vtpnet: variant-transcription factor-phenotype networks

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1 Introduction

In a wide-ranging paper (PMID 22955828 Maurano et al. (2012)), Maurano and colleagues illustrate the concept of "common networks for common diseases" with a bipartite graph. One class of nodes is a set of autoimmune disorders, the other class is a set of transcription factors (TFs). In this graph, an edge exists between a disorder node and a TF node if a SNP that is significantly associated with the risk of the disorder lies in a genomic region possessing a strong match to the binding motif of the TF. This package defines tools to investigate the construction and statistical interpretation of such bipartite graphs, which we will denote VTP (variant-transcription factor-phenotype) networks.

2 Illustrative example of an unpruned VTP

The following code uses the `graphNEL` class to construct an approximation to the complete bipartite graph underlying Figure 4A of the Maurano paper; Figure 1 illustrates an arbitrary complete subgraph. The elements of `diseaseTags` are formatted to allow multiline rendering of the strings in node displays. It will be useful to distinguish a display token type and an analysis token type to simplify programming.

```
> #
> # tags formatted for display
> #
> diseaseTags = c("Ankylosing\\\nspondylitis", "Asthma",
+       "Celiac\\\ndisease", "Crohn's\\\ndisease",
+       "Multiple\\\nsclerosis", "Primary\\\nbiliary\\\ncirrhosis",
+       "Psoriasis", "Rheumatoid\\\narthritis",
+       "Systemic\\\nlupus\\\nerythematosus",
+       "Systemic\\\nsclerosis", "Type 1\\\ndiabetes",
```

```
+       "Ulcerative\\\ncolitis"
+ )
> TFtags = c("ELF3", "MEF2A", "TCF3", "PAX4", "STAT3",
+    "ESR1", "POU2F1", "STAT1", "YY1", "SP1", "CDC5L",
+    "NR3C1", "EGR1", "PPARG", "HNF4A", "REST", "PPARA",
+    "AR", "NFKB1", "HNF1A", "TFAP2A")
> # define adjacency matrix
> adjm = matrix(1, nr=length(diseaseTags), nc=length(TFtags))
> dimnames(adjm) = list(diseaseTags, TFtags)
> library(graph)
> cvtp = ugraph(aM2bpG(adjm)) # complete (V)TP network; variants not involved yet
```

3 Data on GWAS variants: their associated phenotype, locations, and other characteristics

We will use the GWAS data provided at `https://www.sciencemag.org/content/suppl/2012/09/04/science.1222794.DC1/1222794-Maurano-tableS2.txt`, which was manually imported to a GRanges instance in hg19 origin-1 coordinates.

```
> library(vtpnet)
> data(maurGWAS)
> length(maurGWAS)

[1] 5654

> names(values(maurGWAS))

[1] "name"                     "disease_trait"
[3] "disease_class"            "internally_replicated"
[5] "independently_replicated" "In_DHS"
[7] "fetal_origin"             "X.LOG.P."
[9] "sample_size"
```

4 Data on transcription factor binding sites

We have included the result of using FIMO Grant et al. (2011) to scan for motif matches for TF PAX4 as modeled in the Bioconductor *MotifDb* collection. The `-max-stored-scores` parameter was set to 10000000 so that p of up to 10^{-4} are retained.

```
> data(pax4)
> length(pax4)
```

```
> library(Rgraphviz)
> #flat = function(x, g) c(x, edges(g)[[x]])
> #sub = subGraph(unique(c(flat("Crohn's\\\ndisease", cvtp),
> #    flat("Ulcerative\\\ncolitis", cvtp))), cvtp)
> sub = subGraph(unique(c(diseaseTags[1:4], TFtags[1:6])), cvtp)
> plot(sub, attrs=list(node=list(shape="box", fixedsize=FALSE)))
> #plot(cvtp, attrs=list(graph=list(margin=c(.5,.5), size=c(4.1,4.1)),
> #   node=list(shape="box", fixedsize=FALSE, height=1)))
```

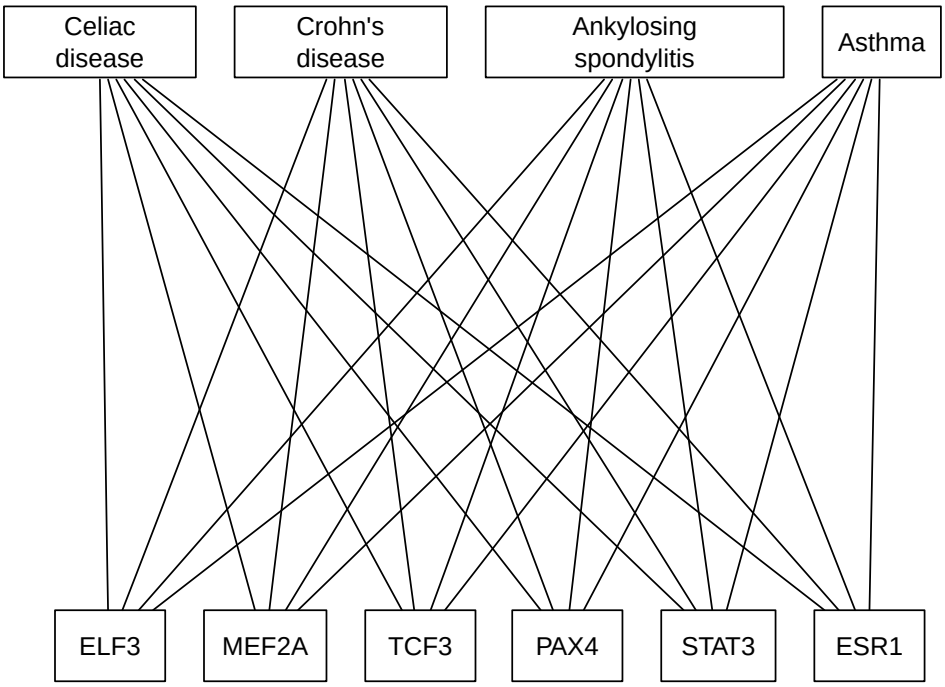

Figure 1: A complete bipartite graph for arbitrarily selected subsets of the autoimmune disorders and TFs found in Figure 4A of Maurano et al.

```
[1] 1862156
```

```
> pax4[1:4]
```

```
GRanges object with 4 ranges and 8 metadata columns:
      seqnames      ranges strand |   source             type     score
         <Rle>   <IRanges>  <Rle> | <factor>         <factor> <numeric>
  [1]     chr1 10273-10302      + |     fimo nucleotide_motif   999.917
  [2]     chr1 10279-10308      + |     fimo nucleotide_motif   999.962
  [3]     chr1 11703-11732      - |     fimo nucleotide_motif   999.999
  [4]     chr1 11704-11733      - |     fimo nucleotide_motif   999.955
         phase                  Name      pvalue      qvalue
      <integer>           <character> <character> <character>
  [1]      <NA> +Mmusculus-JASPAR_CO..    8.35e-05       0.396
  [2]      <NA> +Mmusculus-JASPAR_CO..    3.79e-05       0.361
  [3]      <NA> -Mmusculus-JASPAR_CO..    8.04e-07       0.194
  [4]      <NA> -Mmusculus-JASPAR_CO..    4.46e-05       0.368
                  sequence
                <character>
  [1] TAACCCTAACCCTAACCCCA..
  [2] TAACCCTAACCCCAACCCCA..
  [3] AAAAAAATACACATGGCCAG..
  [4] TAAAAAAATACACATGGCCA..
  -------
  seqinfo: 92 sequences from an unspecified genome; no seqlengths
```

We can also generate our own motif-match ranges. Here is an example of a parallelized search against hg19 using `matchPWM`.

```
> library(foreach)
> library(doParallel)
> registerDoParallel(cores=12)
> library(BSgenome.Hsapiens.UCSC.hg19)
> library(MotifDb)
> sn = seqnames(Hsapiens)[1:24]
> pax4 = query(MotifDb, "pax4")[[1]]
> ans = foreach(i=1:24) %dopar% {
+  cat(i)
+  subj = Hsapiens[[ sn[i] ]]
+  matchPWM( pax4, subj, "75%" )
+ }
> pax4_75 =
+  do.call(c, lapply(1:length(ans), function(x)
```

```
+   {GRanges(sn[x], as(ans[[x]], "IRanges"))}))
> save(pax4_75, file="pax4_75.rda")
```

Results of such searches retaining matches at scores of 85% and 75% of the maximum achievable score have been stored with this package.

5 Building a VTP network: one edge per phenotype

5.1 Raw matches

We can survey the entire GWAS catalog for intersection with putative PAX4 binding sites. First the two Bioconductor internal binding site sets.

```
> data(pax4_85)
> vp_pax4_85 = maurGWAS[ overlapsAny(maurGWAS, pax4_85) ]
> length(vp_pax4_85)
```

```
[1] 0
```

```
> data(pax4_75)
> vp_pax4_75 = maurGWAS[ overlapsAny(maurGWAS, pax4_75) ]
> length(vp_pax4_75)
```

```
[1] 54
```

Then the FIMO-based set.

```
> vp_pax4_fimo = maurGWAS[ overlapsAny(maurGWAS, pax4) ]
> length(vp_pax4_fimo)
```

```
[1] 67
```

The lengths reported here are the numbers of phenotypes linked to PAX4 in a VTP according to various motif matching schemes. For the two non-null results, we have

```
> u75 = unique(vp_pax4_75$disease_trait)
> ufimo = unique(vp_pax4_fimo$disease_trait)
> length(setdiff(u75, ufimo))
```

```
[1] 23
```

```
> length(setdiff(ufimo, u75))
```

```
[1] 28
```

Clearly the identification of TP links is sensitive to the approach used to locate binding sites. However, as noted in the Maurano paper, the use of matching to the reference genome without SNP injection is potentially problematic.

5.2 Filtering

It is useful to restrict the phenotypes of interest, and to map them to broader classes, and to include TFBS matching scores for the purpose of filtering edges. Here we will use the NHGRI GWAS catalog against FIMO-based (reference genome matching only) PAX4 calls.

```
> data(cancerMap)
> requireNamespace("gwascat")
> load(system.file("legacy/gwrngs19.rda", package="gwascat"))
> cangw = filterGWASbyMap( gwrngs19, cancerMap )
> getOneHits( pax4, cangw, "fimo" )

GRanges object with 8 ranges and 41 metadata columns:
        seqnames    ranges strand | Date.Added.to.Catalog  PUBMEDID
           <Rle> <IRanges>  <Rle> |            <character> <integer>
   3475     chr8 129194641      * |             09/12/2013  23535729
   3480    chr11  65583066      * |             09/12/2013  23535729
   6963     chr2  26526419      * |             01/25/2013  23144319
   7155     chr6 143943314      * |             01/15/2013  23108145
   7480    chr20  32588095      * |             11/30/2012  22976474
  12585     chrX  37854727      * |             11/15/2010  20932654
  13650    chr12  14653867      * |             07/12/2010  20543847
  15145    chr10  63752159      * |             09/04/2009  19684604
         First.Author        Date                Journal
          <character> <character>            <character>
   3475  Michailidou K  04/01/2013              Nat Genet
   3480  Michailidou K  04/01/2013              Nat Genet
   6963          Lee Y  11/08/2012         Carcinogenesis
   7155        Wang LE  10/29/2012             Cancer Res
   7480       Siddiq A  09/13/2012           Hum Mol Genet
  12585       Kerns SL  10/05/2010 Int J Radiat Oncol B..
  13650     Turnbull C  06/13/2010              Nat Genet
  15145 Papaemmanuil E  08/16/2009              Nat Genet
                        Link                  Study          Disease.Trait
                  <character>            <character>            <character>
   3475 http://www.ncbi.nlm... Large-scale genotypi..          Breast cancer
   3480 http://www.ncbi.nlm... Large-scale genotypi..          Breast cancer
   6963 http://www.ncbi.nlm... Prognostic implicati.. Non-small cell lung ..
   7155 http://www.ncbi.nlm... Genome-wide associat.. Lung Cancer (DNA rep..
   7480 http://www.ncbi.nlm... A meta-analysis of g..          Breast cancer
  12585 http://www.ncbi.nlm... Genome-wide associat.. Erectile dysfunction..
  13650 http://www.ncbi.nlm... Variants near DMRT1,.. Testicular germ cell..
```

```
15145 http://www.ncbi.nlm... Loci on 7p12.2, 10q2.. Acute lymphoblastic ..
        Initial.Sample.Size Replication.Sample.Size      Region      Chr_id
                <character>             <character> <character> <character>
 3475 10,052 European ance..  45,290 European ance..     8q24.21           8
 3480 10,052 European ance..  45,290 European ance..     11q13.1          11
 6963 348 Korean ancestry ..                     NR      2p23.3           2
 7155 914 European ancestr..  679 European ancestr..      6q24.2           6
 7480 3,666 European ances..  562 European ancestr..    20q11.22          20
12585 27 African American ..                    <NA>      Xp11.4          23
13650 979 European ancestr..  664 European ancestr..     12p13.1          12
15145 907 European ancestr..                    <NA>     10q21.2          10
      Chr_pos.hg38       Reported.Gene.s.         Mapped_gene
         <numeric>             <character>         <character>
 3475    128182395           MIR1208, MYC MIR1208 - LINC01263
 3480     65815595 DKFZp761E198, OVOL1,..   OVOL1-AS1 - SNX32
 6963     26303551                 GPR113      HADHB - GPR113
 7155    143622177               PHACTR2             PHACTR2
 7480     34000289     RALY, EIF2S2, ASIP                RALY
12585     37995474                  SYTL5     CXorf27 - SYTL5
13650     14500933                 ATF7IP              ATF7IP
15145     61992400                 ARID5B              ARID5B
      Upstream_gene_id Downstream_gene_id Snp_gene_ids Upstream_gene_distance
           <character>        <character>  <character>            <character>
 3475        100302281          101927774                               32.21
 3480        101927828             254122                               24.73
 6963             3032             165082                               13.09
 7155             <NA>               <NA>        9749                    <NA>
 7480             <NA>               <NA>       22913                    <NA>
12585            25763              94122                               4.16
13650             <NA>               <NA>       55729                    <NA>
15145             <NA>               <NA>       84159                    <NA>
      Downstream_gene_distance Strongest.SNP.Risk.Allele        SNPs
                   <character>                <character> <character>
 3475                   222.87              rs11780156-T  rs11780156
 3480                    18.24               rs3903072-G   rs3903072
 6963                     4.62               rs6753473-G   rs6753473
 7155                     <NA>               rs9390123-A   rs9390123
 7480                     <NA>               rs2284378-T   rs2284378
12585                    11.11                rs872690-?    rs872690
13650                     <NA>               rs2900333-C   rs2900333
15145                     <NA>               rs7089424-C   rs7089424
          Merged Snp_id_current     Context  Intergenic
```

```
      <character>    <character> <character> <character>
 3475           0       11780156  Intergenic           1
 3480           0        3903072  Intergenic           1
 6963           0        6753473  Intergenic           1
 7155           0        9390123      intron           0
 7480           0        2284378      intron           0
12585           0         872690  Intergenic           1
13650           0        2900333       UTR-3           0
15145           0        7089424      intron           0
      Risk.Allele.Frequency   p.Value Pvalue_mlog   p.Value..text. OR.or.beta
                <character> <numeric>   <numeric>      <character>  <numeric>
 3475                  0.16     3e-11    10.52288                         1.07
 3480                  0.53     9e-12    11.04576                         1.05
 6963                 0.052     4e-06     5.39794 (Additive model)          NA
 7155                0.3957     7e-06     5.15490                          NA
 7480                  0.31     1e-08     8.00000                         1.16
12585                  0.03     9e-06     5.04576                        11.78
13650                  0.62     6e-10     9.22185                         1.27
15145                  0.34     7e-19    18.15490                         1.65
      X95..CI..text. Platform..SNPs.passing.QC.         CNV
         <character>                <character> <character>
 3475   [1.04-1.10]      Illumina & Affymetri..           N
 3480   [1.04-1.08]      Illumina & Affymetri..           N
 6963            NR       Affymetrix [271,817]           N
 7155            NR         Illumina [303,669]           N
 7480   [1.10-1.22]      Illumina [2,608,509]..           N
12585          [NR]        Affymetrix [512,497]           N
13650   [1.12-1.44]          Illumina [298,782]           N
15145   [1.54-1.76]          Illumina [291,473]           N
      num.Risk.Allele.Frequency      dclass     score   tfstart     tfend
                      <numeric> <character> <numeric> <integer> <integer>
 3475                     0.1600      Breast   999.985 129194621 129194650
 3480                     0.5300      Breast   999.952  65583065  65583094
 6963                     0.0520        Lung   999.987  26526415  26526444
 7155                     0.3957        Lung   999.939 143943292 143943321
 7480                     0.3100      Breast   999.928  32588075  32588104
12585                     0.0300    Prostate   999.903  37854721  37854750
13650                     0.6200  Testicular   999.990  14653848  14653877
15145                     0.3400   ALL (ped)   999.962  63752142  63752171
        pvalue    qvalue
      <numeric> <numeric>
 3475  1.49e-05     0.318
```

```
 3480  4.83e-05     0.373
 6963  1.25e-05     0.310
 7155  6.13e-05     0.383
 7480  7.16e-05     0.388
12585  9.72e-05     0.403
13650  1.05e-05     0.301
15145  3.79e-05     0.361
-------
seqinfo: 23 sequences from hg19 genome
```

6 Appendix: generating the ALT-injected genome image

```
> altize = function(chtag = "21",
+ #
+ # from sketch by Herve Pages, May 2013
+ #
+   slpack="SNPlocs.Hsapiens.dbSNP.20120608",
+   hgpack ="BSgenome.Hsapiens.UCSC.hg19",
+   faElFun = function(x) sub("%%TAG%%", x, "alt%%TAG%%chr"),
+   faTargFun = function(x)
+      sub("%%TAG%%", x, "alt%%TAG%%_hg19.fa")) {
+     require(slpack, character.only=TRUE)
+     require(hgpack, character.only=TRUE)
+     require("ShortRead", character.only=TRUE)
+     chk = grep("ch|chr", chtag)
+     if (length(chk)>0) {
+       warning("clearing prefix ch or chr from chtag")
+       chtag = gsub("ch|chr", "", chtag)
+       }
+     snpgettag = paste0("ch", chtag)
+     ggettag = paste0("chr", chtag)
+     cursnps = getSNPlocs(snpgettag, as.GRanges=TRUE)
+     curgenome = unmasked(Hsapiens[[ggettag]])
+     ref_allele =
+      strsplit(as.character(curgenome[start(cursnps)]),
+         NULL, fixed=TRUE)[[1L]]
+     all_alleles = IUPAC_CODE_MAP[cursnps$alleles_as_ambig]
+     alt_alleles = mapply( function(ref,all)
+       sub(ref, "", all, fixed=TRUE),
+         ref_allele, all_alleles, USE.NAMES=FALSE)
+     cursnps$ref_allele = ref_allele
+     cursnps$alt_alleles = alt_alleles
```

```
+     cursnps$one_alt = substr(cursnps$alt_alleles, 1, 1)
+     altg = list(replaceLetterAt(curgenome, start(cursnps),
+       cursnps$one_alt))
+     names(altg) = faElFun(chtag)
+     writeFasta(DNAStringSet(altg), file=faTargFun(chtag))
+ }
```

7 Session information

```
> sessionInfo()

R version 4.1.1 (2021-08-10)
Platform: x86_64-pc-linux-gnu (64-bit)
Running under: Ubuntu 20.04.3 LTS

Matrix products: default
BLAS:   /home/biocbuild/bbs-3.14-bioc/R/lib/libRblas.so
LAPACK: /home/biocbuild/bbs-3.14-bioc/R/lib/libRlapack.so

locale:
 [1] LC_CTYPE=en_US.UTF-8       LC_NUMERIC=C
 [3] LC_TIME=en_GB              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] parallel  stats4    grid      stats     graphics  grDevices utils
 [8] datasets  methods   base

other attached packages:
 [1] vtpnet_0.34.0        doParallel_1.0.16    iterators_1.0.13
 [4] foreach_1.5.1        gwascat_2.26.0       GenomicRanges_1.46.0
 [7] GenomeInfoDb_1.30.0  IRanges_2.28.0       S4Vectors_0.32.0
[10] Rgraphviz_2.38.0     graph_1.72.0         BiocGenerics_0.40.0

loaded via a namespace (and not attached):
 [1] MatrixGenerics_1.6.0        Biobase_2.54.0
 [3] httr_1.4.2                  splines_4.1.1
 [5] bit64_4.0.5                 assertthat_0.2.1
 [7] BiocFileCache_2.2.0         blob_1.2.2
```

```
 [9] BSgenome_1.62.0             GenomeInfoDbData_1.2.7
[11] Rsamtools_2.10.0            yaml_2.2.1
[13] progress_1.2.2              pillar_1.6.4
[15] RSQLite_2.2.8               lattice_0.20-45
[17] glue_1.4.2                  digest_0.6.28
[19] XVector_0.34.0              Matrix_1.3-4
[21] XML_3.99-0.8                pkgconfig_2.0.3
[23] biomaRt_2.50.0              zlibbioc_1.40.0
[25] purrr_0.3.4                 tzdb_0.1.2
[27] BiocParallel_1.28.0         tibble_3.1.5
[29] KEGGREST_1.34.0             generics_0.1.1
[31] ellipsis_0.3.2              cachem_1.0.6
[33] SummarizedExperiment_1.24.0 GenomicFeatures_1.46.0
[35] survival_3.2-13             magrittr_2.0.1
[37] crayon_1.4.1                memoise_2.0.0
[39] fansi_0.5.0                 xml2_1.3.2
[41] tools_4.1.1                 prettyunits_1.1.1
[43] hms_1.1.1                   BiocIO_1.4.0
[45] lifecycle_1.0.1             matrixStats_0.61.0
[47] stringr_1.4.0               DelayedArray_0.20.0
[49] snpStats_1.44.0             AnnotationDbi_1.56.0
[51] Biostrings_2.62.0           compiler_4.1.1
[53] rlang_0.4.12                RCurl_1.98-1.5
[55] rstudioapi_0.13             VariantAnnotation_1.40.0
[57] rjson_0.2.20                rappdirs_0.3.3
[59] bitops_1.0-7                codetools_0.2-18
[61] restfulr_0.0.13             DBI_1.1.1
[63] curl_4.3.2                  R6_2.5.1
[65] GenomicAlignments_1.30.0    dplyr_1.0.7
[67] rtracklayer_1.54.0          fastmap_1.1.0
[69] bit_4.0.4                   utf8_1.2.2
[71] filelock_1.0.2              readr_2.0.2
[73] stringi_1.7.5               Rcpp_1.0.7
[75] vctrs_0.3.8                 png_0.1-7
[77] dbplyr_2.1.1                tidyselect_1.1.1
```

8 Bibliography

References

Charles E Grant, Timothy L Bailey, and William Stafford Noble. Fimo: scanning for occurrences of a given motif. *Bioinformatics (Oxford, England)*, 27(7):1017–8, Apr 2011. doi: 10.1093/bioinformatics/btr064.

Matthew T Maurano, Richard Humbert, Eric Rynes, Robert E Thurman, Eric Haugen, Hao Wang, Alex P Reynolds, Richard Sandstrom, Hongzhu Qu, Jennifer Brody, Anthony Shafer, Fidencio Neri, Kristen Lee, Tanya Kutyavin, Sandra Stehling-Sun, Audra K Johnson, Theresa K Canfield, Erika Giste, Morgan Diegel, Daniel Bates, R Scott Hansen, Shane Neph, Peter J Sabo, Shelly Heimfeld, Antony Raubitschek, Steven Ziegler, Chris Cotsapas, Nona Sotoodehnia, Ian Glass, Shamil R Sunyaev, Rajinder Kaul, and John A Stamatoyannopoulos. Systematic localization of common disease-associated variation in regulatory dna. *Science*, 337(6099):1190–5, Sep 2012. doi: 10.1126/science.1222794.