GeneNetworkBuilder Guide

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Contents

1 Introduction

Transcription factors (TFs), chromatin modifications and microRNAs (miRNAs) are important in regulating gene expression[1]. Chromatin immunoprecipitation (ChIP) followed by high-throughput sequencing (ChIP-seq)[2] or genome tiling array analysis (ChIP-chip)[3] are widely used technologies for identifying genome-wide binding sites of TFs (TFBDs)[4]. The role of the TF on genome-wide gene expression can be determined by expression microarray or RNA-seq experiments[5]. By combining both technologies, researchers have the potential to decipher the regulatory network of the TF. Genes bound by the TF and altered in expression are considered direct targets of the TF, and genes altered in expression but not bound by the TF are the indirect targets of the TF. The indirect targets can potentially form a complex network itself, especially when the TF is a master regulator who regulates other TFs. To facilitate identification of the complex regulatory network of TFs and how indirect targets are inter-connected, we have developed GeneNetworkBuilder (GNB). Each

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genereated network is consisted of a directed acyclic graph with each edge representing TF -> gene, TF -> miRNA, or miRNA -> gene where a -> b represents "a regulates b".

2 Examples of using GeneNetworkBuilder

To use GNB, users need to input a list of genes bound by a given TF and another list of genes/miRNAs with altered expression by knockdown/knockout of the same TF. The bound gene list could be obtained from ChIP-seq or ChIP-chip experiment. The gene list with altered expression are from RNA-seq or expression microarray experiment. ChIP experiments and expression experiments should preferably be performed in similar experimental condition such as tissue type, development stage etc. In addition, users need to select a TF regulatory network from GNB or upload customized TF regulatory network.

GNB provides two embedded regulatory networks. One is designed for *Caenorhabditis elegans* combining database EDGEdb[6] and microCosm Targets[7]. Database MicroCosm Targets contains computationally predicted targets for miRNAs across many species. EDGEdb contains experimentally determined interactions of ~934 worm TFs by high-throughput yeast on-hybrid (Y1H) assay. And the other embedded regulatory network is designed for *Homo sapiens* combining database FANTOM[8], miRGen[9] and microCosm Targets[7]. FANTOM stores physical interactions among the majority of human/mouse DNA-binding TFs. The miRGen is an integrated database of miRNA regulation by TFs and miRNA targets mainly for human. Figure 1 depicts the relationships of various databases and its role in GNB.

2.1 Quick start

Here is an example to use GNB to generate a simple regulatory network for *C. elegans*. There are three steps,

1. buildNetwork, build the network by GNB-embedded or user-defined regulatory network starting from the bound gene list.

2. filterNetwork, filter the network by differential expressed genes/miRNAs.

3. polishNetwork, generate the graphNEL object with display style.

```
> ##
> library(GeneNetworkBuilder)
> ##load C. elegans miRNA ID lists
> data("ce.miRNA.map")
> ##load GNB-embedded regulatory network of C. elegans.
> data("ce.interactionmap")
> ##load data required
> data("example.data")
> ##build the network by binding list and interaction map
> sifNetwork<-buildNetwork(TFbindingTable=example.data$ce.bind,
+                          interactionmap=ce.interactionmap, level=2)
> ##filter the network by expression data
> cifNetwork<-filterNetwork(rootgene="WBGene00000912", sifNetwork=sifNetwork,
+                  exprsData=uniqueExprsData(example.data$ce.exprData),
```
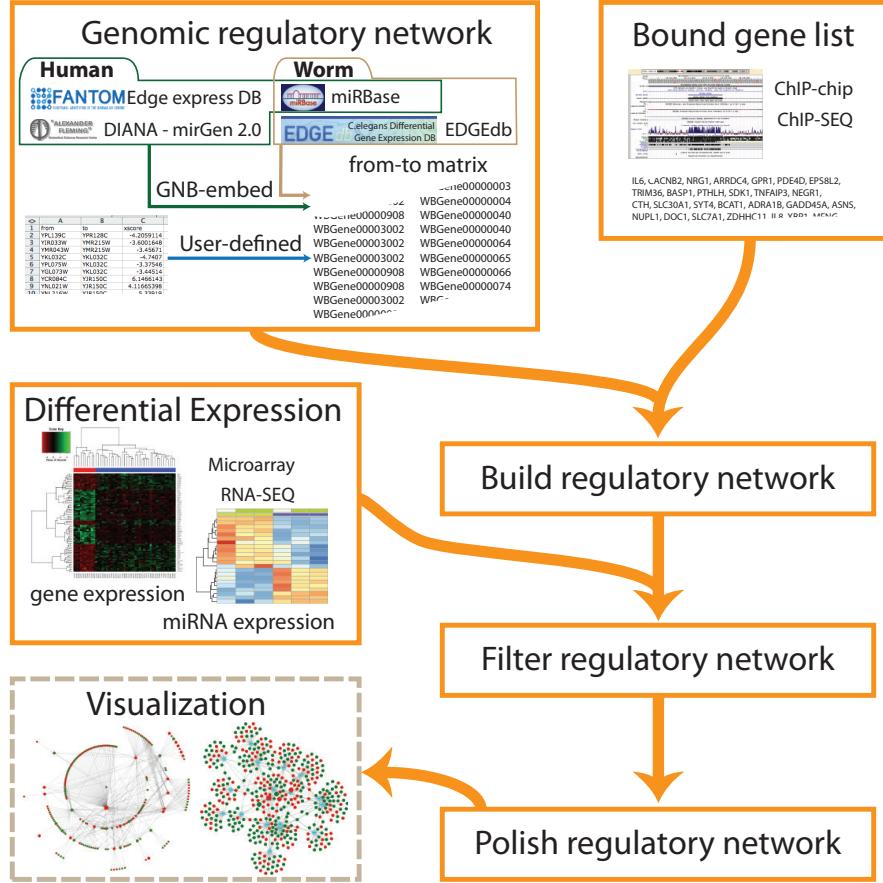
Figure 1: workflow of GeneNetworkBuilder

```
+                       mergeBy="symbols",
+                       miRNAlist=as.character(ce.miRNA.map[ , 1]),
+                       remove_miRNA=FALSE, tolerance=1)
> ##generate graphNEL object for the network
> gR<-polishNetwork(cifNetwork=cifNetwork, nodecolor=colorRampPalette(c("green", "yellow", "red"))(5))
```

2.2   Example using gene expression profile

Here is an example to use GNB to generate a simple regulatory network for *C. elegans*. And also show some examples how to use the *graphNEL* object for further analysis.

```
> library(GeneNetworkBuilder)
> data("example.data")
> ##Initialize a binding matrix by TF and the related gene lists of TFBDs.
> ##For example, TF is daf-16, and the ChIP-chip result indicates that it can bind to
> ##upstream regions of gene "zip-2", "zip-4", "nhr-3" and "nhr-66".
> bind<-cbind(from="daf-16", to=c("zip-2", "zip-4", "nhr-3", "nhr-66"))
> ##For same gene, there are multple gene alias. In order to eliminate the possibility of
> ##missing any interactions, convert the gene symbols to unique gene ids is important.
```

```
> data("ce.IDsMap")
> bind<-convertID(toupper(bind), IDsMap=ce.IDsMap, ByName=c("from", "to"))
> ##build the network by binding list and interaction map
> data("ce.interactionmap")
> sifNetwork<-buildNetwork(TFbindingTable=example.data$ce.bind,
+                          interactionmap=ce.interactionmap, level=2)
> ##filter the network by expression data
> ##For each gene id, it should have only single record for expression change.
> unique.ce.microarrayData<-uniqueExprsData(example.data$ce.exprData,
+                          method="Max", condenseName='logFC')
> data("ce.miRNA.map")
> cifNetwork<-filterNetwork(rootgene="WBGene00000912", sifNetwork=sifNetwork,
+                       exprsData=unique.ce.microarrayData, mergeBy="symbols",
+                       miRNAlist=as.character(ce.miRNA.map[ , 1]),
+                       tolerance=1, cutoffPVal=0.01, cutoffLFC=1)
> ##convert the unique gene ids back to gene symbols
> data("ce.mapIDs")
> cifNetwork<-convertID(cifNetwork, ce.mapIDs, ByName=c("from","to"))
> ##generate graphNEL object for the network
> gR<-polishNetwork(cifNetwork, nodecolor=colorRampPalette(c("green", "yellow", "red"))(10))
> ##plot the figure
> library(Rgraphviz)
> plotNetwork<-function(gR, layouttype="dot", ...){
+     if(!is(gR,"graphNEL")) stop("gR must be a graphNEL object")
+     if(!(GeneNetworkBuilder:::inList(layouttype, c("dot", "neato", "twopi", "circo", "fdp")))){
+         stop("layouttype must be dot, neato, twopi, circo or fdp")
+     }
+     g1<-Rgraphviz::layoutGraph(gR, layoutType=layouttype, ...)
+     nodeRenderInfo(g1)$col <- nodeRenderInfo(gR)$col
+     nodeRenderInfo(g1)$fill <- nodeRenderInfo(gR)$fill
+     renderGraph(g1)
+ }
> plotNetwork(gR)
> ##output the GXL file
> library("XML")
> xml<-saveXML(toGXL(gR)$value())
> z<-textConnection(xml)
> cat(readLines(z, 8), sep="\n")

<?xml version="1.0"?>
<gxl:gxl xmlns:gxl="http://www.gupro.de/GXL/gxl-1.1.dtd">
 <gxl:graph id="graphNEL" edgemode="directed">
  <gxl:node id="zip-4">
   <gxl:attr name="size">
    <gxl:float>72</gxl:float>
   </gxl:attr>
   <gxl:attr name="fill">

> ##calculate shortest path, ...
> library(RBGL)
> sp.between(gR,"daf-16","lam-2")

$`daf-16:lam-2`
$`daf-16:lam-2`$length
[1] 3

$`daf-16:lam-2`$path_detail
[1] "daf-16" "zip-4"  "mir-46" "lam-2"

$`daf-16:lam-2`$length_detail
$`daf-16:lam-2`$length_detail[[1]]
daf-16->zip-4 zip-4->mir-46 mir-46->lam-2
            1             1             1
```
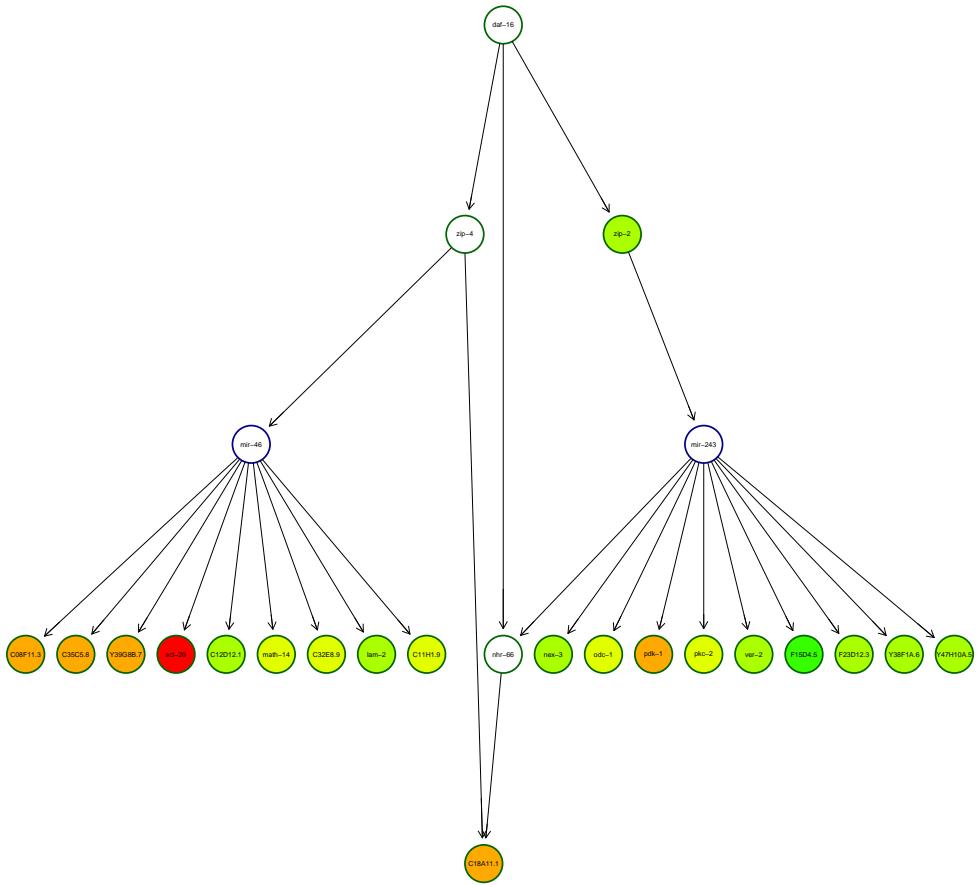
Figure 2: Regulatory network of a *C. elegans* example

2.3 Example using both gene and miRNA expression profile

Using several advanced genomic technologies including micorarray profiling and miRNA sequencing, not only the gene expression profile, but also the miRNA expression profile can be obtained. A more robust network can be built if miRNA expression profiling is available. Here is an example to build *SOX2* response network for *H. sapiens*. The data was downloaded from BMC Genomics[10].

```
> library(GeneNetworkBuilder)
> data("hs.interactionmap")
> data("hs.miRNA.map")
> data("hs.IDsMap")
> data("hs.mapIDs")
> data("example.data")
> rootgene<-"6657"
> sifNetwork<-buildNetwork(example.data$hs.bind, hs.interactionmap, level=5)
> ##example.data$ce$exprData is the combination of gene/miRNA expression profile
> ##note, here should set the miRNAtol to TRUE
```

```
> cifNetwork<-filterNetwork(rootgene=rootgene, sifNetwork=sifNetwork,
+                   exprsData=example.data$hs.exprData, mergeBy="symbols",
+                   miRNAlist=as.character(hs.miRNA.map[,1]),
+                   tolerance=0, miRNAtol=TRUE)
> cifNetwork<-convertID(cifNetwork, hs.mapIDs, ByName=c("from","to"))
> gR<-polishNetwork(cifNetwork)
> ##plot the figure
> library(RCytoscape)
> gR <- initNodeAttribute(gR, "size", "numeric", 36)
> gR <- initNodeAttribute(gR, "fill", "char", "transparent")
> gR <- initNodeAttribute(gR, "borderColor", "char", "black")
> gR <- initEdgeAttribute(gR, "weight", "numeric", 1)
> cw <- new.CytoscapeWindow ('sox2', graph=gR)
> showGraphicsDetails (cw, TRUE)
> displayGraph(cw)
> layoutNetwork (cw, layout.name='force-directed')
> showGraphicsDetails(cw, T)
>  rgb2hex <- function(x){
+               x<-col2rgb(x)
+               unlist(apply(x, 2, function(.ele)
+                                       rgb(.ele[1], .ele[2], .ele[3], maxColorValue=255)))
+ }
> col <- rgb2hex(nodeRenderInfo(gR)$col)
> col <- col[!is.na(col)]
> fill <- rgb2hex(nodeRenderInfo(gR)$fill)
> fill <- fill[!is.na(fill)]
> for(i in 1:length(fill)){setNodeColorDirect(cw, names(fill)[i], fill[i])}
> for(i in 1:length(col)){setNodeBorderColorDirect(cw, names(col)[i], col[i])}
> data <- nodeData(gR)
> for(i in 1:length(data)){
+       setNodeSizeDirect(cw, names(data)[i], data[[i]]$size)
+ }
> redraw (cw)
```

3   References

References

[1] Gene-centered regulatory networks, H. Efsun Arda and Albertha J.M. Walhout, Briefings in Functional Genomics, 9(1): 4-12 (2010)

[2] ChIP-seq: advantages and challenges of a maturing technology, Peter J. Park, Nature Reviews Genetics, 10: 669-680 (2009)

[3] Genome-Wide Analysis of Protein-DNA Interactions, Tae Hoon Kim and Bing Ren, Annual Review of Genomics and Human Genetics, 7: 81-102 (2006)

[4] Chromatin immunoprecipitation (ChIP) of plant transcription factors followed by sequencing (ChIP-SEQ) or hybridization to whole genome arrays (ChIP-CHIP), Kerstin Kaufmann, Jose Muiño, Magne ϕsterås, Laurent Farinelli, Pawel Krajewski and Gerco C Angenent, Nature Protocols, 5: 457-472 (2010)

[5] RNA-Seq: a revolutionary tool for transcriptomics, Zhong Wang, Mark Gerstein and Michael Snyder, Nature Reviews Genetics, 10: 57-63 (2009)
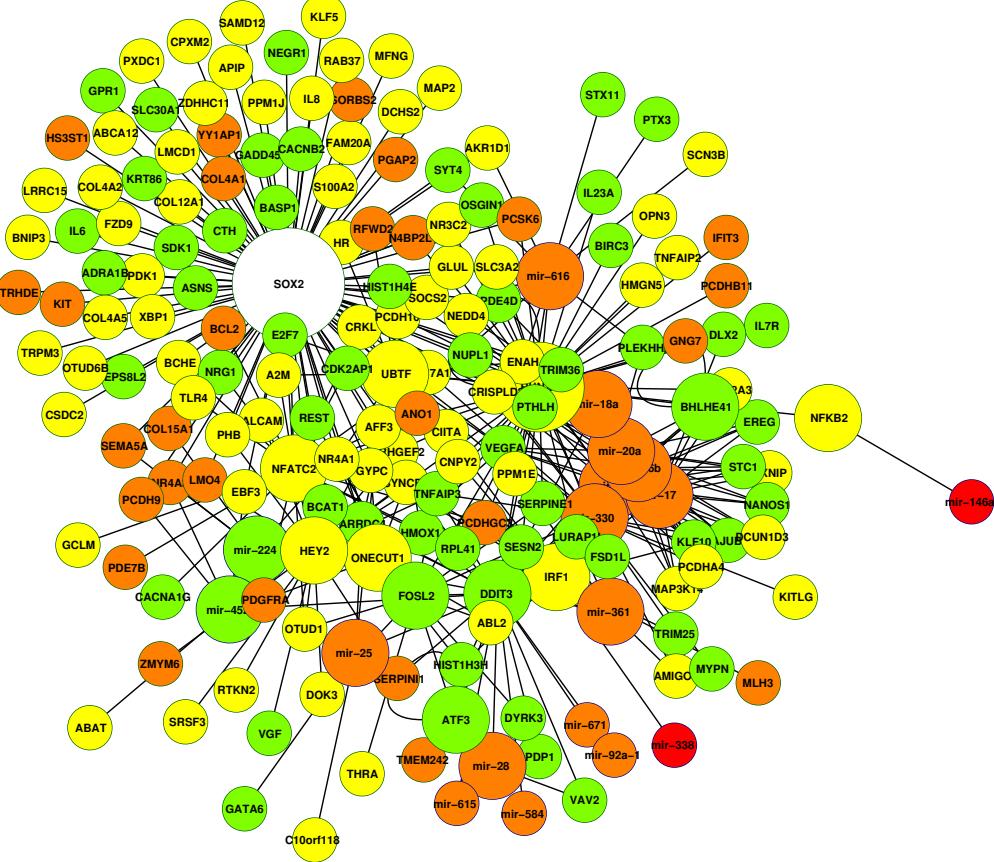
Figure 3: The SOX2 regulatory network.

[6] EDGEdb: a transcription factor-DNA interaction database for the analysis of C. elegans differential gene expression, M Inmaculada Barrasa, Philippe Vaglio, Fabien Cavasino, Laurent Jacotot and Albertha JM Walhout, BMC Genomics, 8:21, (2007)

[7] miRBase: tools for microRNA genomics, Sam Griffiths-Jones, Harpreet Kaur Saini, Stijn van Dongen and Anton J. Enright, Nucleic Acids Research, 36: D153-D158 (2008)

[8] An atlas of combinatorial transcriptional regulation in mouse and man, the FANTOM consortium and RIKEN Omics Science Center, Cell, 140(5):744-752 (2010)

[9] miRGen 2.0: a database of microRNA genomic information and regulation, Panagiotis Alexiou, Thanasis Vergoulis, Martin Gleditzsch, George Prekas, Theodore Dalamagas, Molly Megraw, Ivo Grosse, Timos Sellis and Artemis G. Hatzigeorgiou, Nucleic Acids Research, 38(suppl1):D137-D141, (2010)

[10] The SOX2 response program in glioblastoma multiforme: an integrated ChIP-seq, expression microarray, and microRNA analysis, Xuefeng Fang, Jae-Geun Yoon, Lisha li,

Wei Yu, Jiaofang Shao, Dasong Hua, Shu Zheng, Leroy Hood, David R Goodlett, Gregory Foltz and Biaoyang Lin, BMC Genomics, 12:11, (2011)

4   Session Info

```
> sessionInfo()

R version 3.2.2 (2015-08-14)
Platform: x86_64-pc-linux-gnu (64-bit)
Running under: Ubuntu 14.04.3 LTS

locale:
 [1] LC_CTYPE=en_US.UTF-8       LC_NUMERIC=C
 [3] LC_TIME=en_US.UTF-8        LC_COLLATE=C
 [5] LC_MONETARY=en_US.UTF-8    LC_MESSAGES=en_US.UTF-8
 [7] LC_PAPER=en_US.UTF-8       LC_NAME=C
 [9] LC_ADDRESS=C               LC_TELEPHONE=C
[11] LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C

attached base packages:
[1] grid      stats     graphics  grDevices utils     datasets  methods
[8] base

other attached packages:
[1] RBGL_1.46.0             XML_3.98-1.3
[3] Rgraphviz_2.14.0        GeneNetworkBuilder_1.12.0
[5] graph_1.48.0            Rcpp_0.12.1

loaded via a namespace (and not attached):
[1] plyr_1.8.3        parallel_3.2.2    tools_3.2.2
[4] knitr_1.11        BiocGenerics_0.16.0 stats4_3.2.2
```