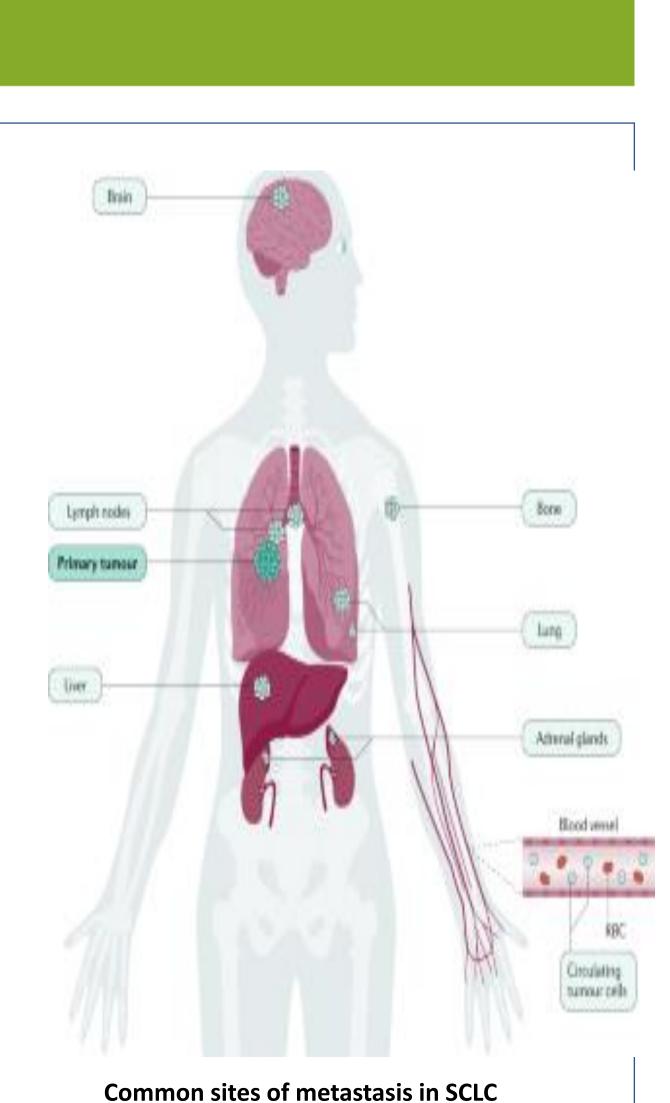


BIOINFORMATICS APPROACH TO UNDERSTAND THE TRANSCRIPTIONAL CHANGES IN SMALL CELL LUNG CANCER (SCLC) PATIENTS WITH BRAIN METASTASES

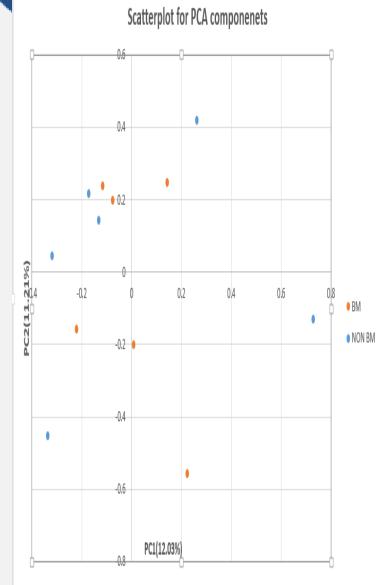
PINE.BIO OMICS DATA SENSE AND MORE

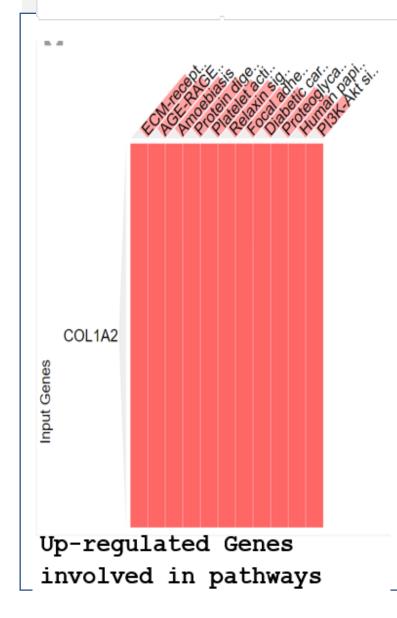
Introduction

Lung cancer is one of the major cause of mortality worldwide, making it a prominent cause of cancer related deaths (Tomazis*et al.*, 2020). It is believed that lung cancer can be cured if detected at an early stage with the help of various screening tools (Thakur *et al.*, 2020). But the early detection of lung cancer is very crucial as the tools available in medical industry are not advanced enough to detect the genes and the pathways responsible for the cause of cancer and even if it is diagnosed at early stage, it cannot be treated completely (Xu et al., 2021). The early detection is a crucial task as only 15% of the patients are detected at an early stage while others, that is 75% of the patients at diagnosed with an advanced or last stage (Jacobsen et al., 2017). Lung cancer can be grouped into two main subtypes namely: NSCLC (Non-Small Cell Lung Cancer) and SCLC (Small Cell Lung Cancer). NSCLC is most prominent with its prevalence of 85%-90% of population suffering from lung cancer (Herbstet al., 2018). Whereas SCLC is present in 10%-15% of population suffering from lung cancer (Molina et al., 2008). NSCLC (Non-Small Cell Lung Cancer) can be further grouped into adenocarcinoma and squamous cell carcinoma (Torre *et al.*,2012). But SCLC, when compared to NSCLC, has a fast rate of forming new cells, and also a high malignancy rate, resulting in a huge risk of metastases (Kalemkerianet al, 2013). SCLC has a very low (almost rare) survival rate and patients usually lose their lives due to this disease. Due to high metastatic rate of this disease, the cancerous (or tumour) cells spread from the primary site to the portions of the body including brain, liver, bones etc. Mostly this spread is diagnosed in (or towards) the brain. Approximately, at the time of primary diagnoses, about 10 percent of the SCLC patients are detected to have brain metastases and about 50 percent develops during the course of cancer from stage to stage (Quanet al, 2004). Hence, it becomes necessary to identify the cause of this spread and to curb it at its root cause only.



Results





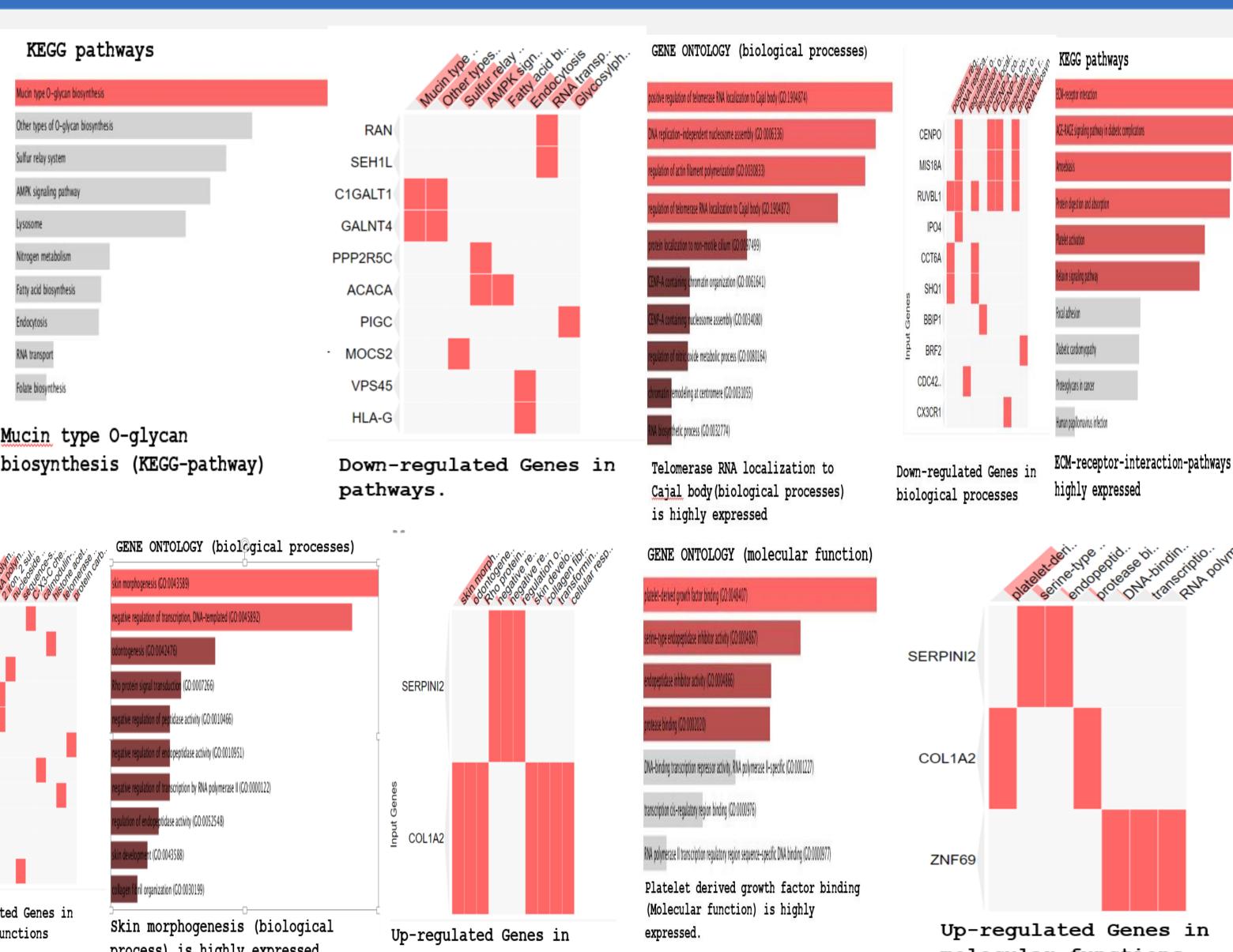
STATISTICAL ANALYSIS

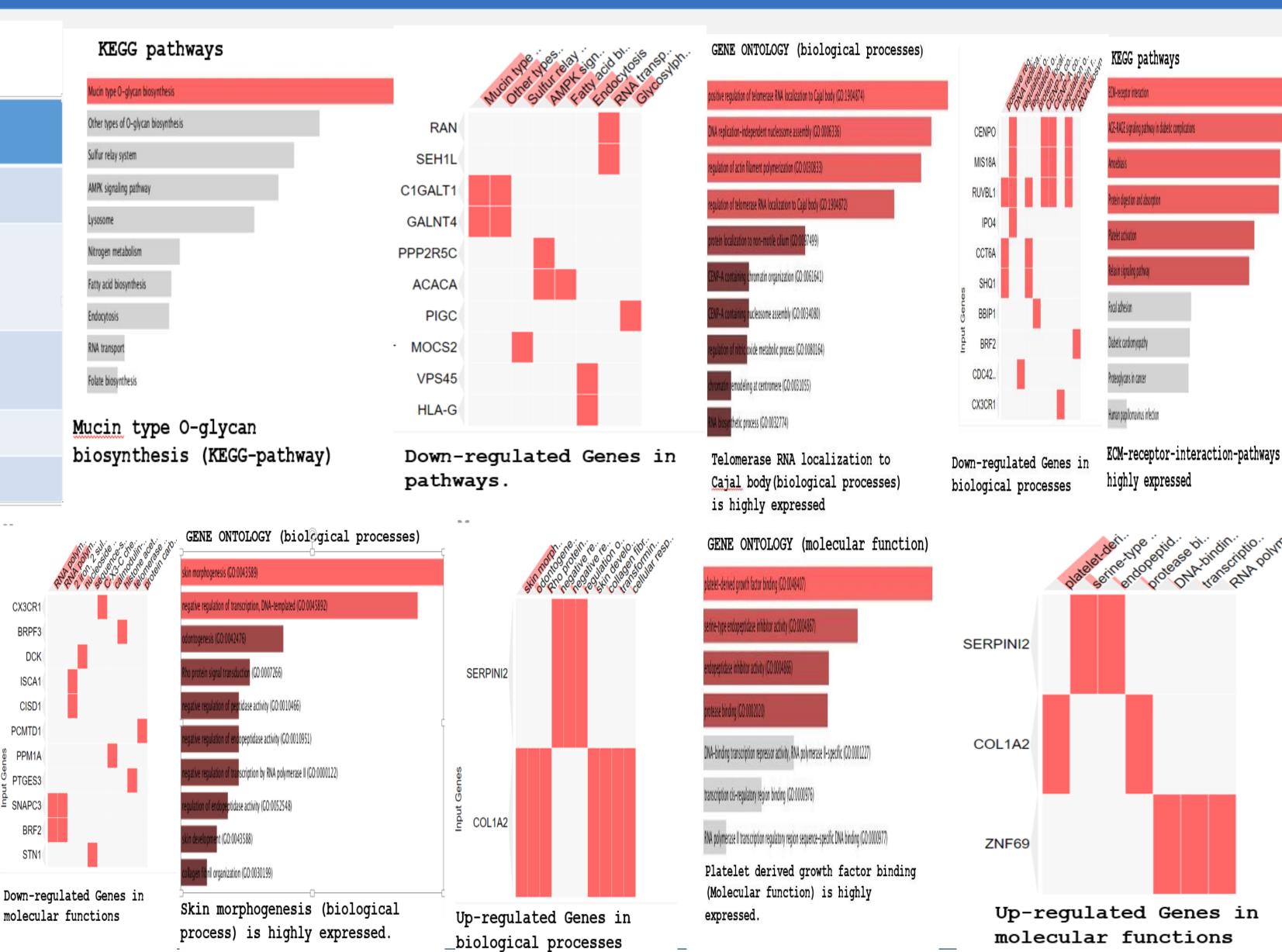
	r	
l	TOTAL GENES	913
((GENES WITH P-VALUE LESS THAN 0.05 (DEGs)	555
	GENES WITH P-adj VALUE LESS THAN 0.05 (significantly expressed genes)	262
	GENES WITH log2Fold change >1.2 or <-1.2	165
	DOWN-REGULATED GENES	158
	UP-REGULATED GENES	7

GENE ONTOLOGY (molecular function)

(
-

region sequence-specific DNA binding (molecular function) is highly expressed





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OMICS RESEARCH SYMPOSIUM

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Methods

- ✓ The data for this project was obtained from the Gene Expression Omnibus (GEO) database after a thorough search, which is a part of NCBI (National Center for Biotechnology Information). ✓ The dataset of series: GSE161968 was selected as the ideal dataset for the study and analysis of the project and was
- associated with BioProject: PRJNA680181 (Zhu H et al, 2021). ✓ This dataset consisted of microarray data (mRNA) of SCLC patients with Brain metastases and without Brain metastases. The samples of these patients were extracted from the peripheral blood (tissue). There were 12 samples in total, with 6 samples as BM positive and 6 samples as BM negative

Exploratory Analysis

- Exploratory analysis enabled the detection of the variance present among all the samples (patients with and without) brain metastases ✓ PCA was performed on all the 12 samples with the help of T-bioinfo server platform that provided with different
- pipelines. The pipeline first performed Quantile normalization of the data (as the data obtained from dataset was already normalized) and then it performed the PCA.
- ✓ As an output result, file that contained information about the total variance in data with respect to PC components: PC1 (representing maximum variance) and PC2 (representing second maximum variance).

Data processing

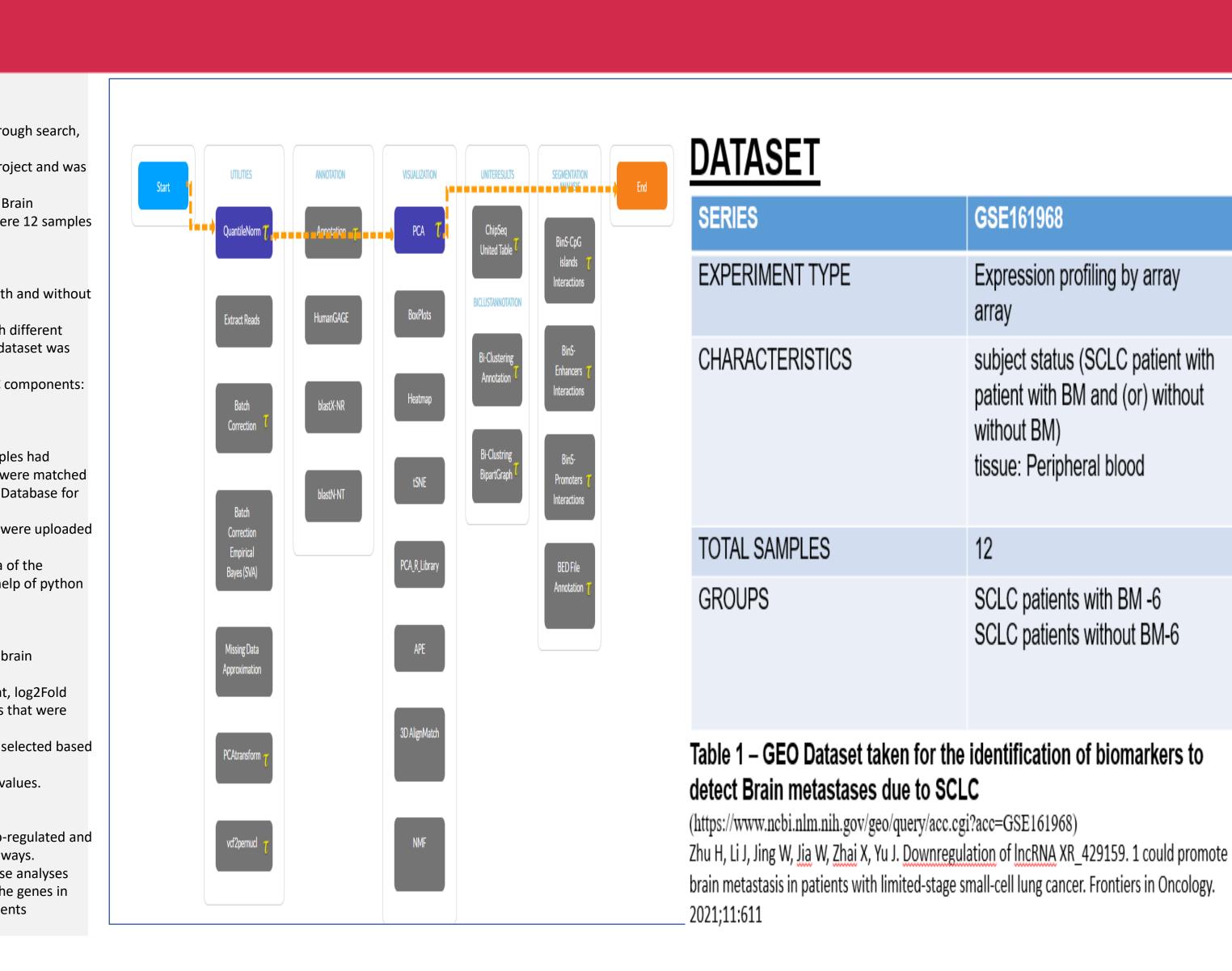
- ✓ The dataset consisted of microarray data. As a result of which genes (present as gene symbols) in samples had multiple probe ids. In order to perform comparative and statistical analysis on the data, the probe ids were matched to gene ids (to avoid and remove duplicate genes (if present)). This process was performed on DAVID (Database for Annotation, Visualization and Integrated Discovery) database. ✓ The probe ids were extracted from the supplementary file, which was obtained from the dataset, and were uploaded
- in DAVID database to obtain gene symbols. From the DAVID database, the gene symbols were obtained that matched with the probe ids. The data of the samples of individual probe ids corresponding to gene symbols were mapped to each other with the help of python

Statistical analysis:

- ✓ To identify differentially expressed genes between samples with brain metastasis vs. samples without brain metastasis, different parametric tests were performed in excel.
- ✓ T-test was performed to evaluate p-value which signifies significant genes present in sample. After that, log2Fold change and P-adj (Q value) were calculated in order to obtain significant differentially expressed genes that were present in samples.
- ✓ The output data was filtered with p-value <0.05 (threshold) and log2Fold change: >-1.2 or <1.2 (range selected based on the number of genes, varies from sample to sample).
- ✓ From this data, the up-regulated and down-regulated genes were identified based on logfold change values.

Gene Enrichment Analysis:

✓ To understand and analyse the biological implications of identified significant genes obtained (both up-regulated and down-regulated) as a result of statistical analysis, we performed enrichment analyses in different pathways. ✓ Towards this, the genes were uploaded (in form of list) on ENRICHR database (reference). This database analyses and identifies the role of genes in different pathways, provides information about the enrichment of the genes in different Geno ontology (GO) terms, i.e. biological processes, molecular functions and cellular components



Conclusions

The results obtained after analysing data depicted that the genes: COL1A2, C1GALT1, GALNT4, .RUVBL1, CCT6A, SHQ1, SNAPC3, **BRF2** might act as effective biomarkers.

Main pathways (KEGG) that were involved include:

- ✓ Mucin type O-glycan biosynthesis pathway (highly supressed due to down-regulation of genes)
- ECM-receptor-interaction-pathways (highly) activated due to up-regulation of genes)

Molecular functions that resulted in metastatic conditions were:

- \checkmark RNA polymerase III cis-regulatory region sequence-specific DNA binding function (highly supressed due to down-regulation of genes)
- \checkmark Platelet derived growth factor binding (PDGF) (highly activated due to upregulation of genes)

Biological processes involved:

- ✓ Telomerase RNA localization to Cajal body (highly supressed due to down-regulation of genes)
- ✓ skin-morphogenesis (highly activated due to up-regulation of genes)

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GSE161968

- Expression profiling by array array
- subject status (SCLC patient with patient with BM and (or) without without BM) tissue: Peripheral blood

12

SCLC patients with BM -6 SCLC patients without BM-6

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Video link- (156) Bioinformatics approach to understand transcriptional changes in SCLC patients with brain metastases - YouTube

