Ben Langmead





#### NATIONAL CANCER INSTITUTE THE CANCER GENOME ATLAS

TCGA produced over

To put this into perspective, 1 petabyte of data



TCGA data describes

including

based on paired tumor and normal tissue sets collected from





#### TCGA RESULTS & FINDINGS



BASIS OF CANCER

Improved our understanding of the genomic underpinnings 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.



Revolutionized how cancer is classified

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



THERAPEUTIC

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

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.



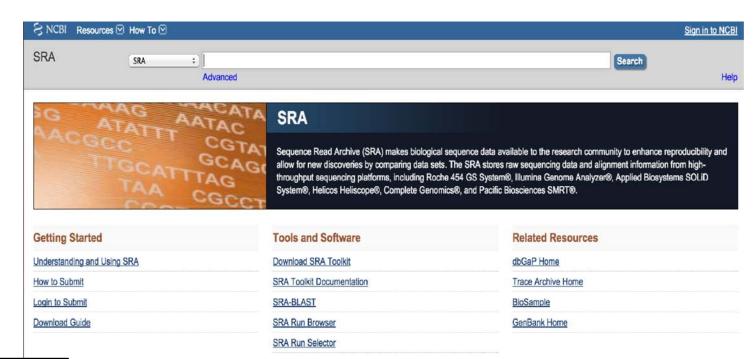
sets for scientists to access from anywhere The GDC also has many expanded capabilities that will allow researchers to across the United States answer more clinically



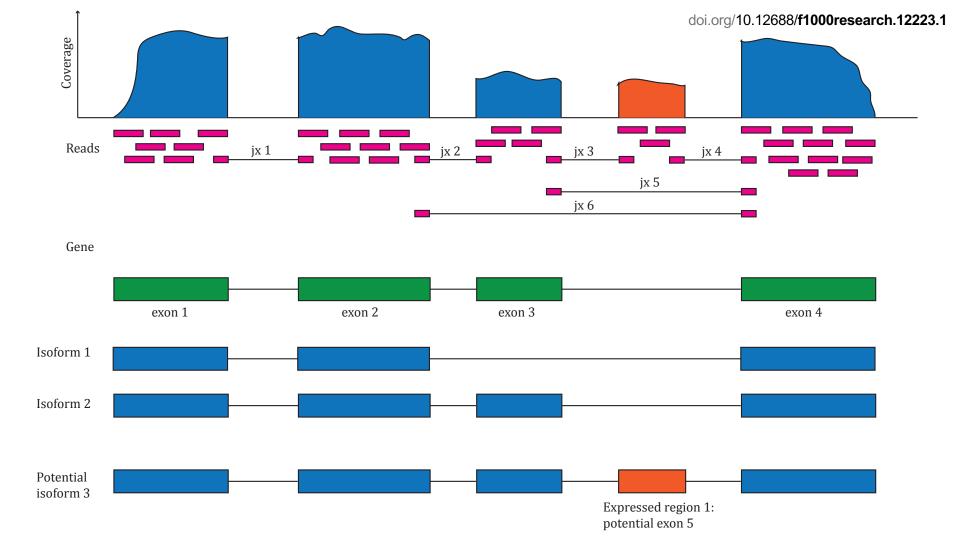
slide adapted from Shannon Ellis

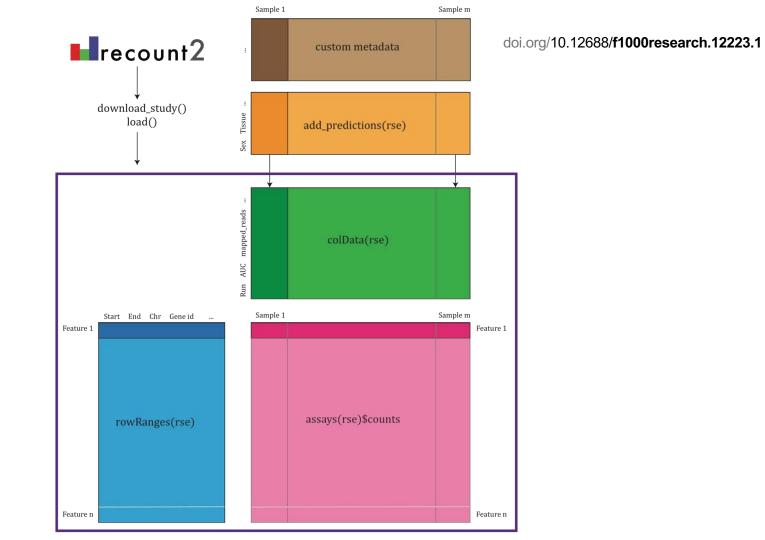
The Genomic Data

NCI-generated data









```
> 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/





Replying to @jtleek

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

RETWEETS









10:17 AM - 11 Apr 2017





LIKES





# recount 2 related projects

- Bioconductor recountWorkflow: doi.org/10.12688/f1000research.12223.1
- Shannon Ellis & Leek: phenotype prediction doi.org/10.1093/nar/gky102
- Jack Fu & Taub: transcript estimations doi.org/10.1101/247346
- Madugundu & Pandey (JHU): proteomics doi.org/10.1002/pmic.201800315
- Luidi-Imada & Marchionni (JHU): FANTOM (non-coding) and cancer doi.org/10.1101/659490
- Kuri-Magaña & Martínez-Barnetche (INSP Mexico): immune expression doi.org/10.3389/fimmu.2018.02679
- Ryten (UCL):

Guelfi: validating expressed region (ER) eQTLs doi.org/10.1101/591156

Zhang: improving the detection of ERs doi.org/10.1101/499103

# Snaptron: querying splicing patterns across tens of thousands of RNA-seq samples 8

Christopher Wilks X, Phani Gaddipati, Abhinav Nellore, Ben Langmead X

Bioinformatics, Volume 34, Issue 1, 01 January 2018, Pages 114–116,

https://doi.org/10.1093/bioinformatics/btx547

Published: 01 September 2017 Article history ▼

Christopher Wilks et al.

http://snaptron.cs.jhu.edu/snapcount\_vignette.html https://github.com/langmead-lab/snapr

#### Installation

```
# Install the development version from GitHub:
# install.packages("devtools")
devtools::install_github("langmead-lab/snapr")
```

#### Usage

snapr can be used with either a procedural interface

```
library(snapcount)

query_jx(compilation = "gtex", genes_or_intervals = "CD99")

#> class: RangedSummarizedExperiment

#> dim: 3485 9662

#> metadata(0):

#> assays(1): counts

#> rownames(3485): 28340058 28340273 ... 28352407 28352408

#> rowData names(12): DataSource:Type snaptron_id ... coverage_median

#> source_dataset_id

#> colnames(9662): 50099 50100 ... 59759 59760

#> colData names(322): rail_id Run ... junction_coverage

#> junction_avg_coverage

query_jx(compilation = "gtex", genes_or_intervals = "CD99", range_filters = exprs(samples_count == 10))

#> class: RangedSummarizedExperiment

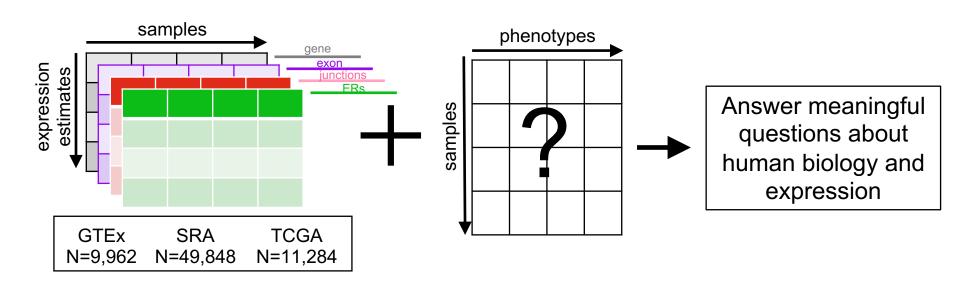
#> dim: 25 9662
```

#### Christopher Wilks et al.

http://snaptron.cs.jhu.edu/snapcount\_vignette.html https://github.com/langmead-lab/snapr



#### expression data for ~70,000 human samples



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

#### sra\_meta\$Sex

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

"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"

to accurately predict critical phenotype information for all samples in recount



gene, exon, exon-exon junction and expressed region RNA-Seq data

### GTEx

Genotype Tissue Expression
Project
N=9.662

#### **TCGA**

The Cancer Genome Atlas N=11,284

#### SRA

Sequence Read Archive N=49.848

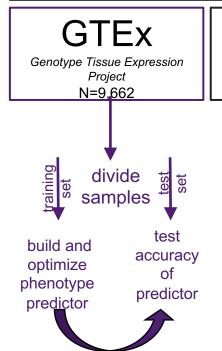
to accurately predict critical phenotype information for all samples in recount



The Cancer Genome

Atlas

N=11,284

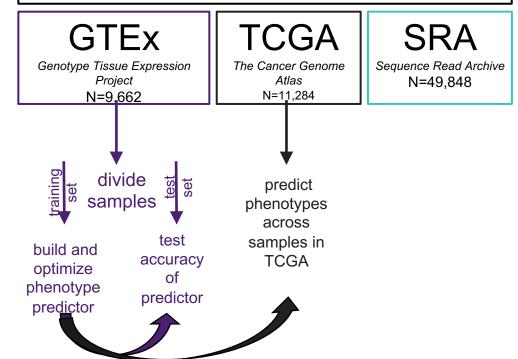


#### CGA | SRA

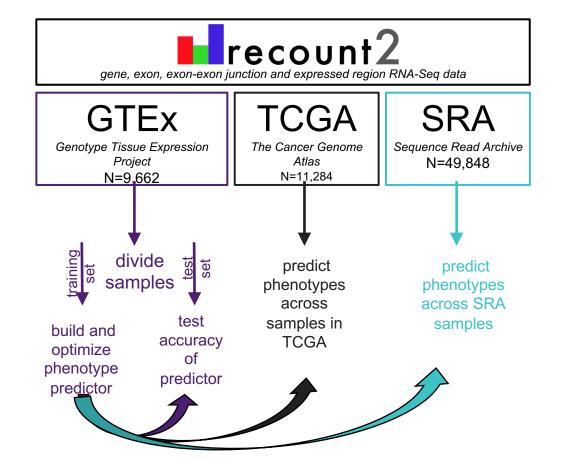
Sequence Read Archive N=49.848

to accurately predict critical phenotype information for all samples in recount

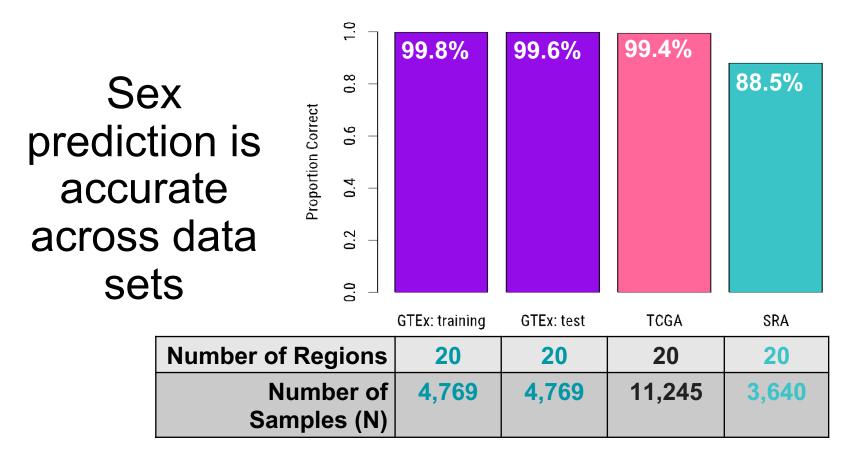




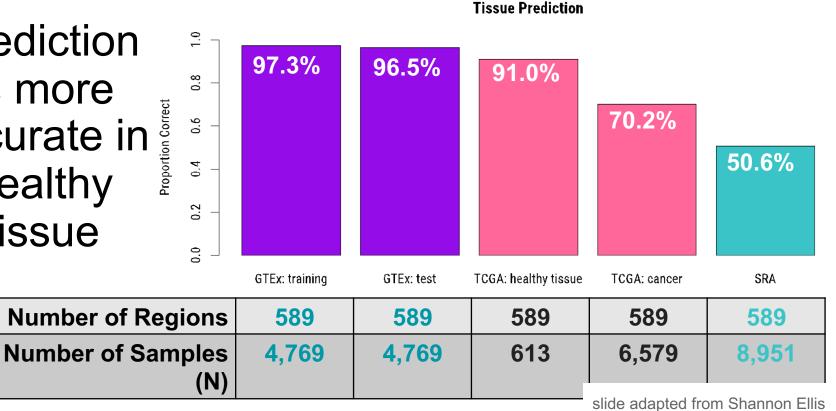
to accurately predict critical phenotype information for all samples in recount



#### **Sex Prediction**



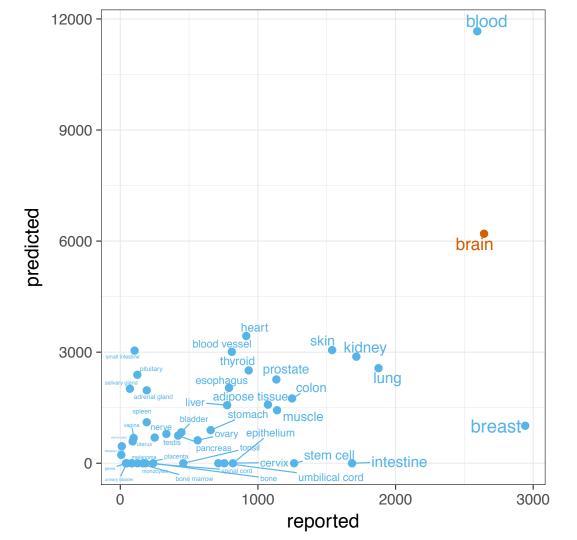
# Prediction is more accurate in healthy tissue



- > 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/







- 62 SRA studies
- 4,431 rows by 48 columns





Sex	Female	Male		
Age/Development	Fetus	Child	Adolescent	Adult
Race/Ethnicity	Asian	Black	Hispanic	White
Tissue Site 1	Cerebral cortex	Hippocampus	Brainstem	Cerebellum
Tissue Site 2	Frontal lobe	Temporal lobe	Midbrain	Basal ganglia
Tissue Site 3	Dorsolateral prefrontal cortex	Superior temporal gyrus	Substantia nigra	Caudate
Hemisphere	Left	Right		
Brodmann Area	1-52			
Disease Status	Disease	Neurological control		
Disease	Brain tumor	Alzheimer's disease	Parkinson's disease	Bipolar disorder
Tumor Type	Glioblastoma	Astrocytoma	Oligodendroglioma	Ependymoma
Clinical Stage 1	Grade I	Grade II	Grade III	Grade IV
Clinical Stage 2	Primary	Secondary	Recurrent	
Viability	Postmortem	Biopsy		
Preparation	Frozen	Thawed		







#### Reproducibility document

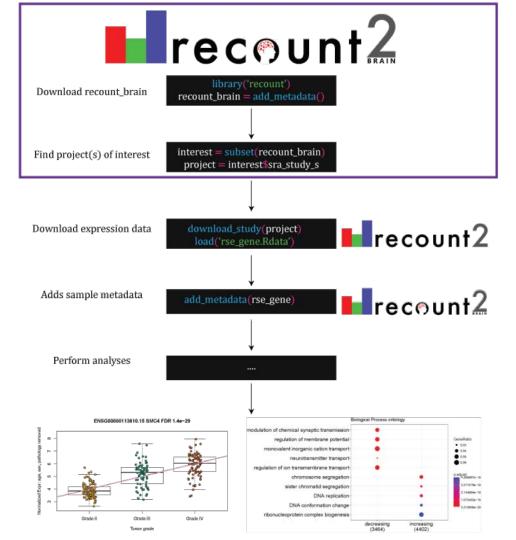
#### https://github.com/LieberInstitute/recount-brain/tree/master/metadata\_reproducibility

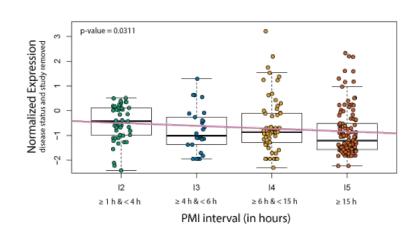
- Overall curation steps: starts by downloading SRA Run Table info, then info from the publications
- Details for each SRA study 6, SRP019762

#### https://doi.org/10.1038/ncomms4584

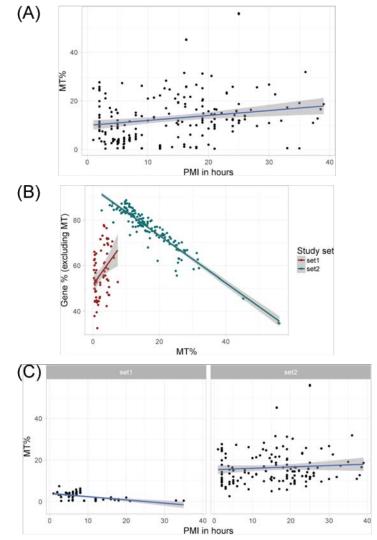
- · Methods:
  - Samples used for metabolite and RNA-seq experiments: "from frozen postmortem tissue" / "corresponding to Brodmann area 10"
    - Viability: Postmortem
    - Preparation: Frozen
    - Brodmann Area: 10
- · Supplementary Information
  - Supplementary Table 11. Sample information for the prefrontal cortex samples used in metabolite and RNA-seq measurements:
    - · Sex:
    - Age:
    - · RIN:
    - · PMI:
    - Brain Bank:







Replicates part of the GTEx PMI paper by Ferreira et al. doi.org/10.1038/s41467-017-02772-x



## Code Example:

research.libd.org/recount-brain/example\_PMI/example\_PMI.html

research.libd.org/recount-brain/example\_PMI/example\_PMI.Rmd

Replicates part of the GTEx PMI paper by Ferreira et al. doi.org/10.1038/s41467-017-02772-x





recount\_brain

- \* Jeff Leek presented Shannon Ellis' prediction work in Toronto (around April 2018) https://docs.google.com/presentation/d/1FgUZZU6pW91J7zH0OqrEgxfnV1Py\_ZGL3ZKHfbOZskY/edit#slide=id.g2f831fd4ae\_0\_306
- \* Dustin J. Sokolowski from Michael D. Wilson's lab is using recount2
- \* Dustin joins the project and merges recount-brain with GTEx and TCGA
- \* Met Sean Davis (NIH) at #biodata18, helped us with mapping to ontologies

The SRA samples in *recount-brain* are complemented by 1,409 GTEx (GTEx Consortium 2015) and 707 TCGA (Brennan et al. 2013; Cancer Genome Atlas Research Network et al. 2015) samples covering 13 healthy regions of the brain and 2 tumor types, respectively. In total, there are 6,547 samples with metadata in *recount-brain* with 5,330 (81.4%) present in *recount2* 



#### github.com/LieberInstitute/recount-brain

The recount-brain team

Hopkins
Ashkaun Razmara
Shannon E. Ellis
Jeff T. Leek

NIH
Sean Davis
LIBD

Andrew E. Jaffe

Funding

NIH R01 GM105705 NIH 1R21MH109956 NIH R01 GM121459 CIHR, NSERC Ontario Ministry of Research

University of Toronto

Dustin J. Sokolowski Michael D. Wilson Hosting recount 2

**IDIES SciServer** 

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```
> library('recount')
```

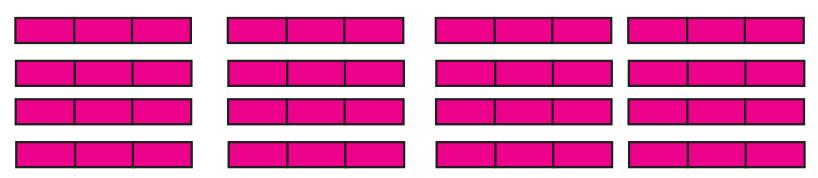
> download\_study( 'ERP001942', type='rse-gene')

> load(file.path('ERP001942', 'rse\_gene.Rdata'))

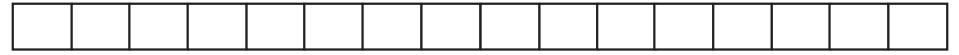
> rse <- scale\_counts(rse\_gene)

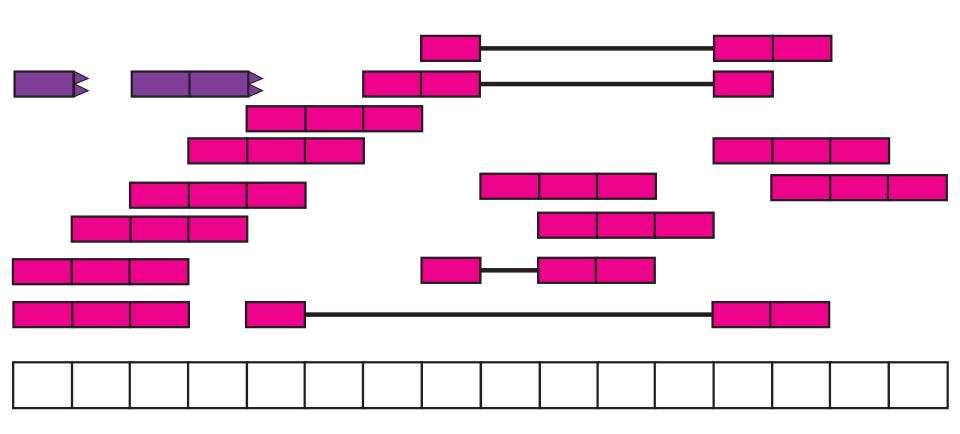
https://github.com/leekgroup/recount-analyses/

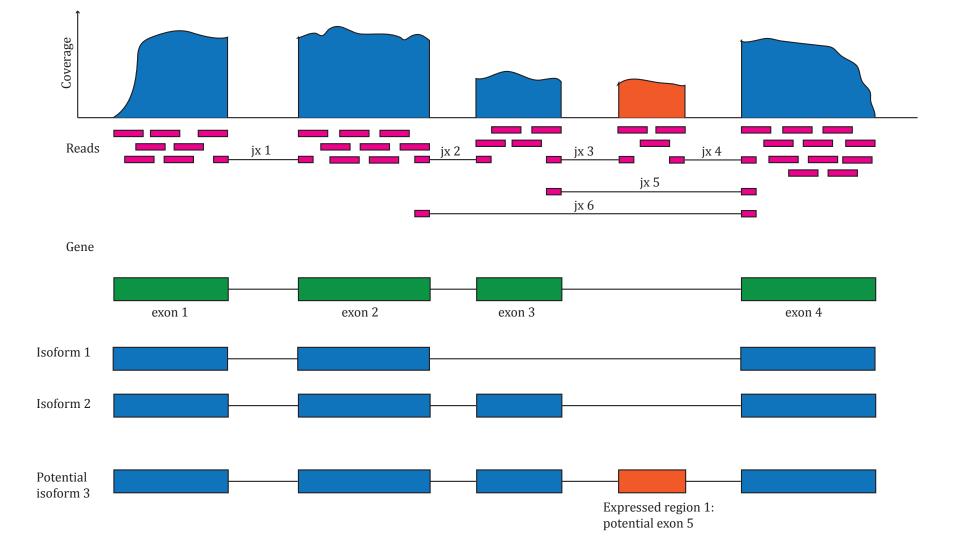
## Reads

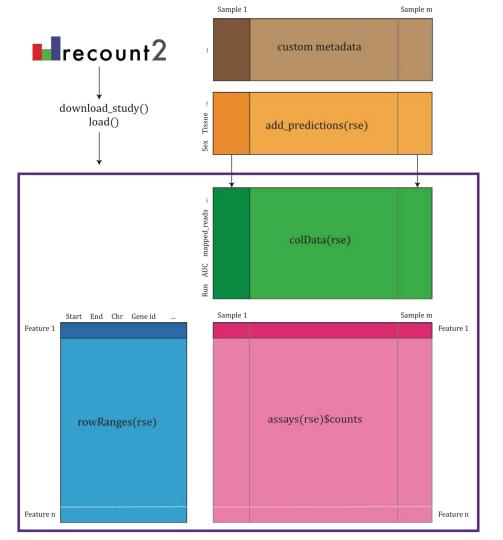


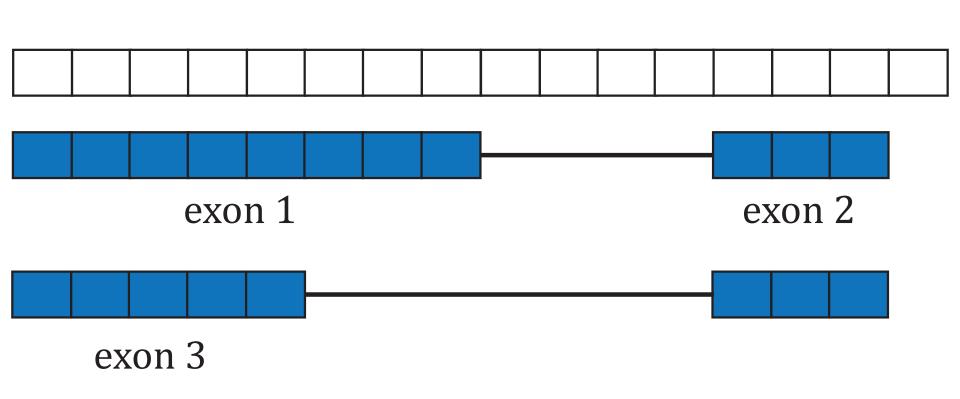
# Reference genome

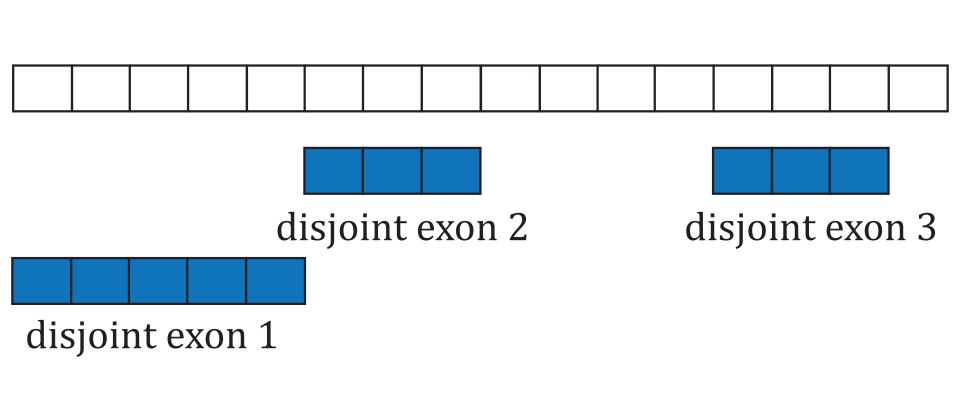


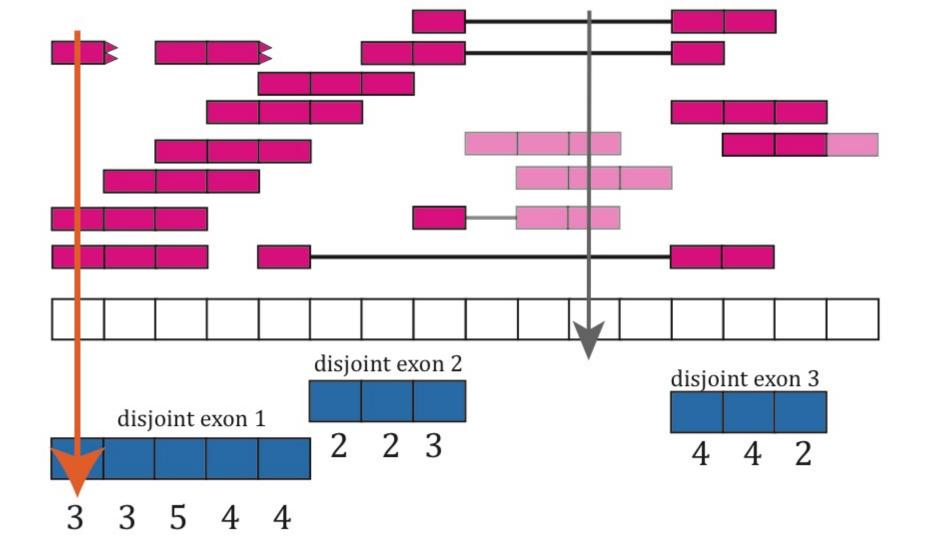


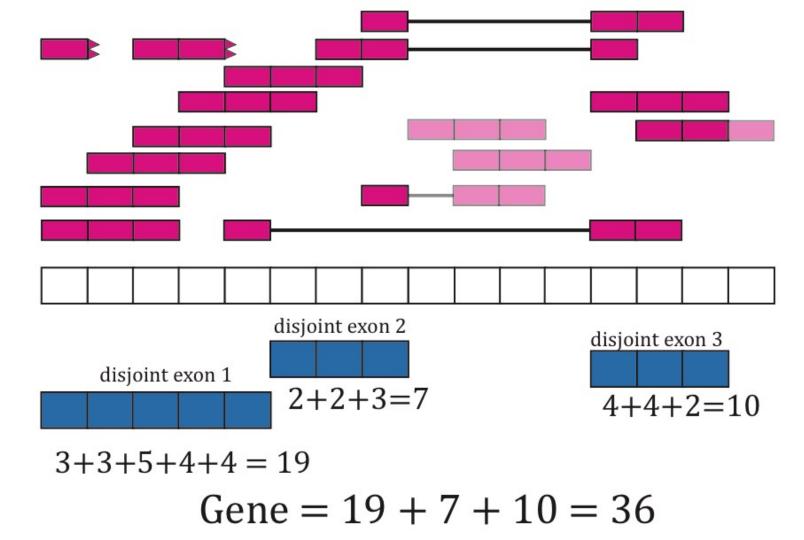




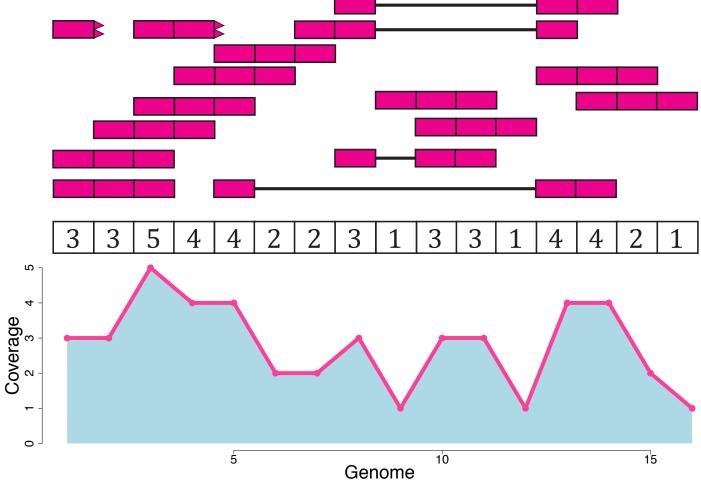








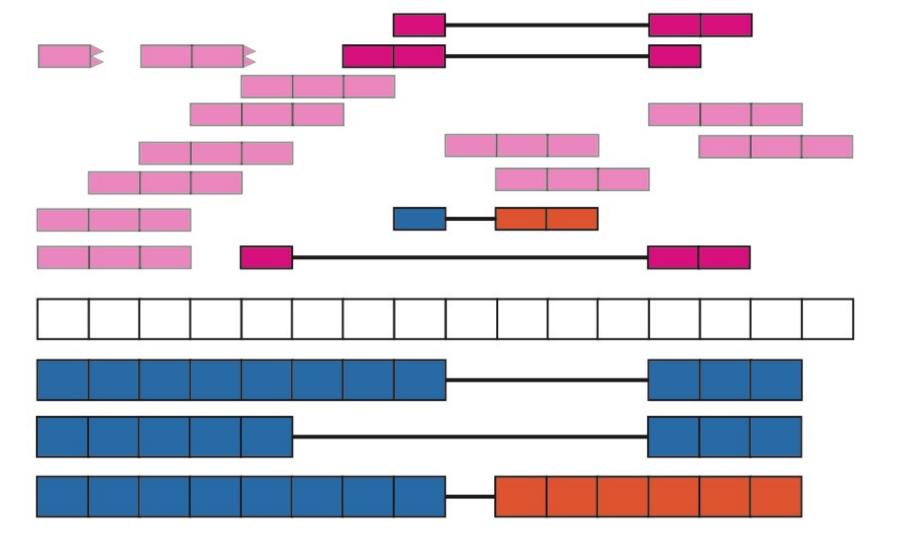
# $\frac{\sum_{i}^{n} coverage_{i}}{Read Length} * \frac{target}{mapped} = scaled read counts$

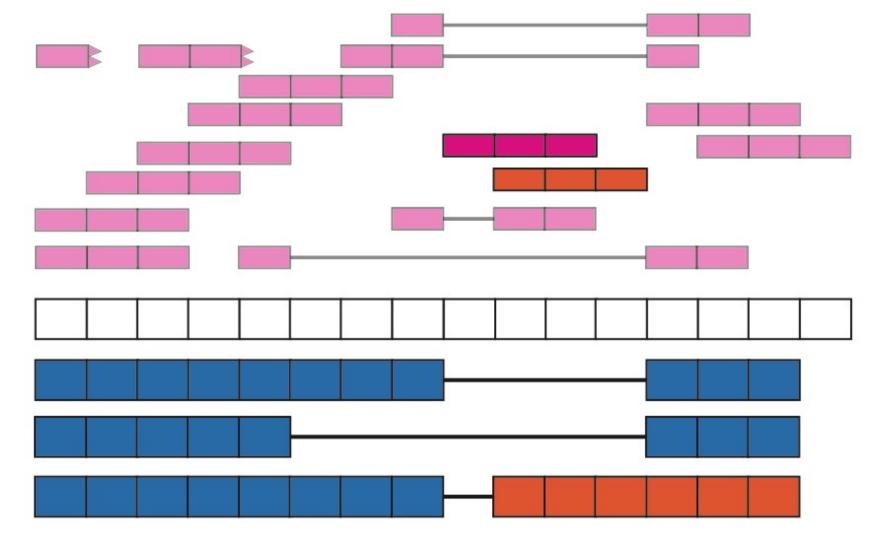


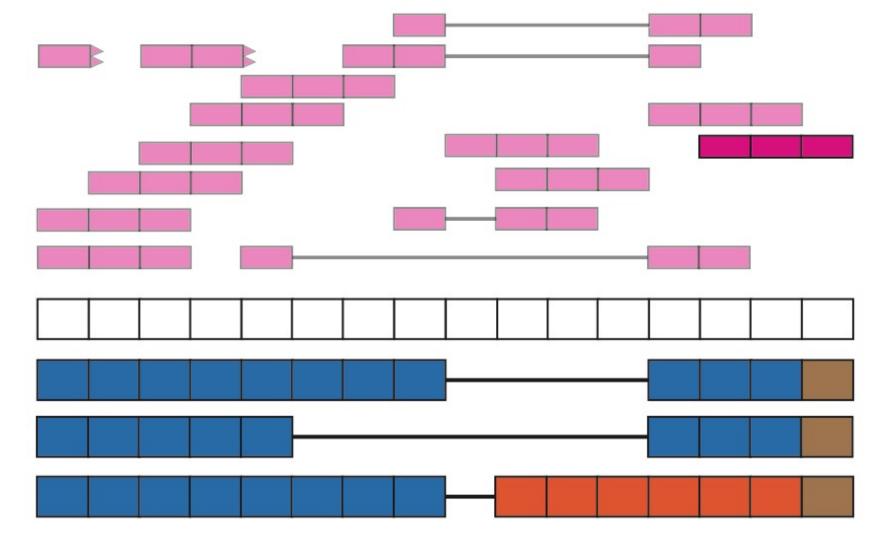
AUC = area under coverage = 45

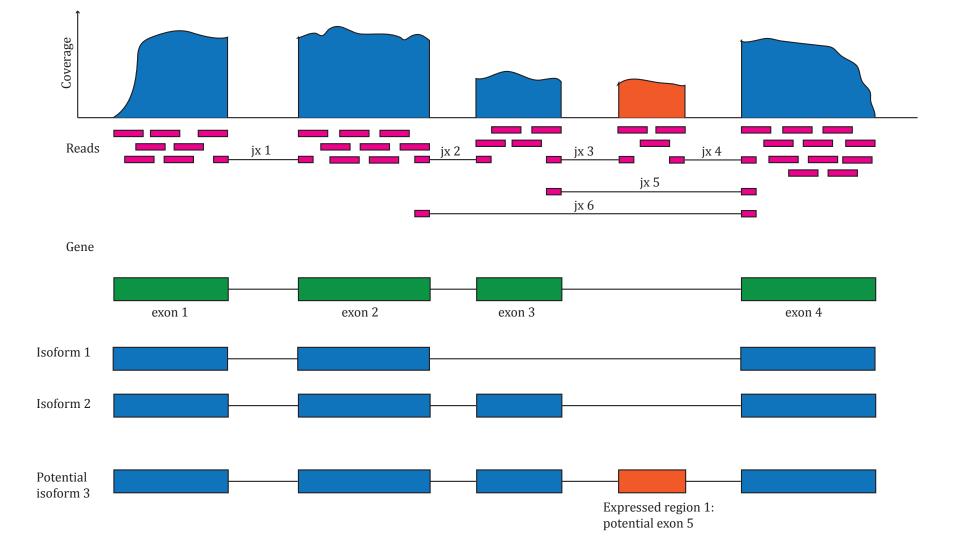
## $\frac{\sum_{i}^{n} coverage_{i}}{Read Length} * \frac{target}{mapped} = scaled read counts$

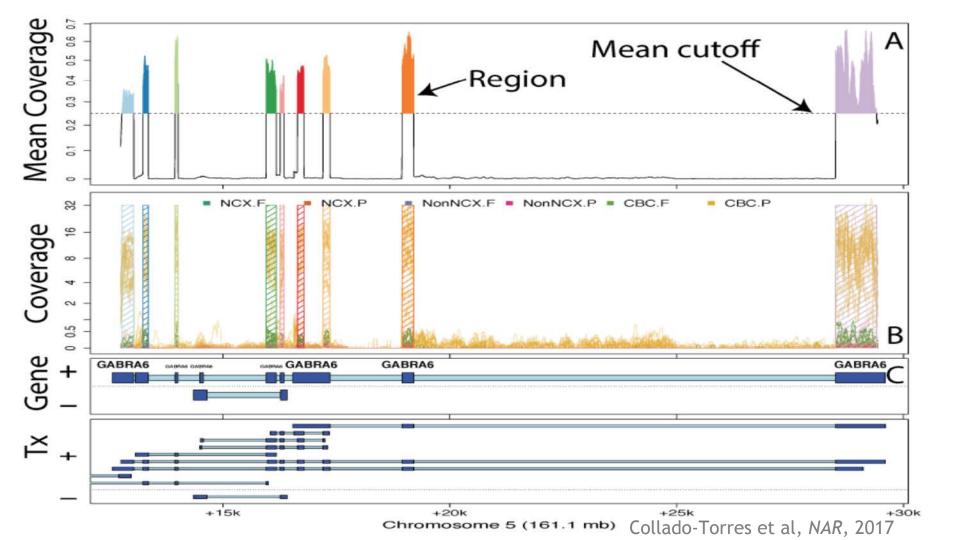
$$\frac{\sum_{i}^{n} coverage_{i}}{AUC} * target = scaled read counts$$











## Postmortem Human Brain Samples

Discovery data

Fetal

Infant

Child

Teen

Adult

50+

6 / group, N = 36

Replication data

Fetal

Infant

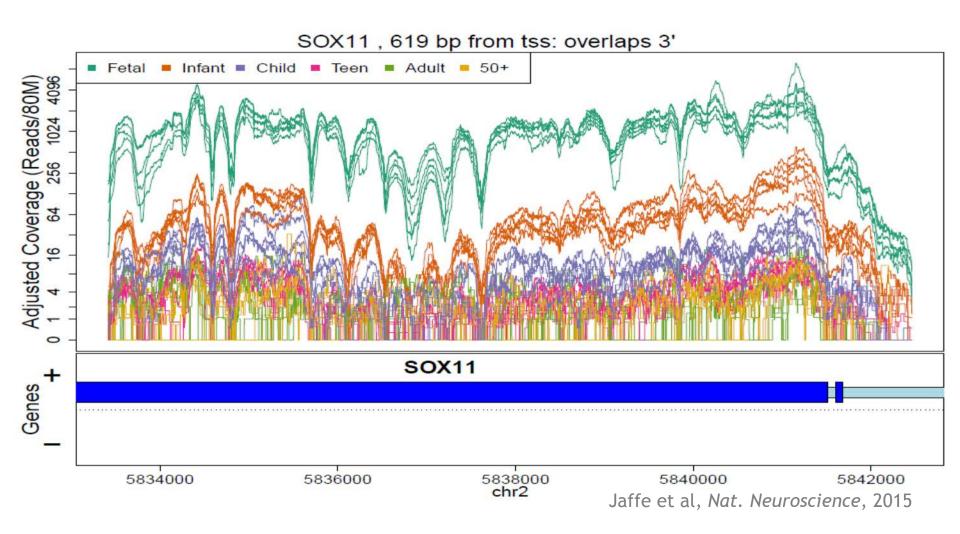
Child

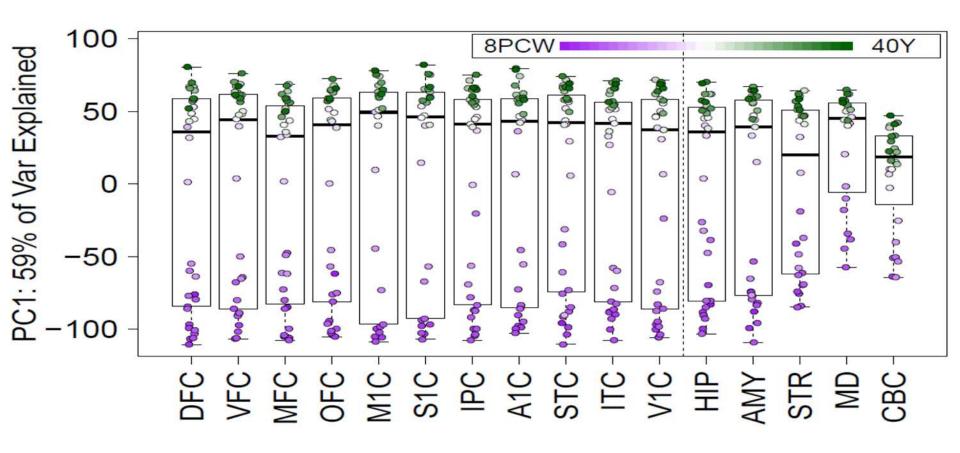
Teen

Adult

50+

6 / group, N = 36





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

### LIBD

Collaborators

UCSD

Shannon Ellis

OHSU

LIBD

Abhinav Nellore

Andrew Jaffe

**Hopkins** 

Jeff Leek Ben Langmead

Christopher Wilks

Kai Kammers

Kasper Hansen

Margaret Taub

Funding

NIH R01 GM105705

NIH 1R21MH109956

CONACyT 351535

AWS in Education

Seven Bridges

IDIES SciServer

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#### 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 | www.hansenlab.org)





help(package = recountWorkshop2019)

vignette('recount-workshop', 'recountWorkshop2019')

https://rebrand.ly/biocworkshops2019

Leonardo Collado-Torres @fellgernon #CONABIO2019

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