# ELMER: An R/Bioconductor Tool Inferring Regulatory Element Landscapes and Transcription Factor Networks Using Methylomes

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need to be done

## **Contents**

1	Introduction       1.1 Example data	<b>1</b>
2	Installing and Loading ELMER	2
3	Input data 3.1 DNA methylation data 3.2 Gene expression data 3.3 Sample information 3.4 Probe information 3.5 Gene information 3.6 MEE.data	3 3
4	Illustration of ELMER analysis 4.1 Selection of probes within biofeatures 4.2 Identifying differentially methylated probes 4.3 Identifying putative probe-gene pairs 4.4 Motif enrichment analysis on the selected probes 4.5 Identifying regulatory TF	6
5	TCGA.pipe	8
6	Generating figures 6.1 Scatter plots	10

# 1 Introduction

This document provides an introduction of the *ELMER*, which is designed to use DNA methylation and gene expression data sets from a large number of tissue samples to infer regulatory element landscapes and transcription factor network. It includes functions for identifying probes at distal regulatory regions with differential DNA methylation levels, predicting

genes whose expression associates with the differentially methylated probes and discovering the functional regulatory TFs. This package can be easily applied to TCGA public available cancer data sets and to custom DNA methylation and gene expression data sets.

## 1.1 Example data

The following steps can be used to download example data set for *ELMER* 

## 2 Installing and Loading ELMER

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## 3 Input data

The whole pipeline analyses in *ELMER* needs at least 4 input files: a matrix of DNA methylation from HM450K platform; a matrix of gene expression for the same samples; a GRanges object containing the information for probe on HM450K such as names and coordinates; a gene annotation which is also a GRanges object. When TCGA data are used, the samples information will be automatically generated by fetch.mee function. However samples information should be provided when using custom data.

## 3.1 DNA methylation data

Raw DNA methylation data can be processed by *Methylumi* or *minfi* generating DNA methylation information for each CpG. The DNA methylation level at each CpG is referred to as a beta  $(\beta)$  value, calculated as (M/(M+U)), where M represents the methylated allele intensity and U the unmethylated allele intensity. Beta values range from 0 to 1, reflecting the fraction of methylated alleles at each CpG in the each tumor; beta values close to 0 indicates low levels of DNA methylation and beta values close to 1 indicates high levels of DNA methylation. Generate a matrix with DNA methylation beta value for all the samples (columns) and probe loci (rows) and save matrix as meth.rda

```
> load("./ELMER.example/Result/LUSC/LUSC_meth_refined.rda")
> str(Meth)

num [1:1725, 1:268] 0.819 0.842 0.91 0.875 0.333 ...
- attr(*, "dimnames")=List of 2
..$ : chr [1:1725] "cg00045114" "cg00050294" "cg00066722" "cg00093522" ...
..$ : chr [1:268] "TCGA-43-3394-11A-01D-1551-05" "TCGA-43-3920-11B-01D-1551-05" "TCGA-56-8305-01A-11D-229
```

#### 3.2 Gene expression data

Gene expression values can be generated from different platforms such as array or RNA-seq, gene level or transcript level gene expression calling. Generate a matrix with gene expression values for all the samples (columns) and gene (rows) and save matrix as RNA.rda

```
> load("./ELMER.example/Result/LUSC/LUSC_RNA_refined.rda")
> str(GeneExp)
num [1:3898, 1:234] 0 0.43 10.08 6.45 8.59 ...
- attr(*, "dimnames")=List of 2
    ..$ : chr [1:3898] "ID126767" "ID343066" "ID26574" "ID24" ...
    ..$ : chr [1:234] "TCGA-22-5472-01A-01R-1635-07" "TCGA-22-5489-01A-01R-1635-07" "TCGA-22-5491-11A-01R-183
```

## 3.3 Sample information

Sample information should be stored as data.frame object contains sample ID, groups label (such as tumor, normal) and other description for each sample. When TCGA data were used, tumor, normal group label will be automatically generated by fetch.mee function by specify option TCGA=TRUE.

```
> mee <- fetch.mee(meth=Meth, exp=GeneExp, TCGA=T)
~~~ MEE.data: initializator ~~~
> head(getSample(mee))
                             ID
                                                     meth.ID
                                                                                   exp.ID
TCGA-43-3394-11 TCGA-43-3394-11 TCGA-43-3394-11A-01D-1551-05 TCGA-43-3394-11A-01R-1758-07
TCGA-56-8305-01 TCGA-56-8305-01 TCGA-56-8305-01A-11D-2294-05 TCGA-56-8305-01A-11R-2296-07
TCGA-56-8307-01 TCGA-56-8307-01 TCGA-56-8307-01A-11D-2294-05 TCGA-56-8307-01A-11R-2296-07
TCGA-56-8308-01 TCGA-56-8308-01 TCGA-56-8308-01A-11D-2294-05 TCGA-56-8308-01A-11R-2296-07
TCGA-56-8309-01 TCGA-56-8309-01 TCGA-56-8309-01A-11D-2294-05 TCGA-56-8309-01A-11R-2296-07
TCGA-58-8386-01 TCGA-58-8386-01 TCGA-58-8386-01A-11D-2294-05 TCGA-58-8386-01A-11R-2296-07
TCGA-43-3394-11 Normal
TCGA-56-8305-01 Tumor
TCGA-56-8307-01 Tumor
TCGA-56-8308-01 Tumor
TCGA-56-8309-01 Tumor
TCGA-58-8386-01 Tumor
```

## 3.4 Probe information

Probe information should be stored as GRanges object containing coordinate of each probe on the DNA methylation array and names of each probe. The default probe information is for HM450K.

name	- 1	strand	ranges		seqnames	
<character></character>	- 1	<rle></rle>	<pre><iranges></iranges></pre>		<rle></rle>	
cg00045114	- 1	*	172674159]	[172674159,	chr1	[1]
cg00050294	- 1	*	2886818]	[ 2886818,	chr1	[2]
cg00066722	- 1	*	43634520]	[ 43634520,	chr1	[3]
cg00093522	- 1	*	2252019]	[ 2252019,	chr1	[4]
cg00107046	- 1	*	16465562]	[ 16465562,	chr1	[5]

```
chr1 [215147891, 215147891]
[1721]
                                                     cg27589988
[1722]
          chr1 [ 46632696, 46632696]
                                                     cg27636310
          chr1 [ 3472204,
                            3472204]
                                                     cg27637706
[1723]
          chr1 [ 3283394, 3283394]
[1724]
                                             ch.1.131529R
          chr1 [174947362, 174947362]
                                              | ch.1.173213985R
[1725]
```

seqinfo: 24 sequences from an unspecified genome; no seqlengths

#### 3.5 Gene information

Gene information should be stored as GRanges object containing coordinates of each gene, gene id, gene symbol and gene isoform id. The default gene information is UCSC gene annotation.

- > load(system.file("extdata","UCSC\_gene\_hg19.rda",package = "ELMER"))
- > ## In TCGA expression data, geneID were used as rowname for each row. However numbers
- > ## can't be rownames, "ID" was added to each gene id functioning as rowname.
- > ## If your geneID is consistent with rownames of gene expression matrix, adding "ID"
- > ## to geneID can be skipped.
- > txs\$GENEID <- pasteO("ID",txs\$GENEID)</pre>
- > geneInfo <- promoters(txs,upstream = 0, downstream = 0)
- > save(geneInfo,file="./ELMER.example/Result/LUSC/geneAnnot.rda")
- > mee <- fetch.mee(meth=Meth, exp=GeneExp, TCGA=T, geneInfo=txs)
- ~~~ MEE.data: initializator ~~~
- > getGeneInfo(mee)

GRanges object with 13741 ranges and 3 metadata columns:

seqnames			ranges	strand	- 1	tx_name	GENEID	SYMBOL
	<rle></rle>	•	<iranges></iranges>	<rle></rle>	- 1	<character></character>	<character></character>	<character></character>
10000	chr1	[243651535, 24	44006584]	-		uc001hzz.1	ID10000	AKT3
10000	chr1	[243663021, 24	44006584]	-		uc001iab.2	ID10000	AKT3
10000	chr1	[243663021, 24	44006886]	-	- 1	uc021plu.1	ID10000	AKT3
10001	chr14	[ 71050957,	71067384]	-		uc001xmf.3	ID10001	MED6
10001	chr14	[ 71050957,	71067384]	-		uc010tth.2	ID10001	MED6
9988	chr7	[86781677, 8	86825648]	+		uc003uik.3	ID9988	DMTF1
9988	chr7	[86781677, 8	86825648]	+		uc011khb.2	ID9988	DMTF1
9988	chr7	[86781870, 8	86825648]	+	- 1	uc003uil.3	ID9988	DMTF1
9988	chr7	[86792198, 8	86809018]	+		uc003uim.1	ID9988	DMTF1
9988	chr7	[86792198, 8	86825648]	+	- 1	uc003uin.3	ID9988	DMTF1

-----

seqinfo: 93 sequences (1 circular) from hg19 genome

#### 3.6 MEE.data

The above 5 components will form MEE.data object as the main input for mulitple functions in ELMER.

- > mee <- fetch.mee(meth=Meth, exp=GeneExp, TCGA=T, probeInfo=probe, geneInfo=txs)
- ~~~ MEE.data: initializator ~~~
- > mee
- \*\*\* Class MEE.data, method show \*\*\*
- \* meth

```
num [1:1725, 1:234] 0.819 0.842 0.91 0.875 0.333 ...
 - attr(*, "dimnames")=List of 2
  ..$ : chr [1:1725] "cg00045114" "cg00050294" "cg00066722" "cg00093522" ...
  ..$ : chr [1:234] "TCGA-43-3394-11A-01D-1551-05" "TCGA-56-8305-01A-11D-2294-05" "TCGA-56-8307-01A-11D-2294-05"
NULL
* exp
 num [1:3894, 1:234] 0 0.214 10.048 5.007 8.63 ...
 - attr(*, "dimnames")=List of 2
  ..$ : chr [1:3894] "ID126767" "ID343066" "ID26574" "ID24" ...
  ..$: chr [1:234] "TCGA-43-3394-11A-01R-1758-07" "TCGA-56-8305-01A-11R-2296-07" "TCGA-56-8307-01A-11R-2296-07"
NULL
* sample
'data.frame':
                     234 obs. of 4 variables:
 $ ID : chr "TCGA-43-3394-11" "TCGA-56-8305-01" "TCGA-56-8307-01" "TCGA-56-8308-01" ...
 $ meth.ID: chr "TCGA-43-3394-11A-01D-1551-05" "TCGA-56-8305-01A-11D-2294-05" "TCGA-56-8307-01A-11D-2294-05"
 $ exp.ID : chr "TCGA-43-3394-11A-01R-1758-07" "TCGA-56-8305-01A-11R-2296-07" "TCGA-56-8307-01A-11R-2296-07"
 $ TN : chr "Normal" "Tumor" "Tumor" "Tumor" ...
NULL
* probeInfo
GRanges object with 1725 ranges and 1 metadata column:
```

	seqnames		ranges	strand		name
	<rle></rle>		<iranges></iranges>	<rle></rle>	- 1	<character></character>
[1]	chr1	[172674159,	172674159]	*	- 1	cg00045114
[2]	chr1	[ 2886818,	2886818]	*	1	cg00050294
[3]	chr1	[ 43634520,	43634520]	*	-	cg00066722
[4]	chr1	[ 2252019,	2252019]	*	- 1	cg00093522
[5]	chr1	[ 16465562,	16465562]	*	- 1	cg00107046
[1721]	chr1	[215147891,	215147891]	*	-	cg27589988
[1722]	chr1	[ 46632696,	46632696]	*	-	cg27636310
[1723]	chr1	[ 3472204,	3472204]	*	-	cg27637706
[1724]	chr1	[ 3283394,	3283394]	*	-	ch.1.131529R
[1725]	chr1	[174947362,	174947362]	*	-	ch.1.173213985R
	_					

seqinfo: 24 sequences from an unspecified genome; no seqlengths

\* geneInfo

GRanges object with 13741 ranges and 3 metadata columns:

	seqnames		ranges	strand	- 1	tx_name	GENEID	SYMBOL
	<rle></rle>		<iranges></iranges>	<rle></rle>	- 1	<character></character>	<character></character>	<character></character>
10000	chr1	[243651535,	244006584]	-	- 1	uc001hzz.1	ID10000	AKT3
10000	chr1	[243663021,	244006584]	-	- 1	uc001iab.2	ID10000	AKT3
10000	chr1	[243663021,	244006886]	-	- 1	uc021plu.1	ID10000	AKT3
10001	chr14	[ 71050957,	71067384]	-	- 1	uc001xmf.3	ID10001	MED6
10001	chr14	[ 71050957,	71067384]	-	- 1	uc010tth.2	ID10001	MED6
9988	chr7	[86781677]	, 86825648]	+	- 1	uc003uik.3	ID9988	DMTF1
9988	chr7	[86781677]	, 86825648]	+	- 1	uc011khb.2	ID9988	DMTF1
9988	chr7	[86781870]	, 86825648]	+	- 1	uc003uil.3	ID9988	DMTF1
9988	chr7	[86792198]	, 86809018]	+	- 1	uc003uim.1	ID9988	DMTF1
9988	chr7	[86792198]	, 86825648]	+	- 1	uc003uin.3	ID9988	DMTF1

seqinfo: 93 sequences (1 circular) from hg19 genome
\*\*\*\*\*\* End Print (MEE.data) \*\*\*\*\*\*\*

## 4 Illustration of ELMER analysis

A subset of chromosome 1 data from TCGA LUSC were used as illustruation.

## 4.1 Selection of probes within biofeatures

Function, get.feature.probe, is used to select probes that locates within the biofeatures such as H3K27ac ChIP-seq peaks. As default, get.feature.probe function will automatically select distal enhancer probes on HM450K which are at least 2kb away from TSS annotated by GENCODE V15 and UCSC-gene and locate within the putative comprehensive enhancers from REMC, ENCODE and FANTOM5.

```
> #get distal enhancer probes that are 2kb away from TSS and overlap with REMC and FANTOM5
> #enhancers on chromosome 1
> Probe <- get.feature.probe(probe=probe, rm.chr=paste0("chr",c(2:22,"X","Y")))
> save(Probe,file="./ELMER.example/Result/LUSC/probeInfo_feature.rda")
```

## 4.2 Identifying differentially methylated probes

Function, get.diff.meth, will be used to identify differentially methylated probes among the ones within biofeatures, which are selected in the above step.

```
> ## fetch.mee can take path as input.
> mee <- fetch.mee(meth="./ELMER.example/Result/LUSC/LUSC_meth_refined.rda",
                   exp="./ELMER.example/Result/LUSC/LUSC_RNA_refined.rda", TCGA=T,
+
+
                   probeInfo="./ELMER.example/Result/LUSC/probeInfo_feature.rda",
                   geneInfo="./ELMER.example/Result/LUSC/geneAnnot.rda")
~~~ MEE.data: initializator ~~~
> sig.diff <- get.diff.meth(mee, cores=detectCores()/2, dir.out ="./ELMER.example/Result/LUSC",
                            diff.dir="hypo", pvalue = 0.01)
> str(sig.diff$hypo)
                       ## significantly hypomethylated probes
'data.frame':
                    742 obs. of 4 variables:
$ probe
                 : chr "cg00045114" "cg00050294" "cg00093522" "cg00163018" ...
                : num 7.31e-13 4.44e-09 1.14e-23 2.24e-21 2.77e-18 ...
$ pvalue
$ tumorMinNormal: num -0.35 -0.5 -0.36 -0.384 -0.374 ...
$ adjust.p
                 : num 2.42e-12 1.06e-08 1.00e-22 1.56e-20 1.46e-17 ...
> # get.diff.meth automatically save output files.
> # getMethdiff.hypo.probes.csv contains statistics for all the probes.
> # getMethdiff.hypo.probes.significant.csv contains only the significant probes.
> dir(path = "./ELMER.example/Result/LUSC", pattern = "getMethdiff")
[1] "getMethdiff.hypo.probes.csv"
                                              "getMethdiff.hypo.probes.significant.csv"
```

## 4.3 Identifying putative probe-gene pairs

Function, get.pair function, will be used to identify putative target genes for selected probes. This step is the most time consuming step since it contains a large amount calculations for permutation. The more permutation time is, the longer it will take. It is recommended to use multiple cores for this step. Default permutation time is 1000 which may need 12 hrs by 4 cores. However 10,000 permutations is recommended which may cost 2 days, if high confident results are desired.

```
> ### identify target gene for significantly hypomethylated probes.
>
> Sig.probes <- read.csv("./ELMER.example/Result/LUSC/getMethdiff.hypo.probes.significant.csv",
                        stringsAsFactors=F)[,1]
> head(Sig.probes) # significantly hypomethylated probes
[1] "cg00045114" "cg00050294" "cg00093522" "cg00163018" "cg00173804" "cg00223046"
> ## Collect nearby 20 gene for Sig.probes
> nearGenes <-GetNearGenes(TRange=getProbeInfo(mee,probe=Sig.probes),
                          geneAnnot=getGeneInfo(mee),cores=detectCores()/2)
> ## Identify significant probe-gene pairs
> Hypo.pair <-get.pair(mee=mee,probes=Sig.probes,nearGenes=nearGenes,
                      permu.dir="./ELMER.example/Result/LUSC/permu",permu.size=300,Pe = 0.01,
                      dir.out="./ELMER.example/Result/LUSC",cores=detectCores()/2,label= "hypo")
> head(Hypo.pair) ## significant probe-gene pairs
                        Probe
                                GeneID
                                         Symbol Distance Sides
                                                                      Raw.p
cg20701183.ID8543
                  cg20701183
                                ID8543
                                          LMO4
                                                    2563 L1 7.453984e-14 0.003322259
cg19403323.ID255928 cg19403323 ID255928
                                                   87477
                                                            R1 1.671937e-12 0.003322259
                                          SYT14
cg12213388.ID84451 cg12213388 ID84451 KIAA1804 993548 L4 2.527644e-12 0.003322259
cg26607897.ID55811 cg26607897 ID55811
                                         ADCY10 292476 R3 4.593610e-12 0.003322259
cg10574861.ID8543
                   cg10574861
                                ID8543
                                           LMO4
                                                    4715
                                                            L1 4.770162e-12 0.003322259
cg26607897.ID23432 cg26607897 ID23432
                                                            R6 8.048248e-12 0.003322259
                                         GPR161
                                                  563308
> # get.pair automatically save output files.
> #getPair.hypo.all.pairs.statistic.csv contains statistics for all the probe-gene pairs.
> #getPair.hypo.pairs.significant.csv contains only the significant probes.
> dir(path = "./ELMER.example/Result/LUSC", pattern = "getPair")
[1] "getPair.hypo.all.pairs.statistic.csv" "getPair.hypo.pairs.significant.csv"
```

#### 4.4 Motif enrichment analysis on the selected probes

Function, get.enriched.motif, will be used to calculate enrichment of the motifs from factorbook and JASPER for the selected probes. Odds Ratio is used to assess the enrichment levels and 95% confidence interval of Odds Ratio is calculated.

```
> ### identify enriched motif for significantly hypomethylated probes which
> ##have putative target genes.
> Sig.probes.paired <- read.csv("./ELMER.example/Result/LUSC/getPair.hypo.pairs.significant.csv",
                                stringsAsFactors=F)[,1]
> head(Sig.probes.paired) # significantly hypomethylated probes with putative target genes
[1] "cg20701183" "cg19403323" "cg12213388" "cg26607897" "cg10574861" "cg26607897"
> enriched.motif <-get.enriched.motif(probes=Sig.probes.paired,
                                      dir.out="./ELMER.example/Result/LUSC", label="hypo",
                                      min.incidence = 10,lower.OR = 1.1)
6 motifs are enriched.
> names(enriched.motif) # enriched motifs
             "BARHL2" "IRF"
                               "PRDM1" "TCF7L2" "TP53"
[1] "AP1"
> # get.enriched.motif automatically save output files.
> # getMotif.hypo.enriched.motifs.rda contains enriched motifs and the probes with the motif.
```

```
> # getMotif.hypo.motif.enrichment.csv contains summary of enriched motifs.
> dir(path = "./ELMER.example/Result/LUSC", pattern = "getMotif")
[1] "getMotif.hypo.enriched.motifs.rda" "getMotif.hypo.motif.enrichment.csv"
> # motif enrichment figure will be automatically generated.
> dir(path = "./ELMER.example/Result/LUSC", pattern = "motif.enrichment.pdf")
[1] "hypo.motif.enrichment.pdf"
```

## 4.5 Identifying regulatory TF

Function, get.TFs, will use the anti-correlation of a particular TF and the level of demethylation at its binding sites to predict the regulatory TF.

```
> ### identify regulatory TF for the enriched motifs
> load("./ELMER.example/Result/LUSC/getMotif.hypo.enriched.motifs.rda")
> TF <- get.TFs(mee=mee, enriched.motif=enriched.motif,dir.out="./ELMER.example/Result/LUSC",
                cores=detectCores()/2, label= "hypo")
> # get.TFs automatically save output files.
> # getTF.hypo.TFs.with.motif.pvalue.rda contains statistics for all TF with average
> # DNA methylation at sites with the enriched motif.
> # getTF.hypo.significant.TFs.with.motif.summary.csv contains only the significant probes.
> dir(path = "./ELMER.example/Result/LUSC", pattern = "getTF")
[1] "getTF.hypo.significant.TFs.with.motif.summary.csv"
[2] "getTF.hypo.TFs.with.motif.pvalue.rda"
> # TF ranking plot based on statistics will be automatically generated.
> dir(path = "./ELMER.example/Result/LUSC/TFrankPlot", pattern = "pdf")
[1] "AP1.TFrankPlot.pdf"
                            "BARHL2.TFrankPlot.pdf" "IRF.TFrankPlot.pdf"
[4] "PRDM1.TFrankPlot.pdf" "TCF7L2.TFrankPlot.pdf" "TP53.TFrankPlot.pdf"
```

# 5 TCGA.pipe

Function, TCGA.pipe, is easy usage for downloading TCGA data and perform all the analyses in ELMER. For illustration purpose, we skip downloading step. User could use getTCGA function to download TCGA data or use TCGA.pipe include "download" to analysis option.

## 6 Generating figures

## 6.1 Scatter plots

## 6.2 Schematic plot

Schematic plot shows a breif view of linkages between genes and probes. To make schematic plot, "Pair" object should be generated first.

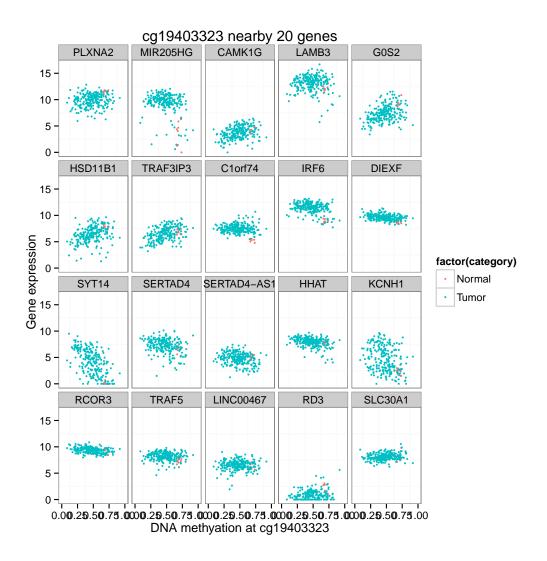


Figure 1: Each scatter plot shows the methylation level of an example probe cg19403323 in all LUSC samples plotted against the expression of one of 20 adjacent genes.

- a. Generate schematic plot for one probe with nearby 20 genes and label the gene significantly linked with the probe in red. Figure44
- > grid::grid.newpage()
- > schematic.plot(pair=pair, by Probe="cg19403323", dir.out="./ELMER.example/Result/LUSC", save=F)
- > schematic.plot(pair=pair, byProbe="cg19403323",dir.out="./ELMER.example/Result/LUSC",save=TRUE)
- b. Generate schematic plot for one gene with the probes which the gene significanlty linked to. Figure 5??
- > grid::grid.newpage()
- > schematic.plot(pair=pair, byGene="ID255928",dir.out="./ELMER.example/Result/LUSC",save=F)
- > schematic.plot(pair=pair, byGene="ID255928",dir.out="./ELMER.example/Result/LUSC",save=TRUE)

## 6.3 Motif enrichment plot

Motif enrichment plot shows the enrichment levels for the selected motifs. Figure 6??

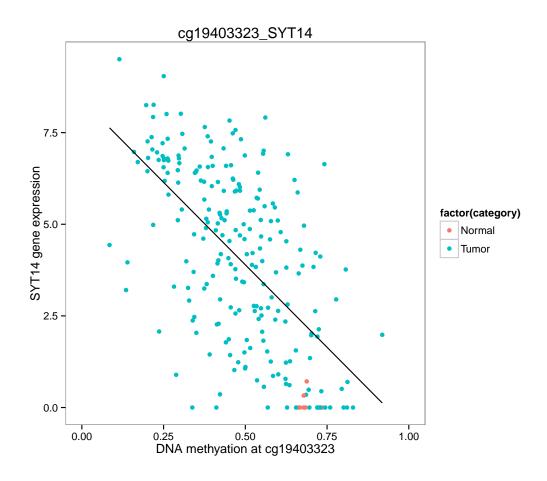


Figure 2: Scatter plot shows the methylation level of an example probe cg19403323 in all LUSC samples plotted against the expression of putative target gene SYT14.

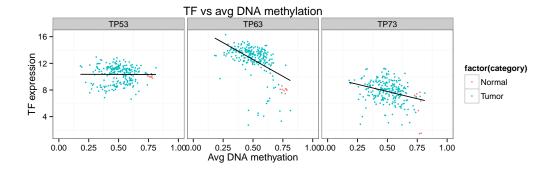


Figure 3: Each scatter plot shows the average methylation level of sites with TP53 motif in all LUSC samples plotted against the expression of transcription factor TP53, TP63, TP73 respectively.

#### chr1:209605478-211752100

2.15Mb

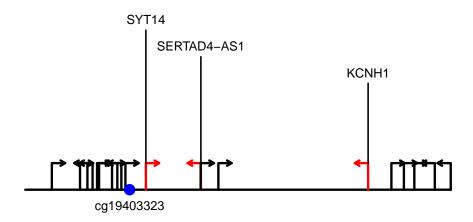


Figure 4: The schematic plot shows probe colored by blue and nearby 20 genes location. The gene significantly linked to the probe were shown in red.

## 6.4 TF ranking plot

TF ranking plot shows statistic -log10(P value) assessing the anti-correlation level of TFs expression level with average DNA methylation level at sites with a given motif. Figure 7??

```
> load("./ELMER.example/Result/LUSC/getTF.hypo.TFs.with.motif.pvalue.rda")
> TF.rank.plot(motif.pvalue=TF.meth.cor, motif="TP53", TF.label=list(TP53=c("TP53","TP63","TP73")),
+ dir.out="./ELMER.example/Result/LUSC/TFrankPlot", save=F)
$TP53
```

> TF.rank.plot(motif.pvalue=TF.meth.cor, motif="TP53", TF.label=list(TP53=c("TP53", "TP73")),

dir.out="./ELMER.example/Result/LUSC/TFrankPlot", save=T) # save to pdf







Figure 5: The schematic plot shows the gene colored in red and all probes in blue, which are significantly linked to the expression of this gene.

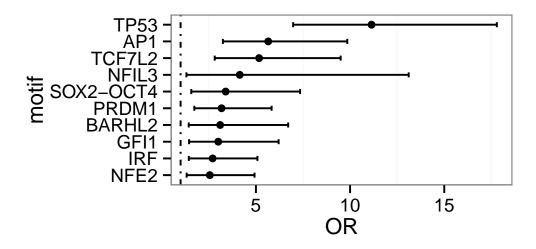


Figure 6: The plot shows the Odds Ratio (x axis) for the selected motifs with OR above 1.3 and lower boundary of OR above 1.3. The range shows the 95% confidence interval for each Odds Ratio.

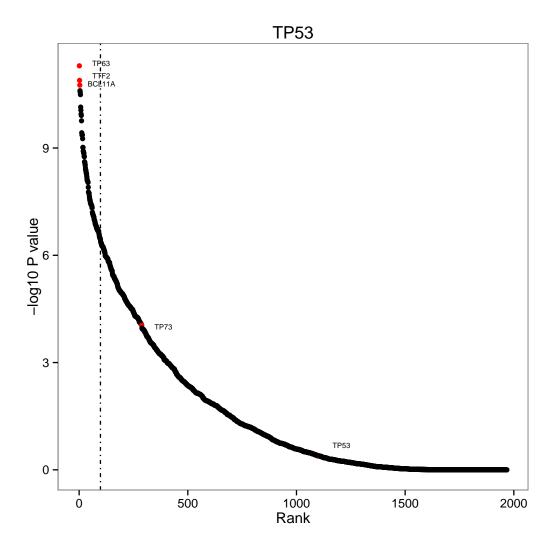


Figure 7: Shown are TF ranking plots based on the score ( $-\log(P \text{ value})$ ) of association between TF expression and DNA methylation of the TP53 motif in the LUSC cancer type . The dashed line indicates the boundary of the top 5% association score. The top 3 associated TFs and the TF family members (dots in red) that are associated with that specific motif are labeled in the plot.