annotation tutorial

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0.1 Example for creating biological annotations for BANNs

BANNs framework requires constructing an annotation mask in order to guide the model structure based on biological annotations. We create SNP-sets based on the genomic annotations of SNPs used in the model. SNPs that fall into the neighborhood of the same gene are grouped together into a SNP-set that corresponds to this gene. If we consider intergenic regions (based on user input), SNPs that fall into unannotated regions of the genome are grouped together into intergenic regions. For example, two SNPs that fall into the intergenic region between $gene\ A$ and $gene\ b$ would be grouped together into a intergenic region called $Intergenic_gene\ A_gene\ B$.

In this example, we show how to carry out the annotation and create mask matrices (and annotation dataframes that store information on which SNPsets contain which SNPs) using the annotation.py script provided.

We need two inputs for carrying out biological annotations:

- 1) A SNP list file of .map or .bim format
- 2) A gene range file

SNP list file: We expect .map formatted SNP lists to be tab-delimited files have four fields/columns (with no header): Chromosome, Variant ID, Morgans, Position(bp). Our annotation script works with position information. If morgan information is missing, you can replace this column with 0s (zeros) or leave it empty. We expect a similar format from .bim files but with six fields/columns: Chromosome, Variant ID, Morgans, Position, Minor Allele, Major Allele. If the files contain a different number of columns, annotation will not carry out and you will receive a warning about the error.

To get more information on file formats, you can visit: https://www.cog-genomics.org/plink2/formats#map https://www.cog-genomics.org/plink2/formats#bim

Important Note: In order to obtain accurate results, please make sure the order of SNPs in the SNP List file correspond to the order of SNPs in the genotype matrix fed into the model because indices of variables matter for the model and annotations. This means if variant rs7412 is the first SNP in the genotype matrix (i.e. the first column of the genotype matrix corresponds to variant rs7412), then the first row of the SNP list should have information on rs7412.

Gene range file: We expect gene range file to be a tab-delimited file with four fields/columns (with no header): Chromosome, Start, End, GeneID. We expect that the Start and End coordinates will be in basepaires. To get more information on this file format, you can visit the "Gene range lists" section in:

https://www.cog-genomics.org/plink/1.9/resources and download gene range files for human genome.

Example Files In this tutorial, we will work with the very small SNP list and gene range lists provided in /sampleData/TestSNPList.txt and /sampleData/TestGeneList.txt files for the sake of demonstration.

Here is what these files look like:

```
[1]: SNPList_file = open("sampleData/TestSNPList.txt", "r")
    GeneList_file = open("sampleData/TestGeneList.txt", "r")

print("Printing SNP list file contents (.map formatted): \n")
print(SNPList_file.read())
print("\n\n Printing Gene range list file contents: \n")
print(GeneList_file.read())
```

Printing SNP list file contents (.map formatted):

```
2
         rs2
                  0
                            11
2
                  0
                            2
         rs1
18
         rs3
                  0
                            13
18
         rs4
                  0
                            1021
18
         rs5
                  0
                            1800
19
         rs5
                  0
                            609
                  0
19
         rs6
                            5227
19
                  0
         rs7
                            10187
19
         rs8
                  0
                            12148
         rs9
                  0
Х
                            1
Х
         rs10
                  0
                            392
Х
                  0
         rs11
                            1107
X
         rs12
                  0
                            4331
```

Printing Gene range list file contents:

18	1023	1803	Gene4
7	1267	9569	Gene2
19	9532	10187	Gene7
Х	1	501	Gene8
19	587	791	Gene6
18	240	391	Gene3
Х	2675	5092	Gene9
Х	4061	9582	Gene10
18	21200	29080	Gene5
7	241	905	Gene1

0.1.1 Carrying out annotations

In this tutorial, we will create two files:

1) an annotation dataframe that contains information on which SNP-sets contain which SNPs and what the genomic coordinate of these SNP-sets are, saved in a tab-delimited .txt file.

This dataframe is sorted in ascending order based on the chromosomal location of the SNP-sets. Carried out by the *annotate()* function

2) an annotation mask matrix, which is a sparse matrix of 0s and 1s used to guide the model architecture, saved in a tab-delimited .txt file.

This is a matrix of size: (number of SNPs by number of SNPsets). An entry in the matrix at location (row i, column j) tells us whether SNP i is contained in SNP-set j (yes if 1, no if 0).

The rows of the mask matrix (which correspond to SNPs) are in the same order as the SNP list file, meaning the first row contains annotation information of the first SNP in the SNP list. The columns of the mask matrix (which correspond to SNP-set) are in the same order as the annotation dataframe, so columns correspond to SNP-sets in an ascending order based on chromosomal location. Carried out by the *qetMaskMatrix()* function

Parameters There are a few parameters we require from the user to specify how to carry out the annotations and save the results, apart from giving the path to the input SNP list file and gene range list file:

output file: Both annotate() and getMaskMatrix() functions require us to specify this. It is supposed to be path to the .txt file where we want to save the results (either annotation dataframe or mask matrix).

intergenic: For *annotate()* function. This is a boolean parameter, that expects either of the True or False values. The default value is True.

If *True*, the annotation script creates intergenic SNP-sets and considers them when annotating the SNPs. If *False*, it simply groups SNPs into genes defined in the gene range list and all SNPs that are unannotated (in intergenic regions) are grouped into one SNP-set called "Unannotated". This "Unannotated" SNP-set is the last entry of the annotation dataframe and would be the last column of the mask matrix to be created from the annotation dataframe. If the gene range list provided by the user has all chromosomes, there will be no "Unannotated" SNP-set when this parameter is set to *True*. However, if there are chromosomes missing (as in the case of this example, just for demonstrations), SNPs in the missing chromosomes will still be grouped into the "Unannotated" SNP-set.

buffer: For annotate() function. This is an integer and the default value is 0.

It tells the annotation script how many basepairs to allow for as a buffer when considering a SNP-set neighborhood. If set to, let's say, 50000, then we would allow a 50kb window around SNP-sets when considering whether a SNP is in the SNP-set or not.

dropSingletons: For *annotate()* function. This is a boolean parameter, that expects either of the True or False values. The default value is False.

When annotating, it is possible that some SNP-sets will only contain one SNP. We call these SNP-sets "singletons". In this case, whether they should still be considered a "SNP-set" is debatable. If you set this parameter value to *True*, then these SNP-sets will be dropped from the annotation, and SNPs will be re-annotated with the remaining SNP-sets. Otherwise (if set at *False*), singleton SNP-sets will be kept in the model.

Creating the annotation dataframe:

0

[rs4]

```
[2]: # Importing annotation.py from the src folder:
     import sys
     sys.path.insert(0, '.../src/') #Need to do this for this notebook example since
     → the code lives in a different directory
     from annotation import * # Import annotation code
     #Changing back to the original directory:
     sys.path.insert(0, '../examples_docs/')
     #Specifying the path to the SNP list file and gene range list file:
     path_to_SNPList="sampleData/TestSNPList.txt"
     path_to_geneList="sampleData/TestGeneList.txt"
     #Specifying the path to the file we want the save annotation dataframe into:
     file_toSave_annotationDF="sampleData/TestAnnotationDF.txt"
     #Carrying out the annotation:
     annotationDF=annotate(path_to_SNPList, path_to_geneList,_
      →outputFile=file toSave annotationDF,
                           intergenic=True, buffer=0, dropSingletons=False)
     print(annotationDF)
    100%|
               | 10/10 [00:00<00:00, 241.81it/s]
    100%|
               | 13/13 [00:00<00:00, 689.29it/s]
    You have chosen to annotate SNP-sets with intergenic regions and with a buffer
    of Obp
    Creating Intergenic SNP-sets
    Annotating SNP-sets with the corresponding SNPs
    Saving annotation results to file sampleData/TestAnnotationDF.txt
                        GeneID Chromosome
                                              Start
                                                         End
                                                                SNPindex \
        Intergenic_Gene3_Gene4
                                              392.0
                                                      1022.0
                                                                     [3]
    0
                                                                     [4]
    1
                         Gene4
                                        18
                                             1023.0
                                                      1803.0
    2
                         Gene6
                                        19
                                              587.0
                                                       791.0
                                                                     [5]
    3
                                              792.0
                                                      9531.0
                                                                     [6]
        Intergenic_Gene6_Gene7
                                        19
    4
                          Gene7
                                        19
                                             9532.0 10187.0
                                                                     [7]
    5
              Downstream_Gene7
                                        19
                                           10188.0 12148.0
                                                                     [8]
    6
                                                                 [9, 10]
                          Gene8
                                         X
                                                1.0
                                                       501.0
    7
                                         Х
                                              502.0
                                                      2674.0
                                                                    [11]
        Intergenic_Gene8_Gene9
    8
                         Gene9
                                                      5092.0
                                         X
                                             2675.0
                                                                    [12]
    9
                         Gene10
                                         Х
                                             4061.0
                                                      9582.0
                                                                    [12]
    10
                   UnAnnotated
                                       NaN
                                                NaN
                                                              [1, 0, 2]
                                                         NaN
              VariantID
```

```
[rs5]
1
2
                [rs5]
3
                [rs6]
4
                [rs7]
5
                [rs8]
         [rs9, rs10]
6
7
               [rs11]
               [rs12]
8
               [rs12]
   [rs1, rs2, rs3]
```

Showing annotation results with different parameters:

```
[3]: #Carrying out the annotation:
annotationDF=annotate(path_to_SNPList, path_to_geneList,
outputFile=file_toSave_annotationDF,
intergenic=False, buffer=500, dropSingletons=True)

print(annotationDF)
```

```
100% | 13/13 [00:00<00:00, 859.38it/s]
100% | 13/13 [00:00<00:00, 838.38it/s]
```

You have chosen to annotate SNP-sets without intergenic regions and with a buffer of $500\mbox{bp}$

Annotating SNP-sets with the corresponding SNPs

Dropping SNP-sets that are singletons (containing only one SNP) and reannotating SNPs without them

Annotating SNP-sets with the corresponding SNPs

Saving annotation results to file sampleData/TestAnnotationDF.txt

	${\tt GeneID}$	Chromosome	Start	End	$SNPindex \setminus$
0	Gene3	18	240.0	391.0	[2]
1	Gene4	18	1023.0	1803.0	[3, 4]
2	Gene6	19	587.0	791.0	[5]
3	Gene7	19	9532.0	10187.0	[7]
4	Gene8	X	1.0	501.0	[9, 10]
5	Gene9	X	2675.0	5092.0	[12]
6	Gene10	X	4061.0	9582.0	[12]
7	UnAnnotated	NaN	NaN	NaN	[1, 0, 6, 8, 11]

```
VariantID
0 [rs3]
1 [rs4, rs5]
2 [rs5]
3 [rs7]
4 [rs9, rs10]
5 [rs12]
```

```
6 [rs12]
7 [rs1, rs2, rs6, rs8, rs11]
```

Creating the mask matrix from the annotation dataframe:

[4]: mask_outputFile="sampleData/TestMask.txt"
mask = getMaskMatrix(path_to_SNPList, annotationDF, mask_outputFile)
print(mask)

100%| | 8/8 [00:00<00:00, 3834.79it/s]

creating mask

Saving annotation mask to file sampleData/TestMask.txt in tab-delimited format

- [[0. 0. 0. 0. 0. 0. 0. 1.]
- [0. 0. 0. 0. 0. 0. 0. 1.]
- [1. 0. 0. 0. 0. 0. 0. 0.]
- [0. 1. 0. 0. 0. 0. 0. 0.]
- [0. 1. 0. 0. 0. 0. 0. 0.]
- [0. 0. 1. 0. 0. 0. 0. 0.]
- [0. 0. 0. 0. 0. 0. 1.]
- $[0. \ 0. \ 0. \ 1. \ 0. \ 0. \ 0. \ 0.]$
- [0. 0. 0. 0. 0. 0. 1.]
- [0. 0. 0. 0. 1. 0. 0. 0.]
- [0. 0. 0. 0. 1. 0. 0. 0.] [0. 0. 0. 0. 0. 0. 0. 1.]
- [0. 0. 0. 0. 0. 1. 1. 0.]]