# RNA-Seq Data Analysis of Bipolar Disorder

Dohyoung Ko Akshay Nataraja



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# Project Motivation

This project aims to analyze RNA-Sequence data by preprocessing samples and developing a CLI pipeline to identify genes that are significantly overexpressed or underexpressed among bipolar disorder patient cohorts.



# Background



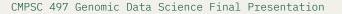
#### ■ Bipolar Disorder

**Overview:** Formerly known as manic-depressive illness, it is a chronic mental health condition characterized by extreme mood swings that include emotional highs and lows such as mania and depression. These mood swings can affect a person's sleep, energy, activity, judgment, behavior, and many other daily functions. In severe cases, visual or auditory hallucinations could occur.

#### Why did we chose to focus on it?

- Genetic Complexity
- Updated Datasets
- Familiarity with Topic





## Data Retrieval

Extra information on cohorts such as age, control or experimental group, and etc.

GEO Metadata

#### NCBI SRA Database

Access the dataset of cohorts from the existing study on SRA via EntreZ

#### **FASTQ**

FASTQ files of raw Illumina HiSeq-2000 high throughput RNA-Seq read data

#### Alignment SAM

SAM files storing alignments to reference genome after running HTSAT2

#### htseq-count with GFF

Download GFF gene mappings from NCBI Homo Sapiens database

#### Raw Count Matrix (CSV)

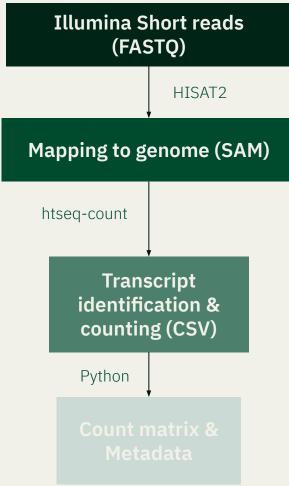
Matrix stored in CSV storing raw counts for each cohort in each row

https://www.ncbi.nlm.nih.gov/geo/geo2r/?acc=GSE80336, https://www.ncbi.nlm.nih.gov/sra/?term=Transcriptome+profiling+of+the+human+dorsal+striatum+in+bipolar+disorder

# Data Preprocessing

**HISAT2**: A sequence alignment program that maps our RNA sequences to a reference genome. Indexes the reference genome by creating a hash table of short and non-overlapping substrings of the reference genome that are used to find potential alignment locations. The tool scans the short reads we inputted and matches it to the substrings previously identified.

**Htseq-count:** Takes in the SAM file along with the GFF file that holds the gene annotations. It then counts the number of reads that overlap with each feature matched with the gene annotations.



# DGE Analysis Pipeline

- Command line application streamlining the whole RNA-Seq DGE analysis process
- Provides one-click solution under High
   Performance Computing Environment
  - BioProject ID on NCBI and GEO
- Written in Python, R, and shell
- https://www.ncbi.nlm.nih.gov/geo/geo2r/?acc=GSE80336

Submission date
Last update date
Contact name
E-mail(s)
Consisting approximation and The Contact page 4

Organization name The Scripps Research Institute - Scripps Florida

Department Neuroscience
Street address 130 Scripps Way
City Jupiter

City Jupiter State/province Florida ZIP/Postal code 33458 Country USA

Platforms (1) GPL11154 Illumina HiSeq 2000 (Homo sapiens)

GSM2124740 control-6

Relations

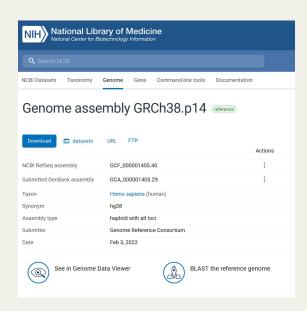
BioProject PRJNA318642 SRA SRP073382



Alliance de recherche numérique du Canada

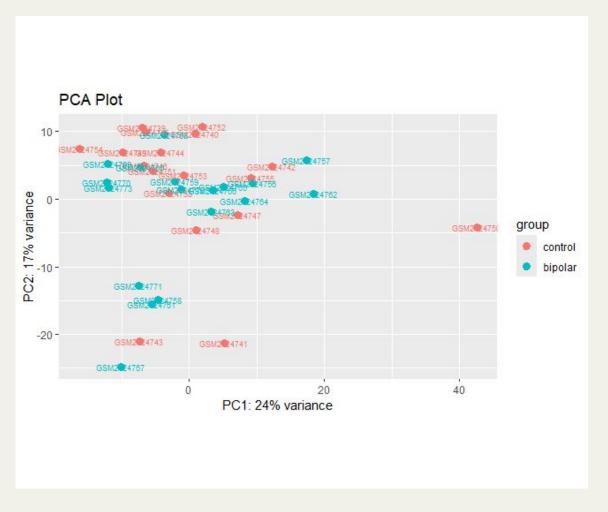
### NCBI Reference Genome

GCF\_000001405.40\_GRCh38.p14\_genomic.gtf.gz GCF\_000001405.40\_GRCh38.p14\_genomic.fna.gz



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arredings coded_creekounds exc	202. 00 27 10.25	2011	

Index of /company of all/CCE/000/001/405/CCE 000001405 40 CDCh20 = 14



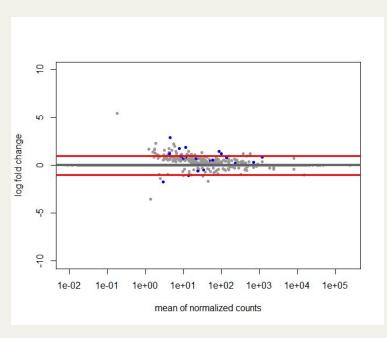
PCA's goal is to reduce the dimensionality between the two groups and visualize the main sources of variability within our data. Our data is split between control and the bipolar cohorts.

#### Observations:

Partial Overlap between the control and bipolar group.

But they key thing to note is that the total variance captured by the plot is only 41% (24% from PC1 and (17% from PC2). Ideally, our total variance should be at least 70% captured.

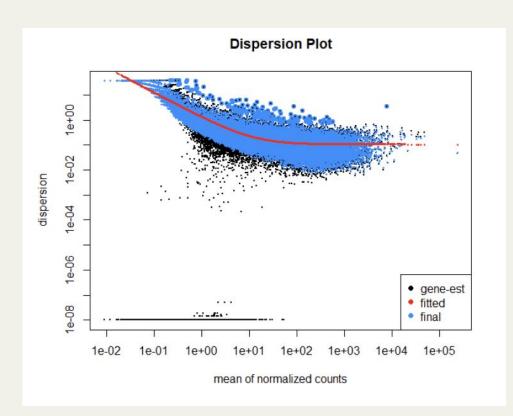
An explanation could be high dimensionality of the RNA-Seq, or notable localized changes in the gene expression.



This MA plot illustrates gene expression changes between two cohorts, "control" and "bipolar," in this case study.

The plot indicates that genes with lower mean expression values exhibit highly variable log fold changes. The blue dots in the figure represent differentially expressed genes (DEGs) that have adjusted p-values below the threshold of 0.05. Genes that meet the filtering criteria of a p-value less than 0.05 and a fold change of either greater than 2 (i.e.,  $\ell > 2$ ) or less than -2 (i.e.,  $\ell < -2$ ) are considered statistically significant.

The red vertical lines in the graph mark these thresholds. A fold change greater than 2 (i.e.,  $\log \ell > 0$ ) indicates up-regulated genes, while a fold change less than 1 (i.e.,  $\log \ell < 0$ ) indicates down-regulated genes. Among the DEGs identified are five up-regulated genes and eight down-regulated genes. This analysis is further illustrated in the following plot.



DESeq2 offers a valuable criterion known as dispersion to assess whether its negative binomial model is appropriate for the given data.

Goal of the dispersion plot is to see if our dataset is well-fitted to the DESeq2 Model.

We can observe that the fitting function is monotonically decreasing. This graph indicates a continuous decrease in dispersion as the mean expression increases. The dispersion estimates generally cluster around the curve, suggesting that it it is well-fitted. However, there is some shrinking as we only have one replicate for each sample in the group.

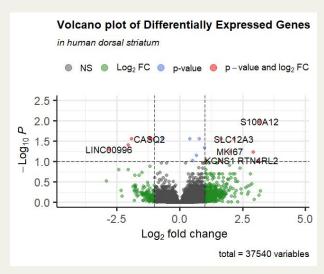


Figure 1: Volcano plot of DEGs in human dorsal striatum  $\,$ 

A fold change greater than 2 (i.e.,  $\log \ell > 0$ ) indicates up-regulated genes, while a fold change less than 1 (i.e.,  $\log \ell < 0$ ) indicates down-regulated genes.

ENSG ID 00000070915	baseMean 11.471846	log2FoldChange 2.156212	IfcSE	stat	pvalue	padj	symbol
00000070915	11.471846	2 156212					
00000010010		2.130212	0.487764	-4.420603	9.8426e-06	0.0275	SLC12A3
00000124134	19.494935	1.609598	0.403496	-3.989133	6.6315e-05	0.0927	KCNS1
00000148773	7.761059	2.001743	0.478666	-4.181919	2.8905e-05	0.0559	MKI67
00000163221	4.440045	3.184795	0.629121	-5.062293	4.1424e-07	0.0104	S100A12
00000165985	14.377425	2.915072	0.700847	-4.159356	3.1914e-05	0.0573	C1QL3
00000186907	4.295899	3.110264	0.782409	-3.975241	7.0308e-05	0.0931	RTN4RL2
00000198743	1185.410114	1.034511	0.258776	-3.997709	6.3958e-05	0.0927	SLC5A3
00000275395	87.451458	1.643967	0.359915	-4.567649	4.9322e-06	0.0275	FCGBP

Table 1: Complete list of up-regulated genes

							3
ENSG ID	baseMean	log2FoldChange	IfcSE	stat	pvalue	padj	symbol
00000118729	13.395064	-1.199622	0.263730	4.548668	5.3986e-06	0.0275	CASQ2
00000132744	3.952897	-2.052703	0.475144	4.320168	1.5591e-05	0.0392	ACY3
00000144681	12.769351	-1.141231	0.251624	4.535454	5.7479e-06	0.0275	STAC
00000239961	2.897106	-1.917298	0.431810	4.440140	8.9900e-06	0.0275	LILRA4
00000242258	2.019025	-2.799696	0.662369	4.226795	2.3704e-05	0.0497	LINC00996
00000242258	2.019025	-2.799696	0.662369	4.226795	2.3704e-05	0.0497	LINC00996

Table 2: Complete list of down-regulated genes

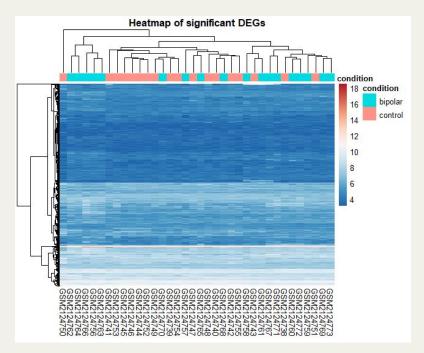


Figure 2: Heatmap of significant DEGs

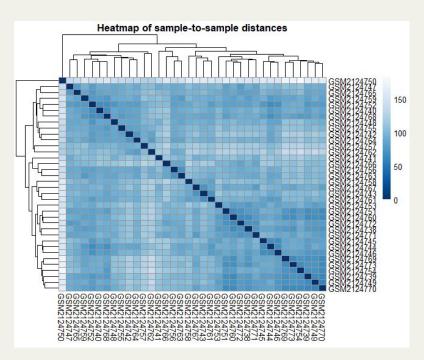


Figure 2: Heatmap of sample-to-sample distances

## Results on Normalized Dataset

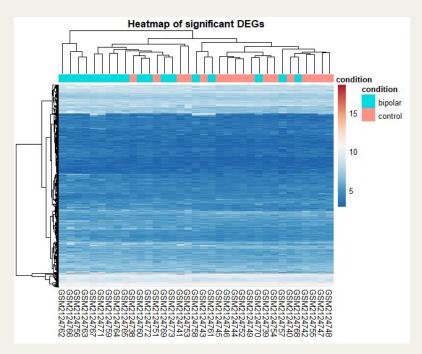


Figure 2: Heatmap of significant DEGs

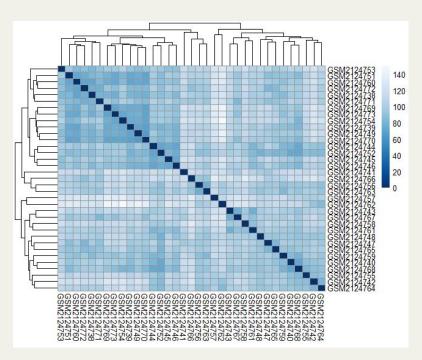


Figure 2: Heatmap of sample-to-sample distances

#### **Under-Expressed**

**LILRA4**- Its expression enhances immune regulation, contributing to the innate and adaptive immune response, including during viral infections and inflammatory diseases.

**ACY3**- This gene plays a significant role in maintaining renal function and detoxification processes.

**LINCO0996**- A long intergenic non-coding RNA, has been shown to suppress lung adenocarcinoma cells, indicating its role in tumor suppression and cellular homeostasis.

**CASQ2**- It is pivotal in regulating calcium homeostasis, influencing muscle contraction and relaxation cycles.

**STAC-** STAC enhances the activity of calcium channels and slows the inactivation of CACNA1C, which is critical for maintaining cellular calcium signaling.

#### Over-Expressed

**SLC12A3**- Down-regulation of SLC12A3 is predicted to disrupt processes in regulating ion homeostasis, cell volume, and neuronal excitability.

**KCNS1**- Plays a part in regulating neuronal excitability and synaptic function. Its down-regulation has been associated with altered synaptic transmission.

**MKI67**- Down-regulation of MKI67 has been linked to reduced cell division, often serving as a marker for low proliferative activity in cells.

**\$100A12-** Plays a part in regulating calcium-dependent processes, including muscle contraction and neuronal signaling, down-regulation may impair immune responses and antibacterial properties.

**C1QL3-** Involved in synaptic regulation, glucose homeostasis, and neuronal projections.

**RTN4RL2-** Is critical for neuronal development and plasticity; its down-regulation correlates strongly with bipolar disorder.

**SLC5A3**- Is essential for glucose uptake and metabolism.

**FCGBP-** Functions as a tumor suppressor gene, particularly in head and neck squamous cell carcinoma.

### Problems

#### Data Intensive Processes

- RAW FASTQ files were data intensive.
- Processing could not be done locally.
- Required a HPC to handle requests.

#### Preprocessing Complexity

- Preprocessing the data required multiple frameworks.
- Data was sourced from different locations.
- No streamlined pipeline to preprocess raw RNA-Seq data.

## Thank You

Any
Questions
:(: