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<u>Q1.</u>

Geo Accession Number: GSE14520

<u>Title</u>: Gene expression data of human hepatocellular carcinoma (HCC)

<u>Platform</u>: GPL3921 [HT_HG-U133A] Affymetrix HT Human Genome U133A Array

<u>Q2.</u>

Printing the summary statistics for the first five probe ids (can be changed as per requirement in the code)

Statistic	N	Mean	St. Dev.	Min	Max
probe_data_list	445	6.887	0.765	5.416	10.172

Statistic	N	Mean	St. Dev.	Min	Max
probe_data_list	445	4.410	0.516	3.211	7.164

[1] "117_at"

Statistic	N	Mean	St. Dev.	Min	Max
probe_data_list	445	3.988	0.453	3.312	8.444

[1] "121_at"

Statistic	N	Mean	St. Dev.	Min	Max
probe_data_list	445	5.711	0.328	4.861	6.607

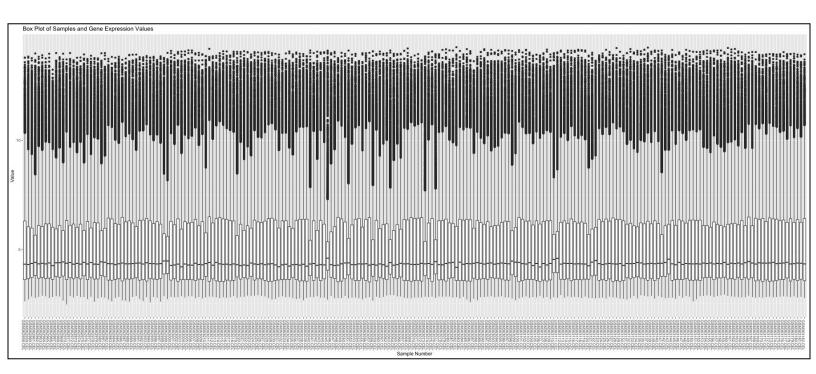
[1] "1255<u>g</u>at"

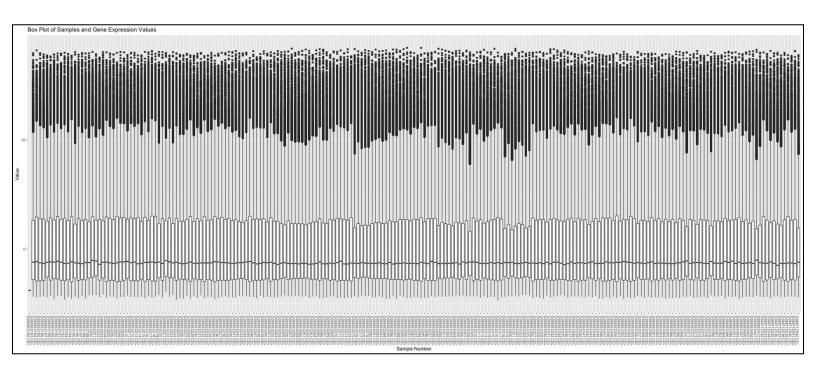
Statistic	N	Mean	St. Dev.	Min	Max
probe_data_list	445	3.197	0.120	2.927	3.682

Box Plot:

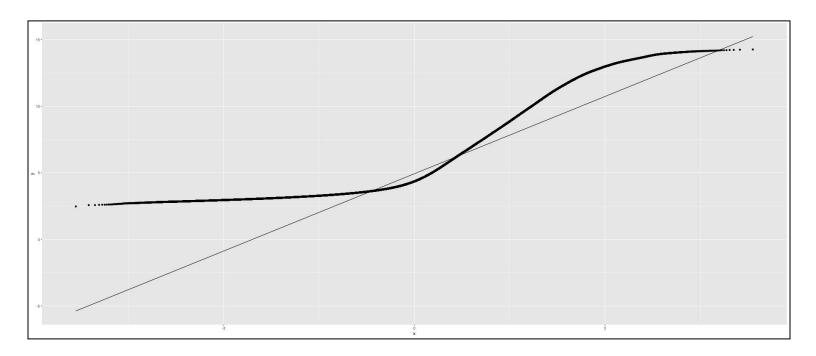
(Split in two for ease of reading)

Plotted the gene values of each sample as a box, so we have 450 boxes (225 boxes in each plot).





QQ Plot:



Skewness of all gene data = 1.33152 Kurtosis of all gene data = 4.287771

Q3.

Since the microarray data is already normalised, we do not need to perform normalisation again.

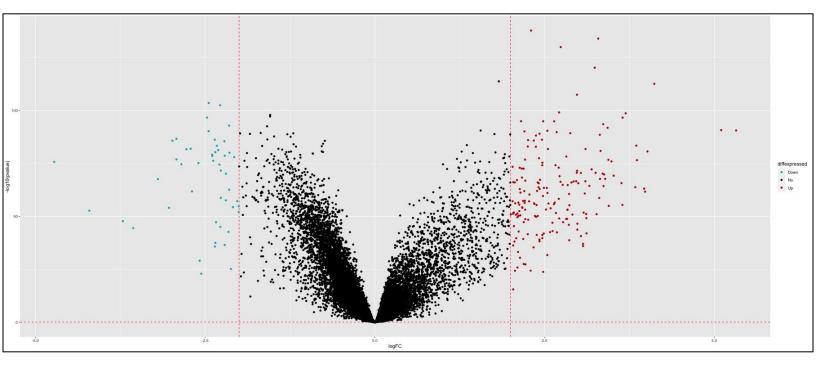
However, if we were to get completely unprocessed data, taking the log transform would transform skew data to approximately conform to normality. The range of the values would change as well, the gene expression after log transform lies between 0 to 16.

If we were to take the log transform of already log transformed data (which is the case with us) the shape of the data would remain approximately the normally distributed. Only the range of values would change from 0 to 16, to, 0 to 4. It also makes data more comparable.

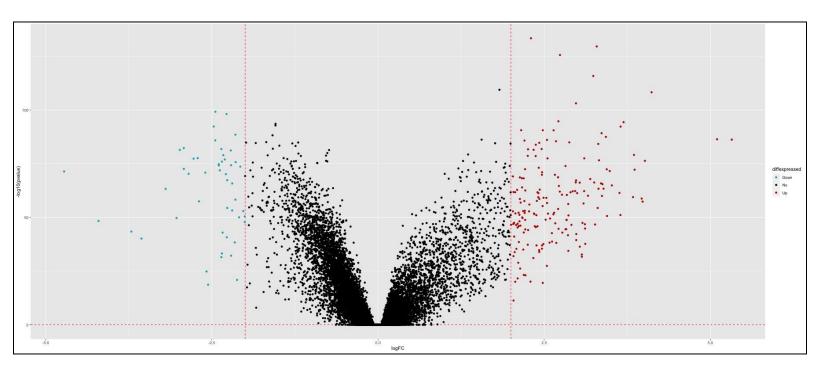
<u>Note</u>: I have not performed log transformation in the code (since data is already log transformed), however, I have written the code that would perform the log transformation (commented it out).

<u>Q4.</u>

Volcano Plot:



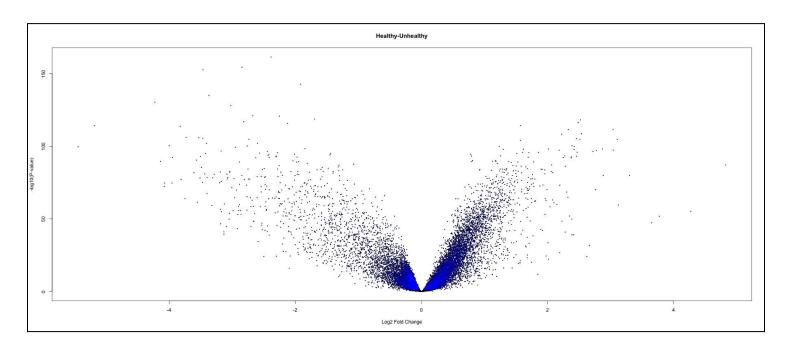
(Without Holm Correction, Up) (With Holm Correction, Down)



Number of Differentially Expressed Genes Obtained: 230

<u>Q5.</u>

Volcano Plot:



Number of Differentially Expressed Genes Obtained: 252

<u>Q6.</u>

log(FoldChange) cutoff = |2|p-value cutoff = 0.05

Normally we choose the fold change cutoff to be around 1 or 1.5, however I chose 2 as the cutoff since 1 and 1.5 were giving a large number of DEGs which did not seem correct to me. The difference between only a few genes is what leads to the difference between a person with cancer and a person without. The number of DEGs were fairly large even with a cutoff of 1.5, hence I chose 2 to be the cutoff value.

We generally we choose the p-value cutoff to be 0.05. Moreover, after Holm correction, a lot of p values were rounded up to 1. This meant that a good number of p-values had been adjusted for errors and therefore there was no need to change the cutoff to a lower value.

A p-value of 0.05 means that we expect an extreme statistic to show up in a chosen sample less than 5% of the time which is quite a low percentage anyway.

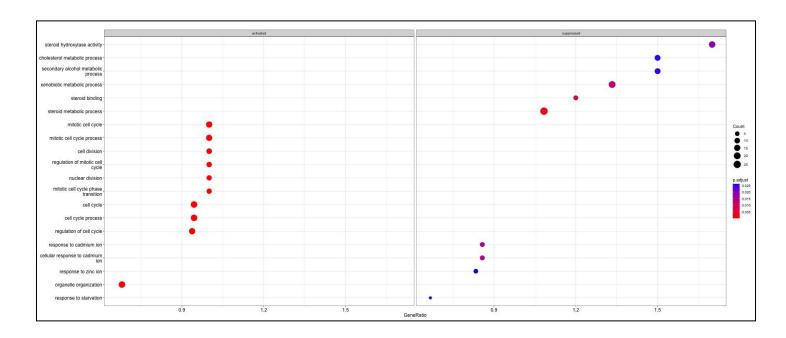
<u>Q7.</u>

Done in R Script

<u>Q8.</u>

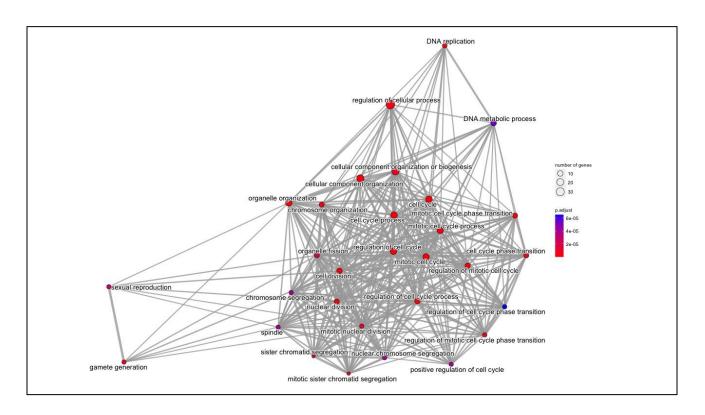
GO Analysis

1. Dot Plot



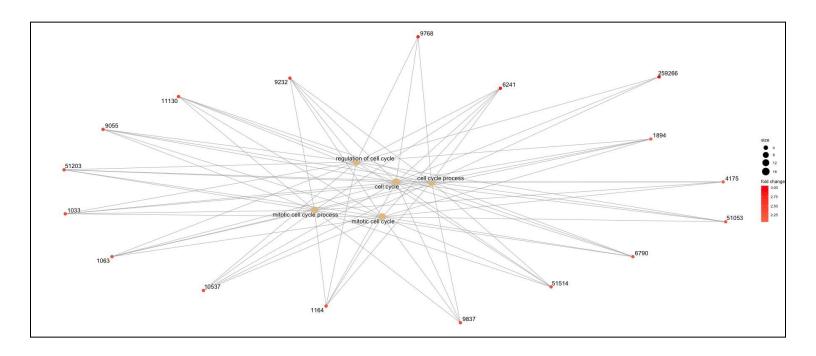
<u>Interpretation</u>: The colour of the dot indicates the p-value with which the particular gene is determined. The count indicates the numbers of genes involved in that particular process.

2. Enrichment Map



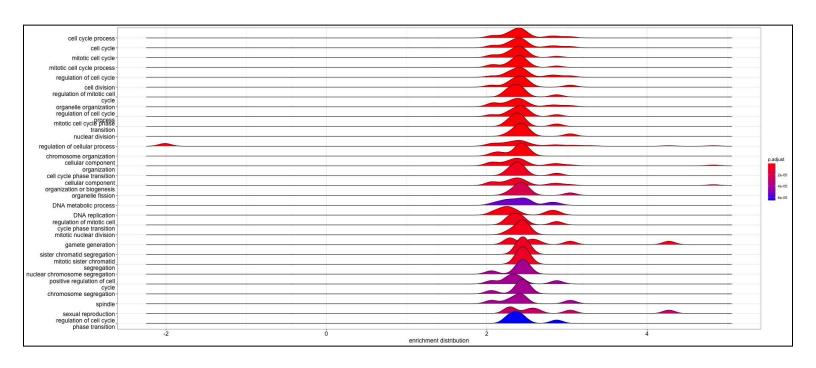
<u>Interpretation</u>: The map shows how functions are related to one another. The colour of the dots indicates the p-value with which that particular process is determined. The number of genes responsible for a particular process is indicated by the size of the dots.

3. Category Netplot



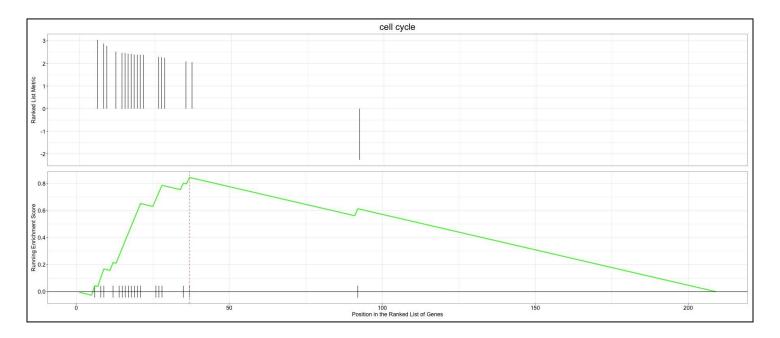
<u>Interpretation</u>: The map shows how a particular gene is related to the above biological processes. The gene is identified by its ENTREZ ID. The colour of the gene indicates the calculated fold change. The size of the dot indicates the numbers of genes responsible for that particular process.

4. Ridge Plot



<u>Interpretation</u>: The above plot shows how functions are correlated with the genes. If the curve is to the right of 0, then the gene list is positively correlated to that particular function. Likewise, if the curve is to the left of 0 then the gene list is negatively correlated to that particular function.

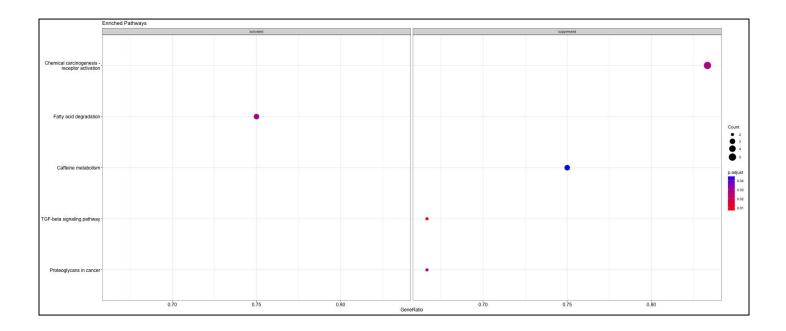
5. GSEA Plot



<u>Interpretation</u>: The above plot shows us the enrichment value (the peak of the curve is the enrichment value) for a particular process (cell cycle in our case). The higher the enrichment value, the more enriched the pathway/function is and vice versa.

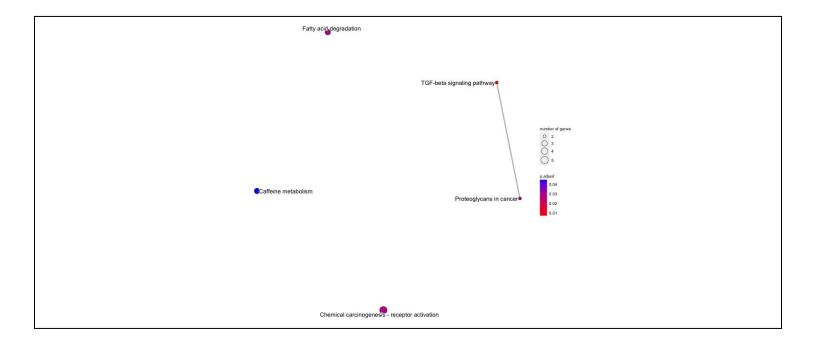
KEGG Analysis

1. Dot Plot



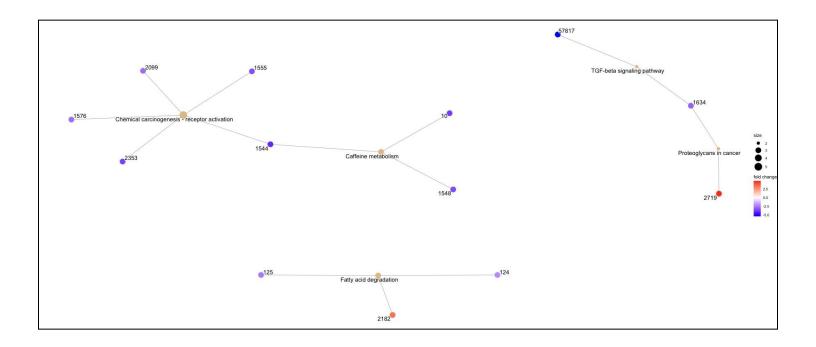
<u>Interpretation</u>: The colour of the dot indicates the p-value with which the particular gene is determined. The count indicates the numbers of genes involved in that particular process.

2. Enrichment Map



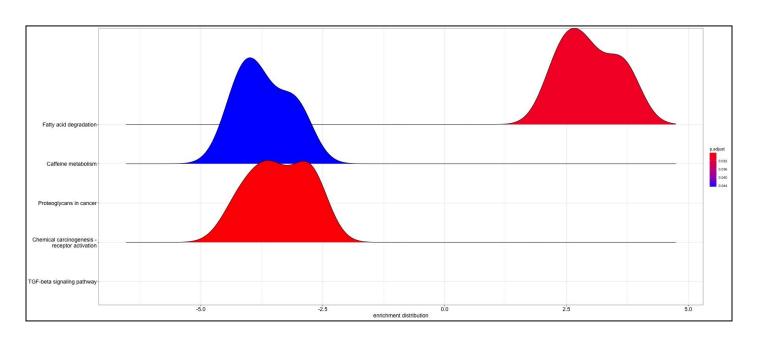
<u>Interpretation</u>: The map shows how functions are related to one another. The colour of the dots indicates the p-value with which that particular process is determined. The number of genes responsible for a particular process is indicated by the size of the dots.

3. Category Netplot



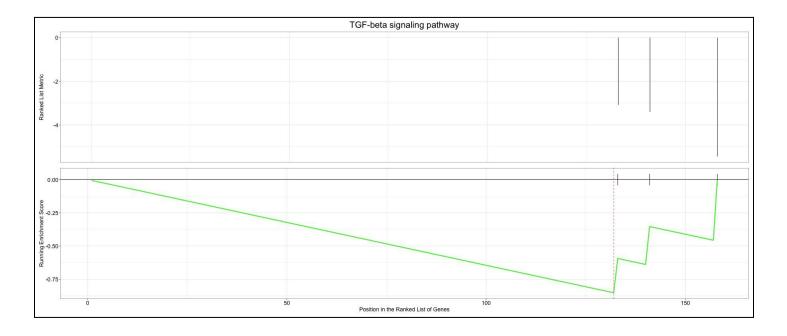
<u>Interpretation</u>: The map shows how a particular gene is related to the above biological processes. The gene is identified by its ENTREZ ID. The colour of the gene indicates the calculated fold change. The size of the dot indicates the numbers of genes responsible for that particular process.

4. Ridge Plot



<u>Interpretation</u>: The above plot shows how functions are correlated with the genes. If the curve is to the right of 0, then the gene list is positively correlated to that particular function. Likewise, if the curve is to the left of 0 then the gene list is negatively correlated to that particular function.

5. GSEA Plot



<u>Interpretation</u>: The above plot shows us the enrichment value (the valley of the curve is the enrichment value) for a particular pathway (TGF-beta signalling pathway in our case). The higher the enrichment value, the more enriched the pathway/function is and vice versa.

<u>Q9.</u>

Pathway: Steroid metabolic process

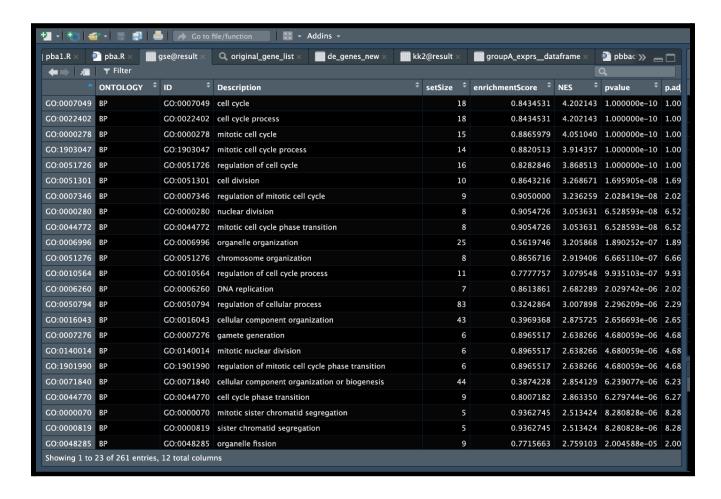
NES Score: -2.039797

Since the NES Score < 0, it means that the above pathway is downregulated by our set of genes. So the above signalling pathway is negatively affected, which is a possible side effect of cancer.

Pathway: Cell division **NES Score**: 3.268671

Since the NES Score > 0, it means that the above pathway is upregulated by our set of genes. So, degradation of fatty acids is higher, which is what happens in cancer (uncontrolled growth of cells).

The same can be done for the rest of the pathways (available in gse@result).



Pathway: TGF-beta signaling pathway

NES Score: -1.789614

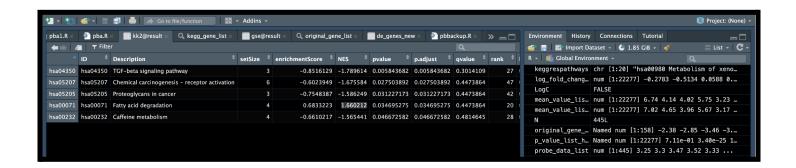
Since the NES Score < 0, it means that the above pathway is downregulated by our set of genes. So the above signalling pathway is negatively affected.

Pathway: Fatty acid degradation

NES Score: 1.660212

Since the NES Score > 0, it means that the above pathway is upregulated by our set of genes. So, degradation of fatty acids is higher.

The same can be done for the rest of the pathways (available in kk2@result).



References:

- 1. https://www.ncbi.nlm.nih.gov/geo/geo2r/?acc=GSE14520&platform=GPL3921
- 2. https://learn.gencore.bio.nyu.edu/rna-seq-analysis/gene-set-enrichment-analysis/