

Supplementary tables, figures and information for

A comprehensive Comparisons of RNA-Seq based transcriptome analysis from reads to differential gene expression and cross comparison with microarrays

Intawat Nookaew^{1,=}, Marta Papini^{1,=}, Natapol Pornputpong¹, Gionata Scalcinati¹, Linn Fagerberg², Matthias Uhlén^{2,3}, Jens Nielsen^{1,3,*}

¹ Novo Nordisk Foundation Center for Biosustainability, Department of Chemical and Biological Engineering, Chalmers University of Technology, Gothenburg, Sweden.

² Novo Nordisk Foundation Center for Biosustainability, Department of Biotechnology, Royal Institute of Technology, Stockholm, Sweden

³ Novo Nordisk Foundation Center for Biosustainability, Technical University of Denmark, DK-2970 Hørsholm, Denmark

= The authors equally contributed to this work

* Corresponding Author: nielsenj@chalmers.se

Table S1 Summary of RNA-seq data generated in this study

Sample	Number of paired reads (millions)	Number of nucleotides (GB)
Batch r1	5.73	1.15
Batch r2	7.62	1.52
Batch r3	5.57	1.11
Chmostat r1	4.03	0.81
Chmostat r2	6.75	1.35
Chmostat r3	6.16	1.23
All	35.85	7.17
Average	5.97	1.19
sd	1.21	0.24

Table S2 Summary of genetic variations (SNV and INDEL) between *S. cerevisiae* CENPK 113-7D and reference strain S288c. The numbers show the genetic variation on specific region. The numbers in parenthesis represent the number of features containing at least one genetic variation in their ORF, upstream sequence, or probe.

Region	SNV	INDEL
In all chromosomes	28139	3520
1000 bp upstream	15494 (2469)	2704 (1175)
Open reading frame ORF	17139 (2269)	702 (335)
Microarray probe	2868 (2472)	151(119)

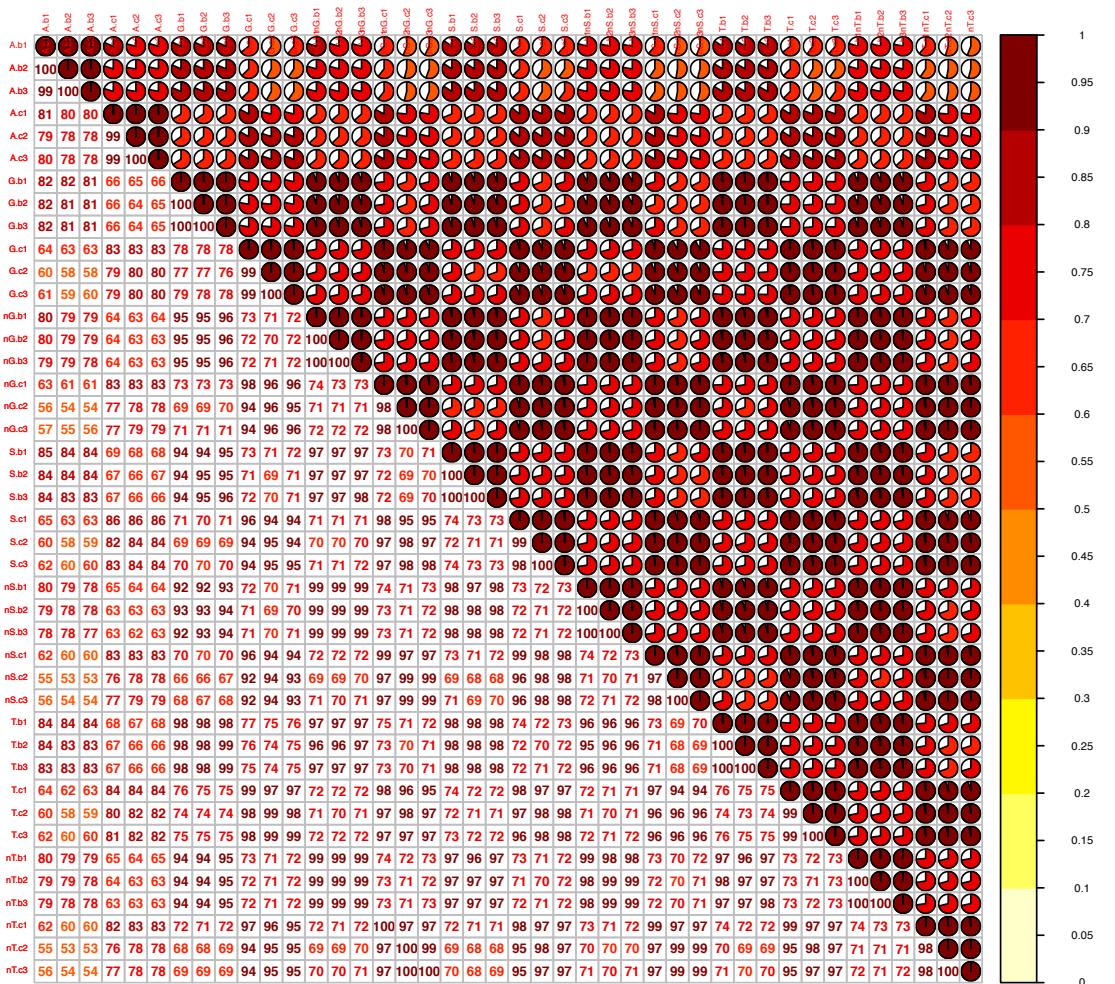


Figure S1 Sample-wise correlation analysis. A = array, G= Gsnap, S = Stumpy, T = TopHat. The non-removed potential PCR duplicates calculations indicate by “n.”

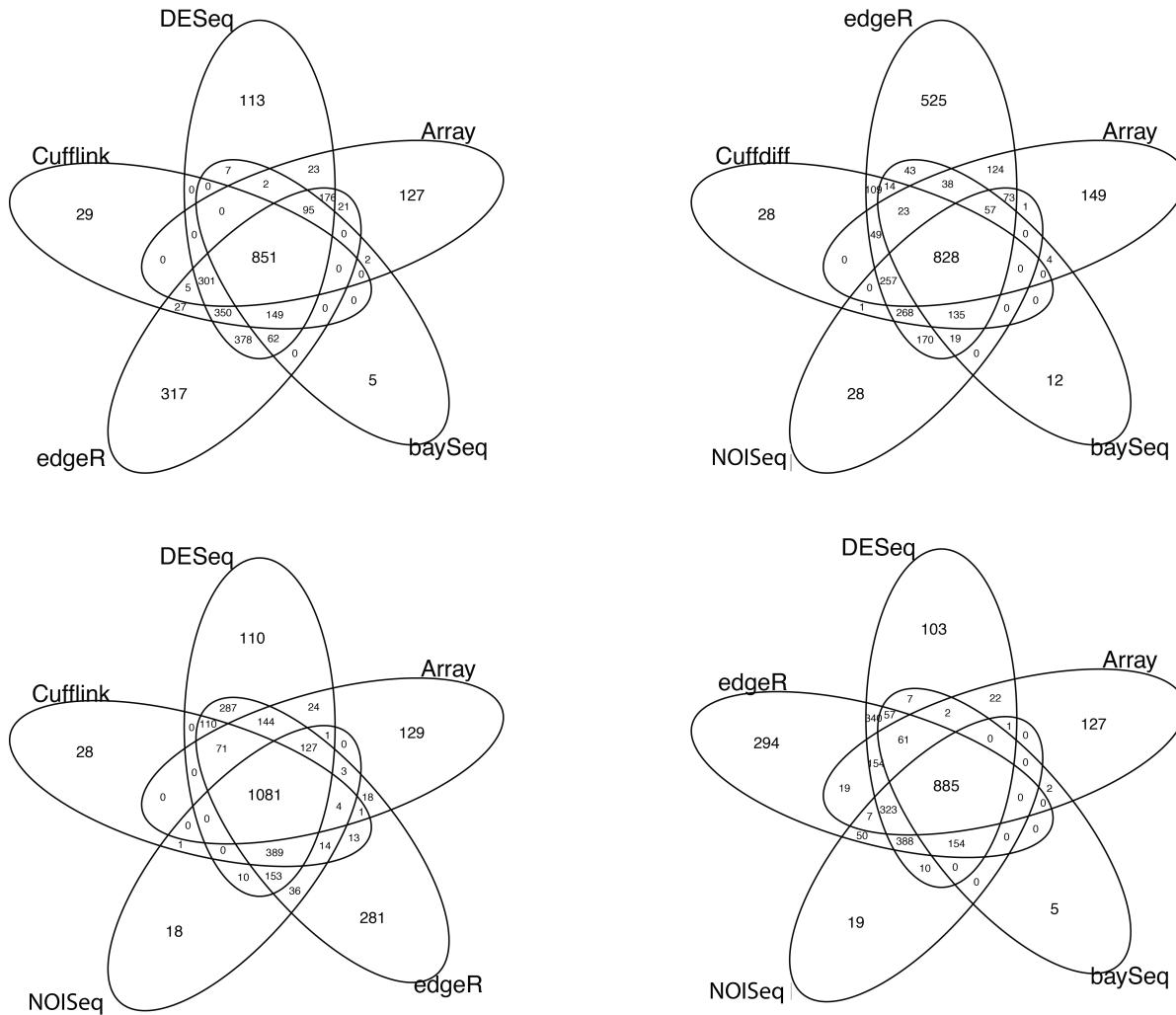


Figure S2 Venn diagram comparison of DGE derived from different statistical analysis with the DGE results from microarray in similar way as Figure 3B

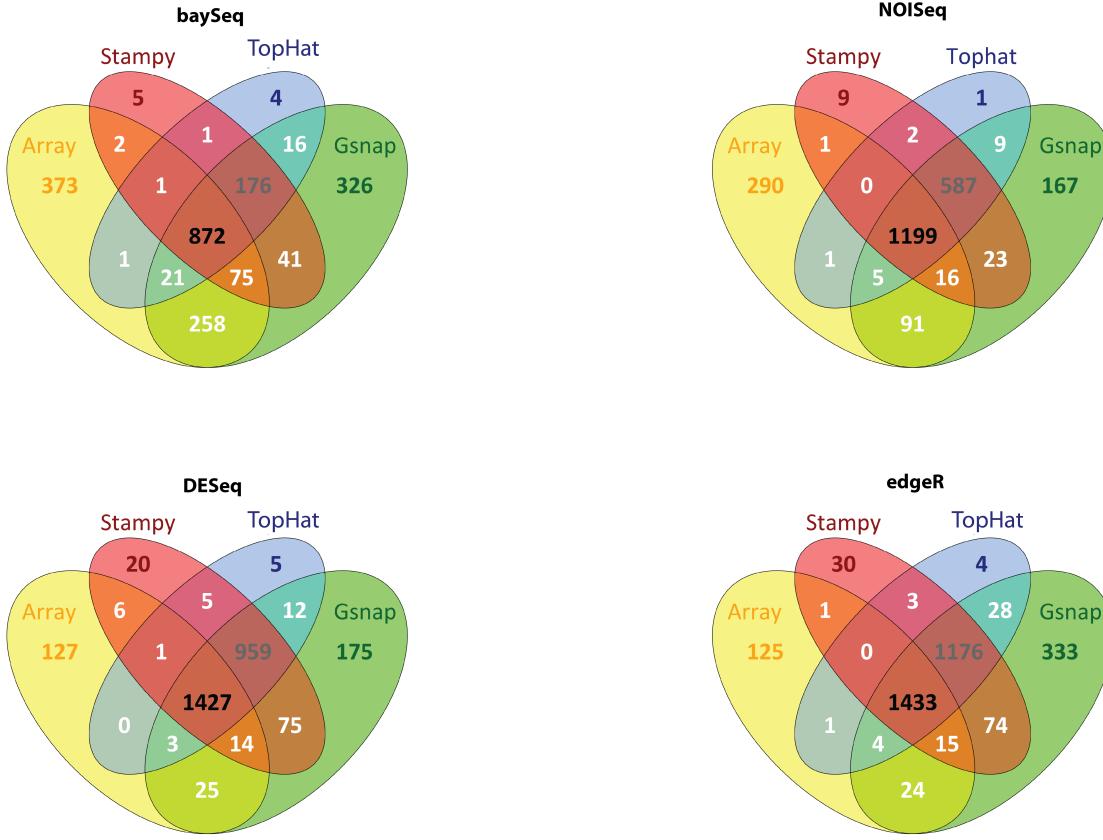


Figure S3 Comparison of DGE using different statistical methods with DGE from microarray in similar way as Figure 3C

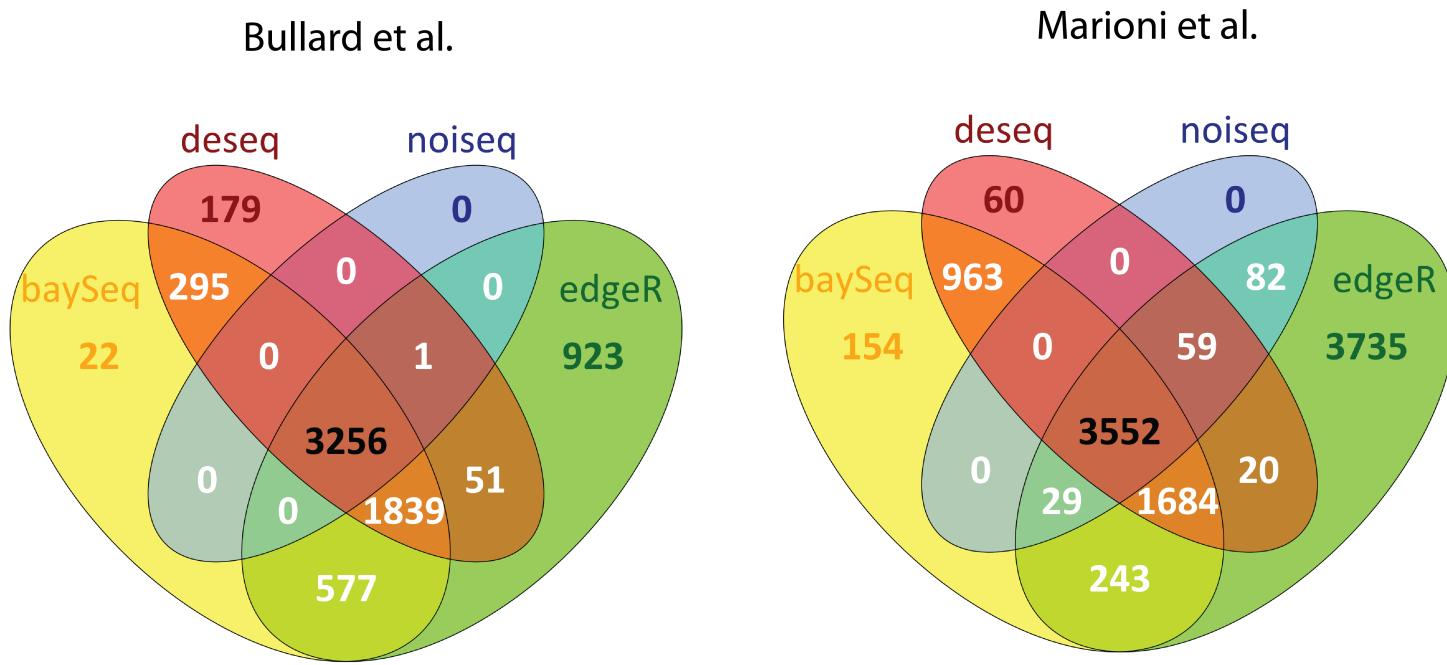
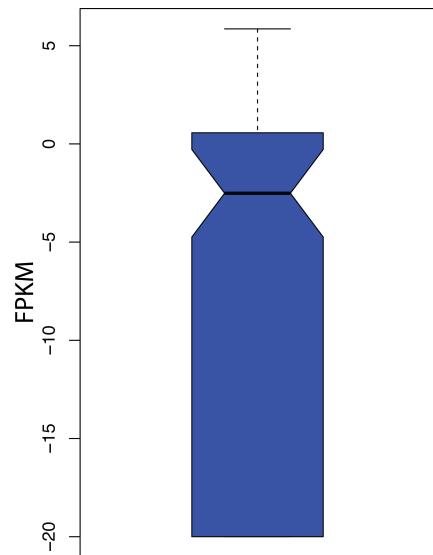


Figure S4 Comparisons of DGE identification using edgeR, baySeq, DESeq and NOISeq methods. The read count table downloaded from 2 published mammalian data sets of Bullard et al ([BMC Bioinformatics](#), 2010 Feb 18;11:94) and Marioni et al. ([Genome Res.](#), 2008 Sep;18(9):1509-17.). The Venn's diagram were made with the cut-off $Q<1e-5$ and $Pr > 0.875$. Cuffdiff method was not included in the comparison because alignment results (SAM/BAM), which are required as input for Cuffdiff, are not provided with the articles.



The 67 genes could not be captured by the
de novo assembly approach

Figure S5 Boxplot shows distribution of gene expression values (FPKM from reference mapping analysis from both batch and chemostat cultures) of the 67 genes could not be captured by the *de novo* assembly approach as full-length transcripts.

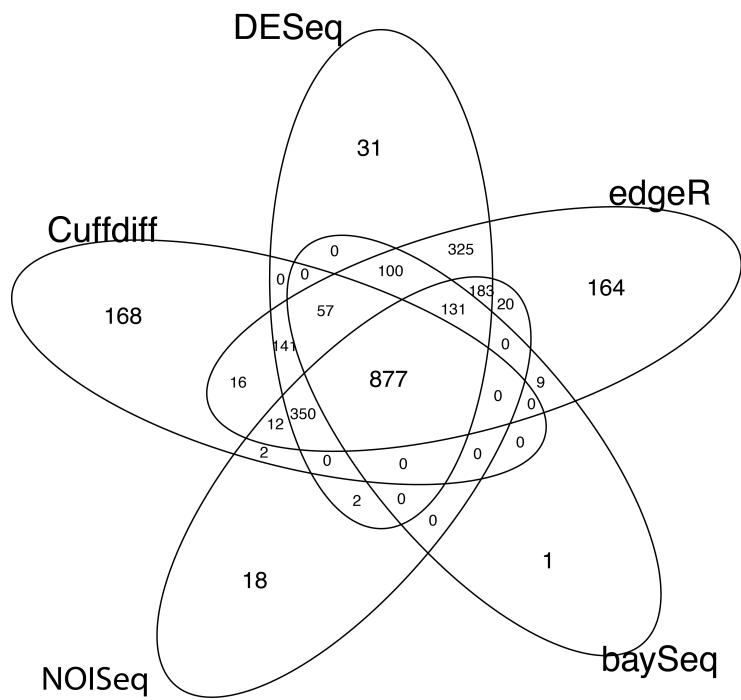


Figure S6 Venn's diagram of the comparison of differential gene expression based on RNA-seq data (result from de novo assembly approach) through five different statistical methods: Cuffdiff, DESeq, NOISeq, edgeR and baySeq.

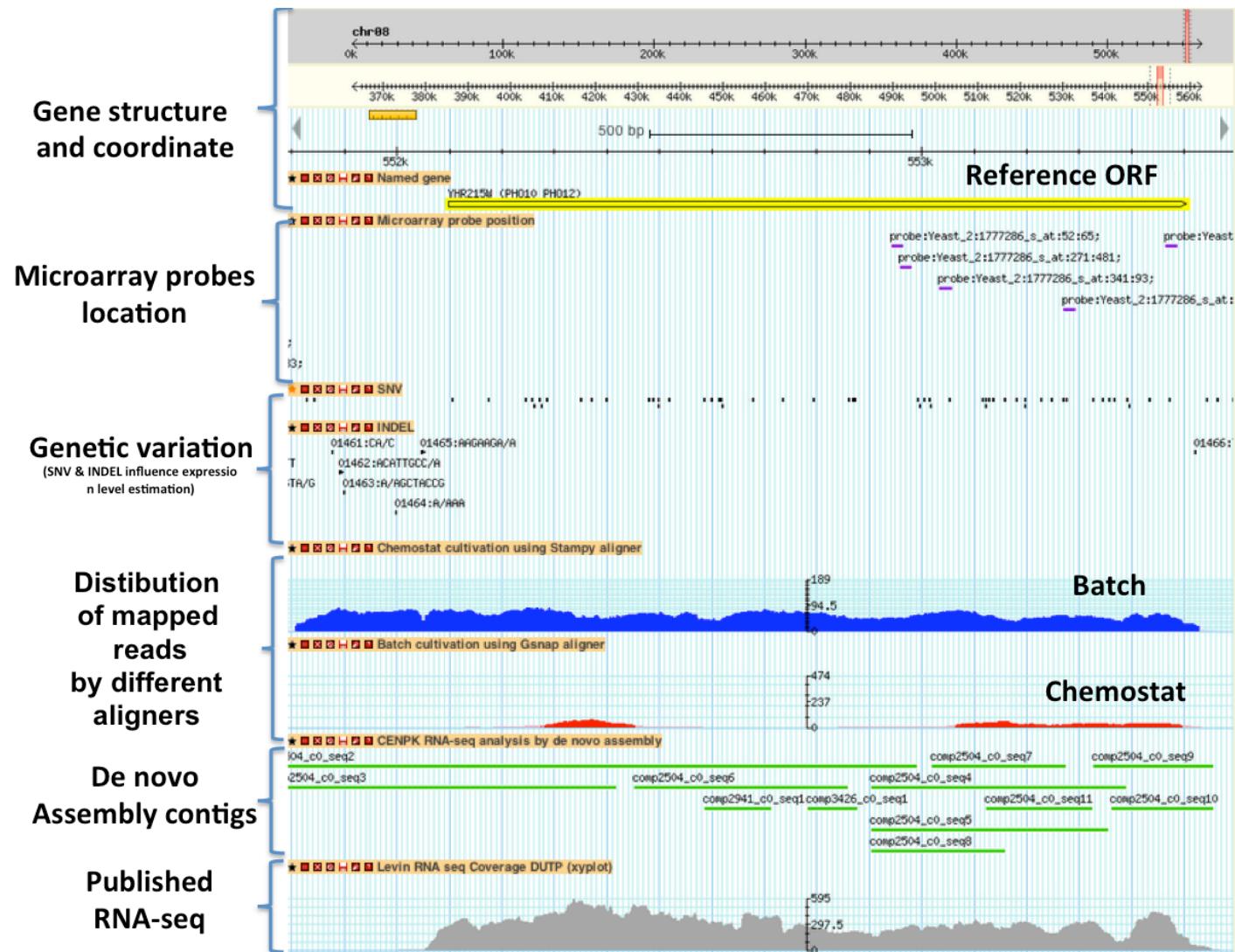


Figure S7 Screen shot of the Yseq browser (www.sysbio.se/Yseq)

Supplementary information

Alignment parameters

Gsnap

```
gsnap -d databasename -D genomedatabase_location -A sam fastq.1 fastq.2 > outfile
```

Stampy

```
stampy.py -f sam --solexa -o outfile -g genomename -h genomehash -M fastq.1 fastq.2
```

TopHat

```
tophat -o outfile indexfile fastq.1 fastq.2
```

Calculate FPKM

```
cufflink -o outfile -G gfffile -b DNAfastafilenames bamfile
```

Generate read count table by HTseq

```
htseq-count -m intersection-strict -s no -t gene -i ID samfile gff_file > outfile
```

Remove PCR duplicate

```
java -jar MarkDuplicates.jar I=input O=output REMOVE_DUPLICATES=true
```

De novo assembly

```
Trinity.pl --seqType fq --jaccard_clip --kmer_method inchworm --CPU 8 --output outfile --left infile_1 --right infile_2
```

Statistical analysis parameters

X = count table

baySeq

```
basysq.test = function(x){  
  repl = c(rep(1,3),rep(2,3))
```

```

groups <- list(NDE = c(1,1,1,1,1,1), DE = c(1,1,1,2,2,2))
CD <- new("countData", data = as.matrix(x), replicates = repl, groups = groups)
CD@libsizes <- getLibsizes(CD)
CDP.NBML <- getPrior.NB(CD, samplesize = 1e6, estimation = "QL", cl = cl)
CDPost.NBML <- getLikelihoods.NB(CDP.NBML, pET = 'BIC', cl = cl)
return(CDPost.NBML)
}

```

DESeq

```

deseq.test = function(x){
conds = c("b","b","b","c","c","c")
cds<-newCountDataSet(x,conds)
cds<-estimateSizeFactors(cds)
sizeFactors(cds)
cds <- estimateDispersions (cds)
res <- nbinomTest (cds, "b", "c")
return(res)
}

```

edgeR

```

edgeR.test = function(x){
group = c("b","b","b","c","c","c")
dat = DGEList(counts=x,group=group)
design <- model.matrix(~factor(group))
dat <- estimateGLMTrendedDisp(dat,design)
dat <- estimateGLMTagwiseDisp(dat,design)
fit <- glmFit(dat,design)
stat.Glm <- glmLRT(dat,fit)
return(stat.Glm)
}

```

NOISeq

```
noiseq.test = function(x) {
mydata <- vector("list", length = 2)
mynames <- rownames(x)
mydata[[1]] <- as.matrix(x[,1:3])
mydata[[2]] <- as.matrix(x[,4:6])
rownames(mydata[[1]]) <- rownames(mydata[[2]]) <- mynames
myresults <- noiseq(mydata[[1]], mydata[[2]], repl = "bio", k = 0.5, norm = "rpkm", long = mylength)
}
```

Cuffdiff

```
cuffdiff -o outfile gfffile -b DNAfastafiles b1.bam,b2.bam,b3.bam c1.bam,c2.bam, c3.bam
```

Software version

FASTX ver. 0.0.13
SolexaQA ver. 1.10
ATLAS2 ver. 1.0
Bowtie ver 0.12.7
BEDTools ver. 2.14.3
SAMTool ver 0.1.18
Gsnap ver. 2012-01-11
GMAP ver. 2012-01-11
Stampy ver. 1.0.13
TopHat ver. 1.4.1
Cufflink/Cuffdiff ver 1.3.0
HTseq ver. 0.5.3p3
Trinity ver. R2012-01-25
baySeq ver. 1.8.3
DESeq ver. 1.6.1
edgeR ver. 2.4.6
NOISeq ver. 2011/04/9