
Last updated: Apr 23, 2020

Trev-seq data analysis pipeline (Original one from October-November):

- 000_Summary data type form Trevor Wilson at Hudson laboratory sin Melbourne

array
(sabre)

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- 02_cutadapt (we have contamination)
- 03_RNAseq
 - 03_STAR mapping mm10 genome and annotation
 - 04_Mapping analysis
- HTSEQ analysis on them....

000_Summary data type from Trevor Wilson at Hudson laboratories in Melbourne.

- HiSeq 3000
- Sample pooling 123 samples
- 8 Sets / Lanes of Multiplex RNA-seq project
- Up to 300M reads per lane
- "In-house developed multiplex method" by Trevor Wilson.
- R1:19 bp this is 20
- R2:76bp in reality, when checking this is 75

Service	Chemistry	Bar-coded	Description of Service and cost per sample	Qty
Multiplex RNAseq project		8 sets of multiplex barcoded samples	1. Sample QC 2. First strand synthesis with custom indexed primer 3. Amplification and library generation of pooled cDNA 4. Library QC Sample QC and First strand \$25 x 123	123
			Multiplexed library prep and QC \$550	8
QC of additional samples			Bioanalyzer RNA 6000 Pico chips \$90 per chip 47 chips run 25/7 – 17/9, less 12 allowed for above	35
	Illumina HiSeq3000 with custom R1 primer		1. Library Denaturation 2. ExAmp clustering 3. Custom primer annealing 4. HiSeq Sequencing (R1 19bp; R2 ~70bp) \$ 1,600 per lane for full 8-lane flowcell ~300million raw reads per lane*	8

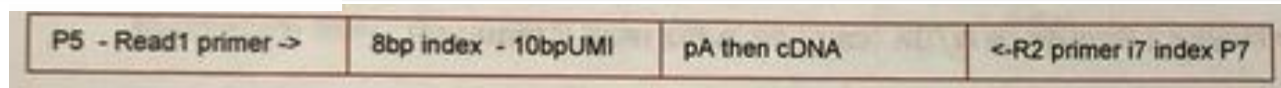
So already we know that if we put from 15-17 samples in a lane where we expect **~300 million raw reads per lane**

We expect between 15M-17M reads per sample! (For me it looks quite low, but talking with Christian he confirms that it is RNA poly A) and it is stem cells rather than tissue

So I guess it should be "posterior"

Also we know that the data has the following format:

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So it contains an 8bp index + a 10 bp UMI!

Be careful with these new concepts as I know barcoded for pool-seq and demultiplexing but never dealt with UMIs before

- Unique molecular identifiers (**UMIs**), or molecular barcodes (MBC) are short sequences or molecular "tags" added to DNA fragments in some next generation sequencing library preparation protocols to identify the input DNA molecule. They can be used to reduce errors and quantitative bias introduced by amplification.
- A **UMI** is then randomly assigned to each molecule and the pool of molecules amplified by duplicating each molecules for every in silico PCR cycle. The final pool of molecules **are** then "sequenced" to produce the reads.

###Number of counts of the original data:

Set or Lane	# Reads in R1	# Samples within set	# Of expected number of reads (+ 10 % of missing not included)	R1 bp	R2 bp
Set1	294,775,915	17	17,339,759.7	20	77
Set2	303,318,042	14	21,665,574.4	20	76
Set3	305,492,703	16	19,093,293.9	20	77
Set4	299,233,225	14	21,373,801.8	20	77
Set5	311,752,725	17	18,338,395.6	20	76
Set6	283,569,294	16	17,723,080.9	20	77
Set7	239,235,722	16	14,952,232.6	20	77
Set8	250,823,201	13	19,294,092.4	20	77

000_Project data sharing:

The data of this project has been backed up in

- **a) External hard-rive**
- **b) UQMDR:** QRISdata/Q1287/00_fastq_gz
- On Friday 1st of November we have a meeting about UQRDM!

"The samples are barcoded, and therefore, demultiplexing is required"

You need to know the exact number of samples that you are demultiplexing. If you have 15 samples, you need to have 15 cellular barcodes similar to those shown above. In Trev-seq cellular barcodes are 8 bases long and UMIs are typically 10 bases long. There is an additional one base in R1 (can't explain this particular well) therefore R1 read has to be 19 bases long. Have a look inside FASTQ file to see that this is the case. Your R2 can be anywhere up to 150 bases (standard illumina length), but I think Trevor normally does 75 cycles - length.

The tool Kirill mentioned is sabre:

<https://github.com/najoshi/sabre>

Sabre is a tool that will demultiplex barcoded reads into separate files. It will work on both single-end and paired-end data in fastq format. It simply compares the provided barcodes with each read and separates the read into its appropriate barcode file, after stripping the barcode from the read (and also stripping the quality values of the barcode bases). If a read does not have a recognized barcode, then it is put into the unknown file.

Sabre also has an option (-m) to allow mismatches of the barcodes.

Finally, after demultiplexing, sabre outputs a summary of how many records went into each barcode file.

Be aware that if you do not format the barcode data file correctly, sabre will not work properly.

The one he points out is: 1. demultiplex with sabre tool <https://github.com/serine/sabre> to get your 15 individual fastq files, all single ended

It should be sent in PBS job scheduling system used on Awoonga!

```
uqmnava1@awoonga1:/QRISdata/Q1287/00_fasq_gz> /home/uqmnava1/software/sabre/src/sabre
-f Set1_S1_L001_R1_001.fastq.gz -r Set1_S1_L001_R2_001.fastq.gz -b
Set1_barcode.txt -c -u -m2 -l 10 -a 1 -s Set1_sabre.txt
```

Running: sabre

Command line args:

```
--fq1 Set1_S1_L001_R1_001.fastq.gz
--fq2 Set1_S1_L001_R2_001.fastq.gz
--barcode Set1_barcode.txt
--unassigned_R1 unassigned_R1.fq
--unassigned_R2 unassigned_R2.fq
--combine 1
--umi 1
--max-mismatch 2
--min-umi-len 10 ###minimum UMI length to keep
--max-5prime-crop 1
--no-comment -1
--stats Set1_sabre.txt
```

```
BC TAAGGCGA
BC CGTACTAG
BC AGGCAGAA
BC TCCTGAGC
BC GGACTCCT
BC TAGGCATG
BC CTCTCTAC
BC CGAGGCTG
BC AAGAGGCA
BC GTAGAGGA
BC GCTCATGA
BC ATCTCAGG
BC ACTCGCTA
BC GGAGCTAC
BC GCGTAGTA
BC CGGAGCCT
BC TACGCTGC
```

Recommended not working on QRIS directory!

Directory: /30days/uqmnava1/RNAseq/00_fasq_gz

```
cat 00_demultiplex_set1.pbs
#!/bin/bash
#PBS -A UQ-IMB
#PBS -l walltime=02:00:00
#PBS -l select=1:ncpus=10:mem=20GB
#cd /QRISdata/Q1287/00_fasq_gz/
cd /30days/uqmnava1/RNAseq/00_fasq_gz/

/home/uqmnava1/software/sabre/src/sabre -f Set1_S1_L001_R1_001.fastq.gz -r
Set1_S1_L001_R2_001.fastq.gz -b Set1_barcode.txt -c -u -m2 -l 10 -a 1 -s
Set1_sabre.txt -t 8
mv unassigned_R2.fq Set1_unassigned_R2.fq
mv unassigned_R1.fq Set1_unassigned_R1.fq
/home/uqmnava1/software/sabre/src/sabre -f Set2_S2_L002_R1_001.fastq.gz -r
Set2_S2_L002_R2_001.fastq.gz -b Set2_barcode.txt -c -u -m2 -l 10 -a 1 -s
Set2_sabre.txt -t 8
mv unassigned_R2.fq Set2_unassigned_R2.fq
mv unassigned_R1.fq Set2_unassigned_R1.fq
/home/uqmnava1/software/sabre/src/sabre -f Set3_S3_L003_R1_001.fastq.gz -r
Set3_S3_L003_R2_001.fastq.gz -b Set3_barcode.txt -c -u -m2 -l 10 -a 1 -s
Set3_sabre.txt -t 8
mv unassigned_R2.fq Set3_unassigned_R2.fq
mv unassigned_R1.fq Set3_unassigned_R1.fq
/home/uqmnava1/software/sabre/src/sabre -f Set4_S4_L004_R1_001.fastq.gz -r
Set4_S4_L004_R2_001.fastq.gz -b Set4_barcode.txt -c -u -m2 -l 10 -a 1 -s
Set4_sabre.txt -t 8
mv unassigned_R2.fq Set4_unassigned_R2.fq
mv unassigned_R1.fq Set4_unassigned_R1.fq
/home/uqmnava1/software/sabre/src/sabre -f Set5_S5_L005_R1_001.fastq.gz -r
Set5_S5_L005_R2_001.fastq.gz -b Set5_barcode.txt -c -u -m2 -l 10 -a 1 -s
Set5_sabre.txt -t 8
mv unassigned_R2.fq Set5_unassigned_R2.fq
mv unassigned_R1.fq Set5_unassigned_R1.fq
```

```

#PBS -l select=1:ncpus=10:mem=20GB

cd /30days/uqmnaval/RNAseq/00_fasq_gz/

/home/uqmnaval/software/sabre/src/sabre -f Set6_S6_L006_R1_001.fastq.gz -r
Set6_S6_L006_R2_001.fastq.gz -b Set6_barcode.txt -c -u -m2 -l 10 -a 1 -s
Set6_sabre.txt -t 8
mv unassigned_R2.fq Set6_unassigned_R2.fq
mv unassigned_R1.fq Set6_unassigned_R1.fq
/home/uqmnaval/software/sabre/src/sabre -f Set7_S7_L007_R1_001.fastq.gz -r
Set7_S7_L007_R2_001.fastq.gz -b Set7_barcode.txt -c -u -m2 -l 10 -a 1 -s
Set7_sabre.txt -t 8
mv unassigned_R2.fq Set7_unassigned_R2.fq
mv unassigned_R1.fq Set7_unassigned_R1.fq
/home/uqmnaval/software/sabre/src/sabre -f Set8_S8_L008_R1_001.fastq.gz -r
Set8_S8_L008_R2_001.fastq.gz -b Set8_barcode.txt -c -u -m2 -l 10 -a 1 -s
Set8_sabre.txt -t 8
mv unassigned_R2.fq Set8_unassigned_R2.fq
mv unassigned_R1.fq Set8_unassigned_R1.fq

(rnasik-1.5.4) uqmnaval@awoonga1:~/30days/uqmnaval/RNAseq/00_fasq_gz>

```

After demultiplexing this is 77bp!

All data then is outputted and we have 148 *fastq files... some have 0Gb that is because they are merged with other datatypes:

We now check per sample and barcode

#Check on the distinct barcodes...

The different

First of all understanding the output in Set1!

#Number of reads in Set1:

```

echo $(( $(zcat Set1_S1_L001_R1_001.fastq.gz | wc -l) / 4 ))
294775915 ### That matches with Hudson's report "Lane 1 total reads passed filter
294.8 Million; Set 1 i7 index TAAGGCGA"
echo $(( $(zcat Set1_S1_L001_R2_001.fastq.gz | wc -l) / 4 ))
294775915

#so as SE: 294775915*2 = 589551830

```

Looking at Sabre output... The Total of reads is inferior 293,571,193 PE reads... that is 587,142,386 SE reads (total number of reads R1+R2)

Set1_m0_sabre.txt

N_pairs

4520189

8240889

P_pairs

0.02

0.03

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GCGTAGTA	12632956	6316478	0.02
GGAGCTAC	13941528	6970764	0.02
ACTCGCTA	12662261	6331130	0.02
ATCTCAGG	15157226	7578613	0.03
GCTCATGA	77588366	38794183	0.13
GTAGAGGA	16244670	8122335	0.03
AAGAGGCA	13848293	6924146	0.02
CGAGGCTG	20684602	10342301	0.04
CTCTCTAC	10809287	5404643	0.02
TAGGCATG	5921828	2960914	0.01
GGACTCCT	7377530	3688765	0.01
TCCTGAGC	5002548	2501274	0.01
AGGCAGAA	8860252	4430126	0.02
CGTACTAG	6979547	3489773	0.01
TAAGGCGA	5792122	2896061	0.01
unassigned	71239682	35619841	0.12
total	587142386	293571193	1.00

###These numbers are lower than the original number of reads : 294775915-293571193=1,204,722 ##One million reads from the original input to sabre are discarded.

FILE output file Lane 1/ Set 1:

Tissue	Cell Type	Y/O	ULN	Index	Index sequer sample	New demultiplexed file	#wcc -i fastq	#Number of reads	PCT Total reads	PE
Liver	CD133+	Y	19-04923	N701	TAAGGCGA 2407-1	2407-1_TAAGGCGA_R1.fastq		23750988	5937747	0.02
	CD133+	O	19-04924	N702	CGTACTAG 2407-2	2407-2_CGTACTAG_R1.fastq		29159404	7289851	0.02
	CD133+	Y	19-04925	N703	AGGCAGAA 2407-3	2407-3_AGGCAGAA_R1.fastq		37510196	9377549	0.03
	CD133+	O	19-04926	N704	TCCTGAGC 2407-4	2407-4_TCCTGAGC_R1.fastq		20771156	5192789	0.02
	CD133+	Y	19-04927	N705	GGACTCCT 2407-5	2407-5_GGACTCCT_R1.fastq		30632836	7658209	0.03
	CD133+	O	19-04928	N706	TAGGCATG 2407-6	2407-6_TAGGCATG_R1.fastq		24682832	6170708	0.02
	mESC CS	CONTROL	19-04929	N707	CTCTCTAC 2407-7	2407-7_CTCTCTAC_R1.fastq		44739080	11184770	0.04
Pancreas	Beta Cells	Y	19-04930	N710	CGAGGCTG 0708-21	0708-21_CGAGGCTG_R1.fastq		58929604	14732401	0.05
	Beta Cells	O	19-04931	N711	AAGAGGCA 0708-22	0708-22_AAGAGGCA_R1.fastq		57104488	14276122	0.05
	Beta Cells	Y	19-04932	N712	GTAGAGGA 0708-23	0708-23_GTAGAGGA_R1.fastq		70464052	17616013	0.06
	Beta Cells	O	19-04933	N714	GCTCATGA 0708-24	0708-24_GCTCATGA_R1.fastq		340609380	85152345	0.29
	Beta Cells	Y	19-04934	N715	ATCTCAGG 0708-25	0708-25_ATCTCAGG_R1.fastq		61829044	15457261	0.05
	Beta Cells	Y	19-04935	N716	ACTCGCTA 0708-26	0708-26_ACTCGCTA_R1.fastq		52096944	13024236	0.04
	Beta Cells	O	19-04936	N718	GGAGCTAC 0708-27	0708-27_GGAGCTAC_R1.fastq		58565596	14641399	0.05
	Beta Cells	Y	19-04937	N719	GCGTAGTA 0708-28	0708-28_GCGTAGTA_R1.fastq		53046780	13261695	0.04
	Beta Cells	O	19-04938	N720	CGGAGCCT 0708-29	0708-29_CGGAGCCT_R1.fastq		68076040	17019010	0.06
	mESC CS	Control	19-04939	N721	TACGCTGC 0708-30	0708-30_TACGCTGC_R1.fastq		66253264	16563316	0.06
						Set1_unassigned_R1.fq		88881976	20220494	0.07
						Total Sum		1179103660	294775915	1.00
						Total sum (samples)		1098221684	274555421	0.93
						total Average (samples)		64601275.53	16150318.88	0.05
						Total median (samples)		53046780	13261695	0.04
						Original number SE reads R1+R2			589551830	
						Original number PE reads			294775915	

check EXCEL file:

Recommendations this far: re-run liver CD133 and Eye RP cohort 2 pooled! # reads less than 10M

- The number of unassigned reads is as Kirill mentioned < than 10% of reads..
- The number of reads we got “single end this time is between 250-300 Million rather than 400-500Millon
- The average number of reads per sample is 15 Million then with some fewer than 10!
- Kirill mentioned the aim was to have between 20-25 Million. This is not eh case in here! (The rule of thumb for RNA-seq gen expression is ~ 20-25K reads per sample. And I'm pretty sure this is what Trevor would aim for when doing the libraries.)

01_Data quality with fastqc

Well we had a general data fastqc report from Trevor and I am not sure if I am convinced on data. Usually at the

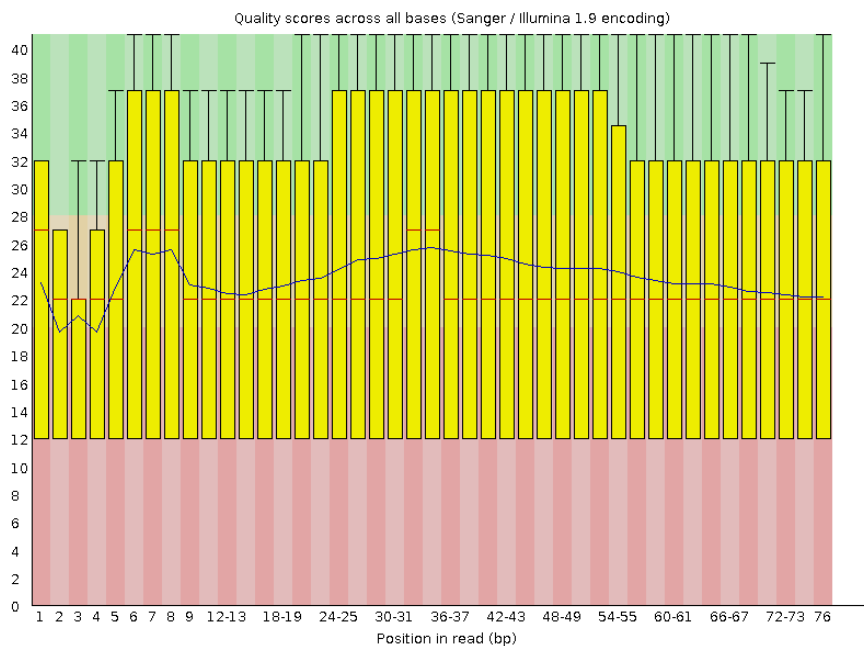
< 10 have a Phred score above 30 and almost close to 40!

Quality scores on the "red" area ==> danger!

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We know that Phred Score is the same as probability of having an error.. I

! Per base sequence quality



When performing FastQC on the data we observe that we have “contamination of primers...” Therefore, before mapping they should be removed.

In particular, there is contamination of the following primer 19% of reads + 10% of reads! They have Cutadapt installed and working

Module load cutadapt (version cutadapt/1.18)

```
uqmnava1@tinaroo2:~/uqmnava1/RNAseq/01_demultiplexed> cutadapt -b
GCAGCGTAGTACTCTGCGTTGATACCACTGCTTCCGCGGACAGGCGTGTA -o 2607-9_clean_test.fastq 2607-
9_TACGCTGC_R1.fastq
This is cutadapt 1.18 with Python 3.6.8
Command line parameters: -b GCAGCGTAGTACTCTGCGTTGATACCACTGCTTCCGCGGACAGGCGTGTA -o
2607-9_clean_test.fastq 2607-9_TACGCTGC_R1.fastq
Processing reads on 1 core in single-end mode ...
```

Cutadapt all options... so let's clean the adapters first on them!

- I should, I do have that information from Christian in Slack where he inputted all the info on how to perform the library preparations...

And let's compare our dataset with ENCODE current standards for small RNA-seq pipeline.

Check on the ENCODE quality standards:

<https://www.encodeproject.org/data-standards/rn>

Current Standards

experiments can be found here.

RNA-seq assay in which the average library insert size is <200 base pairs.

a replicates. Assays performed using EN-TEs samples may be exempted due to limited avail

aligned reads, although older projects aimed for 10 million aligned reads. Best practices for ENCODE2 RNA-seq experiments have been

outlined here.

- Replicate concordance: the gene level quantification should have a [Spearman correlation](#) of >0.9 between isogenic replicates and >0.8 between anisogenic replicates (i.e. replicates from different donors).
- The experiment must pass routine metadata audits in order to be released.

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[a-seq/small-rnas/](#)

01_ Mapping to mm10 with STAR aligner / RNAsik in Awoonga / Tinaroo HPC

Next step is mapping against genome....

We have lots of reads being thrown out..... why? That is weird.. I have never seen sth like that that.... Too short... why?

not sure....

#Some issues with too short alignments.... % of reads unmapped: too short | 24.87%

So I think in our case

We need to lower the trimming value or increase the trimming value...

Because if we have contamination of an adapter to 26bp

I changed the parameters to 0.3

And by default the

--outFilterMatchNmin

default: 0

int: alignment will be output only if the number of matched bases is higher than or equal to this value.

--outFilterMatchNminOverLread

default: 0.66

real: same as outFilterMatchNmin, but normalized to the read length

That is minimum 50 bp, (75 * 0.66 = 49.5bp) but if we have already 75- 26 bp from the Clontech SMART CDS Primer II A (100% over 26bp) that is already 49!!!! So it is already inferior that what expected

I think is just that.... But on the other hand is pointing at quite a bit of contamination in the reads... almost 26bp-30bp of read are with adapter sequences

reply. Do not hesitate to bump up your posts if I do not answer

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reMinOverLread 0 --outFilterMatchNminOverLread 0. This parameter only limits on the mapped length, so even if short portions of the reads align, the reads will be considered mapped. This explains why you are seeing high mapping rate - albeit such short alignments are often multimappers. I do not think this is generally a good idea, as such short alignments will have a high rate of wrong alignments. It is better to figure out why the mappability is poor, e.g.

1. File formatting issues
2. Poor sequencing quality
3. Contamination

I also did what Kirill was mentioning that is to look at the database of unmapped reads and check out what's in there....

```
uqmnaval@tinaroo2:~/uqmnaval/RNAseq/01_demultiplexed> head -21 t
21384 AAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
326 GACGTCGAGTACTCTGCGTTGATACCACTGCTTCCGCGGACAGGCGTGTAGATCTCGGTGGTCGCCGTATCATTAA
314 ATCTTTAACTCTAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
290 CGTCGAGTACTCTGCGTTGATACCACTGCTTCCGCGGACAGGCGTGTAGATCTCGGTGGTCGCCGTATCATTAAA
239 CTTATATATTGAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
224 TTATATACCTATGAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
215 NAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
203 CATATAATAATAAAAAAAAAACAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
197 ACGTCGAGTACTCTGCGTTGATACCACTGCTTCCGCGGACAGGCGTGTAGATCTCGGTGGTCGCCGTATCATTAA
159 ACAGTGATACATACAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
140 TATTATAACAGTAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
126 CAGCTACAGTGTACAAACAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
125 GTATATACACCTCAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
125 GTATATACCCCTCAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
113 CTGTTAAATAAAGCAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
111 CAGCTACAGTGTACAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
107 CATATTGATCAAGCAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
105 TTTGTGGACTGTGAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
103 AAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
99 TAATAATAATAAAACAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
97 CAGCTACAGTGTACTAACAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAAA
uqmnaval@tinaroo2:~/uqmnaval/RNAseq/01_demultiplexed> tail t
1 TTTTTTTTTTTTTTTTTTAACCGCACCGGGCAGCCGACGCGCGCGGCCCCCGCGCCGCTGGCTGGCTGGG
1 TTTTTTTTTTTTTTTTTTAGGGGTCAGCCAGCCACAAGACAACGGGCGGCCCCGCGCTGCCCTCCGGGCGTTTT
1 TTTTTTTTTTTTTTTTTTATTTTAAGAAAAAGAAAAACAAAACAGTACTTTTTTAATGGAACAACCTGACCAAAA
1 TTTTTTTTTTTTTTTTTTTTCGAAGACCGAAGCACACCCAAAGCAACAACCCACAAACCAAGCGACGGGGTCGG
1 TTTTTTTTTTTTTTTTTTTTAGTGTCAGGACGACGACCGAACCCCAACACTTCTGCGGGGGTGCCGCTGGGT
1 TTTTTTTTTTTTTTTTTTTTCAACAACCACCCCCCCCCCGCCCCACCCCCACGCGCCCCACGCGTGTGCA
1 TTTTTTTTTTTTTTTTTTTTCTGAGACAGCGCCGAGCGGGCTCCGGCCCCGCCCCGCGCCCTCCGTCTTGTTCG
1 TTTTTTTTTTTTTTTTTTTTGACAAACCAAGAAGCCAGTGCGGCACACACCTGGGTGACACCTTGTGGATGCA
1 TTTTTTTTTTTTTTTTTTTTAAATAGCGCGCCCTCCCGGCCGCGCTGCCTTTTCTCTGGTGCGTTGTCT
1 TTTTTTTTTTTTTTTTTTTTGATACCACTTCTTCCGCTGTCTGCCGTGTAGGTCTCGTTGGTTGGT
```

Mapping option: 01_demultiplexed/STAR_aling_03.pbs

```
#!/bin/bash
#PBS -A UQ-IMB
#PBS -N STAR
#PBS -l walltime=02:00:00
```

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Alternatively... using STAR we can use the transcriptome to check if the mapping is better as well as quantMode GeneCount

OK!

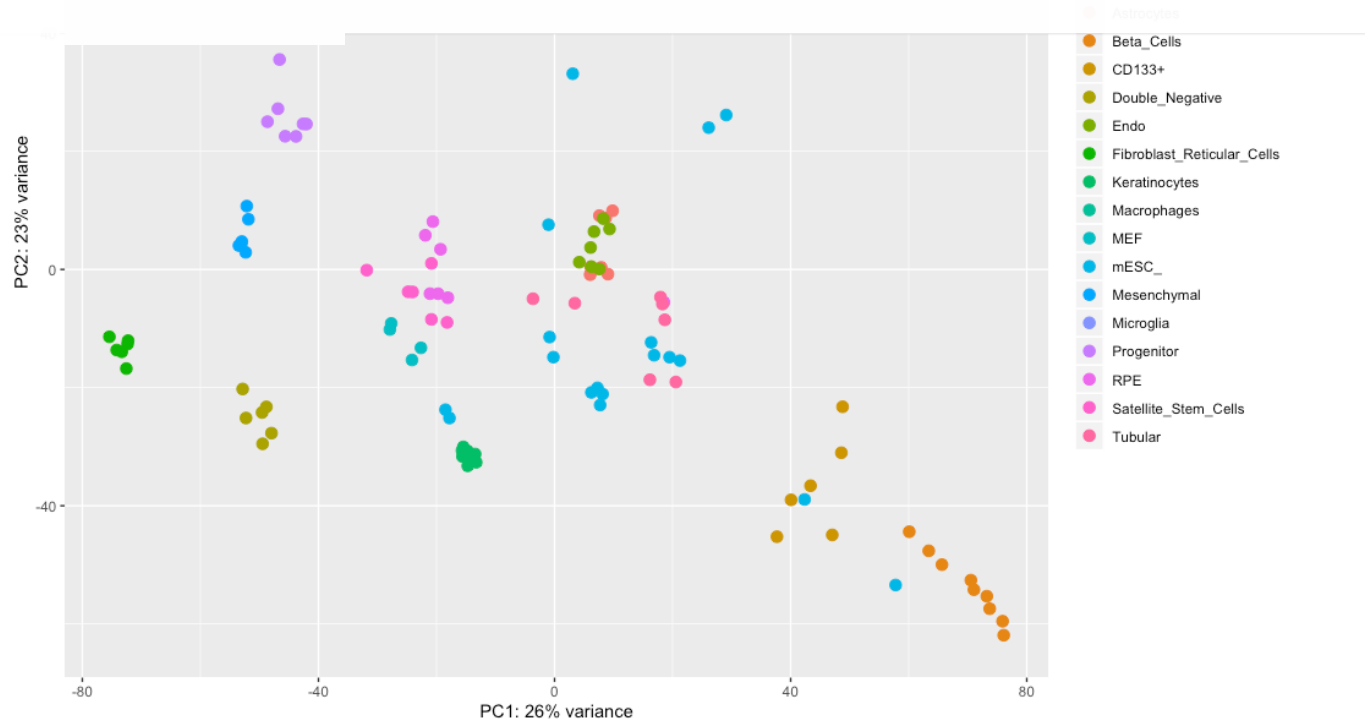
Now we go with featureCounts:
On them.....

However, shall we remove duplicates on them? Though PICARD? ... in RNA-seq data??

Also the counts could be performed with STAR!

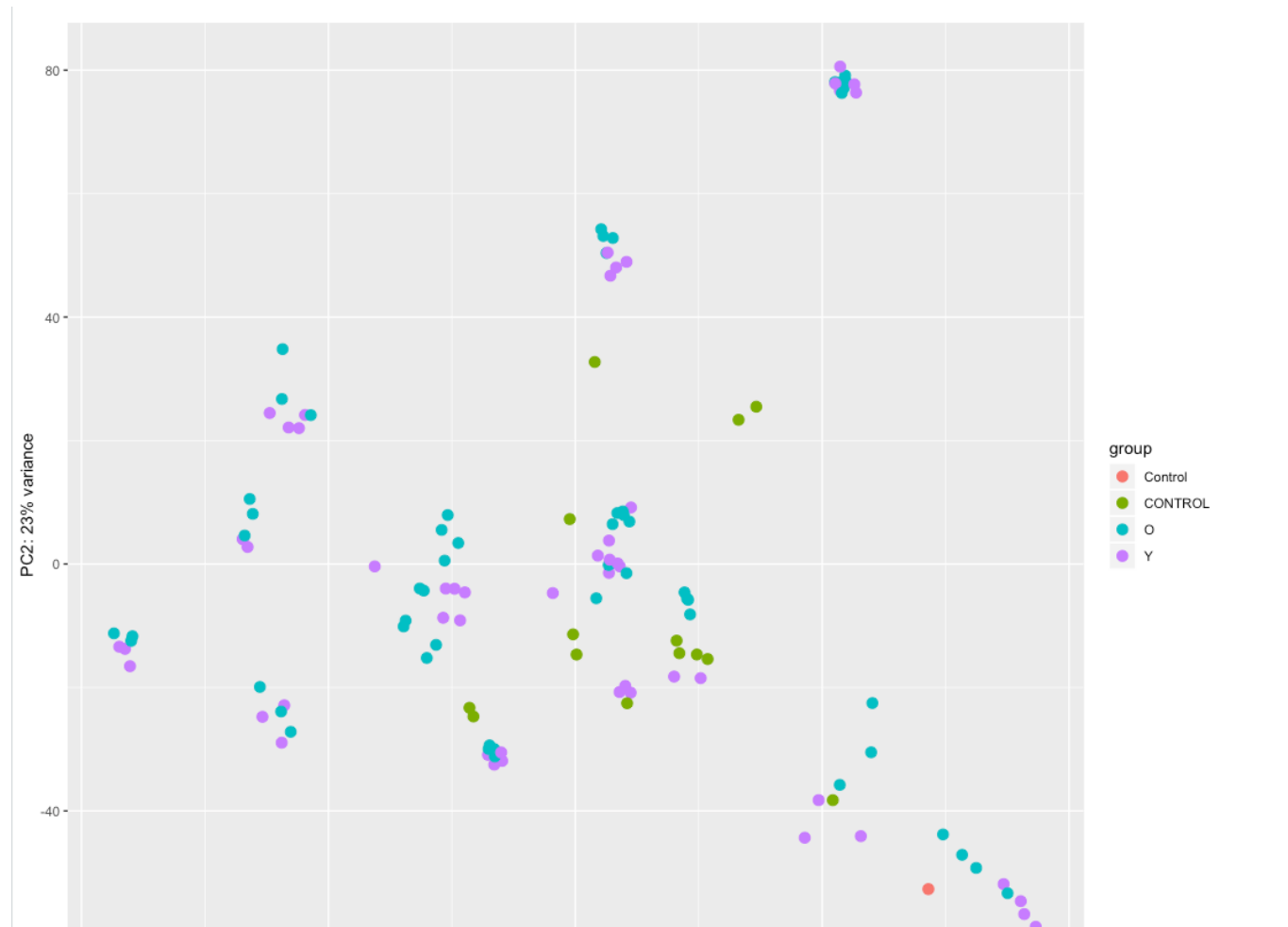
- All_counts.txt
- On the data.... Of them...
-

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##Check here the “batch effect on them”

```
plotPCA(vsd,intgroup='V3')
```





Now we perform on them

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