AlphaBeta

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1 Introduction

AlphaBeta is a computational method for estimating epimutation rates and spectra from high-throughput DNA methylation data in plants.

The method has been specifically designed to:

- 1. Analyze 'germline' epimutations in the context of multi-generational mutation accumulation lines (MA-lines).
- 2. Analyze 'somatic' epimutations in the context of plant development and aging.

Heritable changes in cytosine methylation can arise stochastically in plant genomes independently of DNA sequence alterations. These so-called 'spontaneous epimutations' appear to be a byproduct of imperfect DNA methylation maintenance during mitotic and meitotic cell divisions.

Accurate estimates of the rate and spectrum of these stochastic events are necessary to be able to quantify how epimutational processes shape methylome diversity in the context of plant evolution, development and aging.

Here we describe AlphaBeta, a computational method for estimating epimutation rates and spectra from pedigree-based high-throughput DNA methylation data in plants.

The method requires that the topology of the pedigree is known, which is typically the case in the construction of mutation accumulation lines (MA-lines) in sexually or clonally reproducing plant species.

However, the method also works for inferring somatic epimutations in long-lived perrenials, such as trees, using leaf methylomes and coring data as input. In this case, AlphaBeta treats the tree branching structure as an intra-organismal phylogeny of somatic lineages that carry information about the epimutational history of each branch.

2 Preparing Files

2.1 Generation file

A file containing the list of filenames should be provided for generation of a divergence matrix and calculation of methylation proportions.

```
# SAMPLE FILE
generation.fn <- system.file("extdata", "generations.fn", package = "AlphaBeta")
file <- fread(generation.fn)
head(file)</pre>
```

```
filename generation lineage

1: data/methylome_Col_GO-merged.txt GO

2: data/methylome_Col_G1_L2-merged.txt G1 L2

3: data/methylome_Col_G4_L8-merged.txt G4 L8

4: data/methylome_Col_G11_L2-merged.txt G11 L2
```

2.2 Generate divergence matrix

Estimating epimutation rates from high-throughput DNA methylation data. Generation of divergence matrix and calculation of methylation levels.

```
dMatrix(genTable = generation.fn, cytosine = "CG", posteriorMaxFilter = 0.99)
# Sample output from dMatrix function
head(fread("AB-dMatrix-CG-0.99.csv"))
```

```
pair.1 pair.2 D.value
1: G0 G1-L2 0.01366
2: G0 G4-L8 0.01412
3: G0 G11-L2 0.00806
4: G1-L2 G4-L8 0.03265
5: G1-L2 G11-L2 0.00473
6: G4-L8 G11-L2 0.00904
```

2.3 Generate methylation proportions

```
rc.meth.lvl(genTable = generation.fn, cytosine = "CG", posteriorMaxFilter = 0.99,
   nThread = 4)
# Sample output from proportions function
head(fread(system.file("extdata/dm", "AB-methprop-CG-0.99.csv",
   package = "AlphaBeta")))
    Sample_name context rc.meth.lvls
      G3_26_r1
                CG 0.2542201
                   CG 0.2522355
 2:
       G3_87_r1
      G3_87_r2
                    CG 0.2524761
 3:
                   CG 0.2482041
 4: G31_109_r1
                    CG
 5: G31_109_r2
                          0.2654014
                    CG
                          0.2623544
 6: G31_119_r1
```

2.4 Information about Sample file.

This file containing information on generation times and pedigree lineages

```
# Sample file
head(fread(system.file("extdata/dm", "sampleInfo.csv", package = "AlphaBeta")))

Sample Generation Lineage
```

```
1:
    G3 26 r1
    G3_87_r1
                      3
                             87
2:
                      3
                             87
3:
    G3 87 r2
                            109
                     31
4: G31_109_r1
5: G31_109_r2
                     31
                             109
6: G31_119_r1
                     31
                             119
```

2.5 File containing lineage branch points

39

```
# Sample file
head(fread(system.file("extdata/dm", "branchPoints.csv", package = "AlphaBeta")))
    BP Generation Lineage
         0
 1:
                 none
    1
             2
 2: 2
                   87
 3: 3
            30
                    109
 4: 4
              30
                    119
              30
                     29
 5: 5
```

3 Germline epimutations

30

6: 6

3.1 Calculate divergence times

To calculate divergence times of the pedigree should be provided in the form of 4 files as shown below.

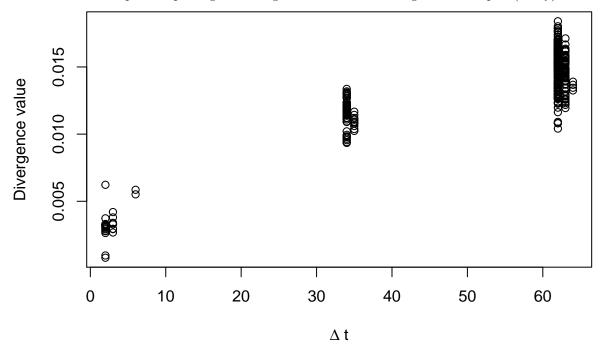
```
package = "AlphaBeta"), sep = "\t", header = TRUE)
context <- "CG"</pre>
```

calculate divergence times of the pedigree:

```
pedigree <- convertDMATRIX(sample.info = sample.info, branch.points = branch.points,
    dmatrix = dmatrix, design = "sibling")
head(pedigree)</pre>
```

```
time0 time1 time2
                             D.value
[1,]
                3
                       3 0.005516667
         0
[2,]
         0
                3
                       3 0.005856857
[3,]
         0
                3
                     31 0.011792749
                     31 0.009345341
[4,]
         0
                3
[5,]
         0
                3
                     31 0.010905316
[6,]
         0
                3
                     31 0.011464732
```

This is a manual step for inspecting the divergence data and removing outlier samples (if any):



Read in the proportions data:

Calculate initial proportions of unmethylated cytosines after removal of outliers:

```
pOuu_in <- 1 - mean(as.numeric(as.character(props[, 3])))
pOuu_in</pre>
```

[1] 0.7435074

3.2 Run Models

3.2.1 Run Model with no selection (ABneutral)

Progress: 0.25 Progress: 0.5 Progress: 0.75 Progress: 1

NOTE: it is recommended to use at least 50 Nstarts to achieve best solutions

Showing summary output of only output:

summary(output)

```
Length Class
                                  Mode
estimates
                   20
                        data.frame list
                   20
estimates.flagged
                        data.frame list
pedigree
                 2457
                       -none-
                                  numeric
settings
                    2 data.frame list
model
                    1
                        -none-
                                  character
                  315
                       -none-
                                  numeric
for.fit.plot
```

head(output\$pedigree)

```
time0 time1 time2
                        div.obs delta.t
                                                       residual
                                          div.pred
[1,]
        0
                   3 0.005516667 6 0.008938422 -0.0034217553
             3
                  3 0.005856857
[2,]
        0
             3
                                     6 0.008938422 -0.0030815654
[3,]
        0
             3 31 0.011792749
                                     34 0.011924748 -0.0001319988
[4,]
       0
            3 31 0.009345341
                                   34 0.011924748 -0.0025794065
[5,]
        0
             3
                  31 0.010905316
                                     34 0.011924748 -0.0010194322
[6,]
        0
             3
                                     34 0.011924748 -0.0004600156
                  31 0.011464732
```

3.2.2 Run model with selection against spontaneous gain of methylation (ABselectMM)

```
output <- ABselectMM(pedigree.data = pedigree, p0uu = p0uu_in,
    eqp = p0uu_in, eqp.weight = 1, Nstarts = 4, out.dir = output.data.dir,
    out.name = "CG_global_estimates_ABselectMM")</pre>
```

Progress: 0.25 Progress: 0.5 Progress: 0.75 Progress: 1

summary(output)

	Length	Class	Mode
estimates	22	${\tt data.frame}$	list
${\tt estimates.flagged}$	22	${\tt data.frame}$	list
pedigree	2457	-none-	numeric
settings	2	${\tt data.frame}$	list
model	1	-none-	${\tt character}$
for fit plot	315	-none-	numeric

3.2.3 Run model with selection against spontaneous loss of methylation (ABselectUU)

```
output <- ABselectUU(pedigree.data = pedigree, p0uu = p0uu_in,
        eqp = p0uu_in, eqp.weight = 1, Nstarts = 4, out.dir = output.data.dir,
        out.name = "CG_global_estimates_ABselectUU")

Progress: 0.25
Progress: 0.5
Progress: 0.75
Progress: 1

summary(output)</pre>
```

```
Length Class
                                Mode
estimates
                  22
                      data.frame list
                  22 data.frame list
estimates.flagged
pedigree
                2457
                      -none- numeric
                  2 data.frame list
settings
model
                  1
                      -none- character
                 315 -none-
for.fit.plot
                                numeric
```

3.2.4 Run model that considers no accumulation of epimutations (ABnull)

```
output <- ABnull(pedigree.data = pedigree, out.dir = output.data.dir,
   out.name = "CG_global_estimates_ABnull")
summary(output)</pre>
```

```
estimates 1 -none- numeric estimates.flagged 0 -none- NULL pedigree 2457 -none- numeric settings 0 -none- NULL model 1 -none- character for.fit.plot 1755 -none- numeric
```

3.3 Comparison of different models and selection of best model

3.3.1 Testing ABneutral vs. ABnull

```
RSS_F RSS_R df_F df_R Fvalue pvalue 7.084342e-04 4.124786e-03 3.460000e+02 3.500000e+02 4.171374e+02 6.260446e-131
```

3.3.2 Testing ABselectMM vs.ABneutral

```
file1 <- system.file("extdata/models/", "CG_global_estimates_ABselectMM.Rdata",
    package = "AlphaBeta")
file2 <- system.file("extdata/models/", "CG_global_estimates_ABnull.Rdata",</pre>
```

```
package = "AlphaBeta")
out <- FtestRSS(pedigree.select = file1, pedigree.null = file2)</pre>
out$Ftest
                        RSS_R
                                       df_F
                                                                               pvalue
          RSS_F
                                                     df_R
                                                                 Fvalue
   6.507729e-04 4.124786e-03 3.460000e+02 3.500000e+02 4.617618e+02 2.662626e-137
      Testing ABselectUU vs.ABneutral
3.3.3
file1 <- system.file("extdata/models/", "CG_global_estimates_ABselectUU.Rdata",
    package = "AlphaBeta")
file2 <- system.file("extdata/models/", "CG_global_estimates_ABnull.Rdata",
    package = "AlphaBeta")
out <- FtestRSS(pedigree.select = file1, pedigree.null = file2)</pre>
out$Ftest
         RSS_F
                        RSS_R
                                       df_F
                                                     df_R
                                                                 Fvalue
                                                                               pvalue
   6.509786e-04 4.124786e-03 3.460000e+02 3.500000e+02 4.615886e+02 2.812040e-137
     Bootstrap analysis with the best model
i.e ABneutral in our case
inputModel <- system.file("extdata/models/", "CG_global_estimates_ABneutral.Rdata",
    package = "AlphaBeta")
# Bootstrapping models CG
output.data.dir <- paste0(getwd(), "/")</pre>
Boutput <- BOOTmodel(pedigree.data = inputModel, Nboot = 4, out.dir = output.data.dir,
    out.name = "Boot_CG_global_estimates_ABneutral")
 Bootstrap interation: 0.25
 Bootstrap interation: 0.5
 Bootstrap interation: 0.75
 Bootstrap interation: 1
summary(Boutput)
                 Length Class
                                    Mode
 standard.errors 24
                        -none-
                                    numeric
 boot.base
                 20
                        data.frame list
 settings
                 2 data.frame list
 N.boots
                 1
                       -none-
                                    numeric
 N.good.boots
                  1
                         -none-
                                    numeric
 boot.results
                  19
                        data.frame list
 model
                  1
                         -none-
                                    character
Boutput$standard.errors
                       SE
                                  2.5%
                                              97.5%
  alpha
             5.508637e-06 0.0001071533 0.0001195495
             1.602496e-05 0.0003111296 0.0003471905
  beta
 beta/alpha 2.507930e-04 2.9035914206 2.9041517430
             1.903058e-03 0.0220164790 0.0259587939
 weight
  intercept 1.780034e-04 0.0016808194 0.0020449250
```

```
PrMMinf 3.281715e-05 0.2557819957 0.2558558709
PrUMinf 3.272287e-05 0.0006369610 0.0007105979
PrUUinf 1.407742e-07 0.7435071671 0.7435074548
```

4 Somatic epimutations

4.1 Loading data and generation of pedigree

4.2 Generate pedigree from the input files

```
0
           297
                  287 0.003796614
[1,]
[2,]
            297
                  324 0.003974756
        0
[3,]
            327
                  287 0.003995156
        0
[4,]
        0
            297
                  287 0.004040671
[5,]
        0
            328
                  287 0.004046553
[6,]
        0
            328
                  287 0.004048672
```

4.3 Calculate the proportion of unmethylated cytosines

```
p0uu_in <- mean(props[, 3])
p0uu_in</pre>
[1] 0.2564926
```

4.4 Run Models

4.4.1 Run Model with no selection (outneutral)

```
outneutral <- ABneutralSOMA(pedigree.data = pedigree.out, p0uu = p0uu_in,
        eqp = p0uu_in, eqp.weight = 0.001, Nstarts = 5, out.dir = output.data.dir,
        out.name = "ABneutralSOMA_CG_estimates")

Progress: 0.2
Progress: 0.4
Progress: 0.6
Progress: 0.8
Progress: 1</pre>
summary(outneutral)
```

```
\begin{array}{cccc} & Length & Class & Mode \\ estimates & 20 & data.frame & list \\ estimates.flagged & 20 & data.frame & list \\ \end{array}
```

```
pedigree 196 -none- numeric settings 2 data.frame list model 1 -none- character for.fit.plot 3275 -none- numeric
```

head(outneutral\$pedigree)

```
time0 time1 time2
                       div.obs delta.t div.pred residual
       0 297
                287 0.003796614
[1,]
                                 584
                                         NA
       0
[2,]
          297
                324 0.003974756
                                 621
                                          NA
                                                  NA
[3,]
       0 327
                287 0.003995156
                                 614
                                          NA
                                                  NA
                                          NA
[4,]
       0 297 287 0.004040671
                                 584
                                                  NA
                287 0.004046553
[5,]
       0 328
                                 615
                                          NA
                                                  NA
          328
                                          NA
[6,]
       0
                287 0.004048672
                                 615
                                                  NA
```

4.4.2 Run model with selection against spontaneous gain of methylation (outselectMM)

```
outselectMM <- ABselectMMSOMA(pedigree.data = pedigree.out, p0uu = p0uu_in,
        eqp = p0uu_in, eqp.weight = 0.001, Nstarts = 10, out.dir = output.data.dir,
        out.name = "ABselectMMSOMA_CG_estimates")

Progress: 0.1
Progress: 0.2
Progress: 0.3
Progress: 0.4
Progress: 0.5
Progress: 0.6
Progress: 0.6
Progress: 0.7
Progress: 0.8
Progress: 0.9
Progress: 0.9</pre>
```

summary(outselectMM)

```
Length Class
                                Mode
estimates
                 22 data.frame list
estimates.flagged
                22 data.frame list
                 196 -none- numeric
pedigree
settings
                  2
                      data.frame list
model
                  1 -none- character
for.fit.plot
                3275 -none-
                                numeric
```

4.4.3 Run model with selection against spontaneous loss of methylation (outselect UU)

```
outselectUU <- ABselectUUSOMA(pedigree.data = pedigree.out, p0uu = p0uu_in,
    eqp = p0uu_in, eqp.weight = 0.001, Nstarts = 10, out.dir = output.data.dir,
    out.name = "ABselectUUSOMA_CG_estimates")</pre>
```

Progress: 0.1
Progress: 0.2
Progress: 0.3
Progress: 0.4
Progress: 0.5
Progress: 0.6
Progress: 0.7
Progress: 0.8
Progress: 0.9
Progress: 1

summary(outselectUU)

	Length	Class	Mode
estimates	22	${\tt data.frame}$	list
estimates.flagged	22	${\tt data.frame}$	list
pedigree	196	-none-	numeric
settings	2	${\tt data.frame}$	list
model	1	-none-	character
for.fit.plot	3275	-none-	numeric

5 R session info

sessionInfo()

R version 3.6.1 (2019-07-05)

Platform: x86_64-pc-linux-gnu (64-bit) Running under: Ubuntu 18.04.2 LTS

Matrix products: default

BLAS: /usr/lib/x86_64-linux-gnu/openblas/libblas.so.3 LAPACK: /usr/lib/x86_64-linux-gnu/libopenblasp-r0.2.20.so

locale:

[1] LC_CTYPE=en_US.UTF-8	LC_NUMERIC=C	LC_TIME=en_US.UTF-8
[4] LC_COLLATE=C	LC_MONETARY=en_US.UTF-8	LC_MESSAGES=en_US.UTF-8

[7] LC_PAPER=en_US.UTF-8 LC_NAME=C LC_ADDRESS=C

[10] LC_TELEPHONE=C LC_MEASUREMENT=en_US.UTF-8 LC_IDENTIFICATION=C

attached base packages:

[1] stats graphics grDevices utils datasets methods base

other attached packages:

[1] data.table_1.12.2 AlphaBeta_0.99.0

loaded via a namespace (and not attached):

[1]	Rcpp_1.0.1	formatR_1.7	compiler_3.6.1	pillar_1.4.1
[5]	iterators_1.0.10	<pre>prettyunits_1.0.2</pre>	remotes_2.1.0	tools_3.6.1
[9]	testthat_2.1.1	digest_0.6.19	pkgbuild_1.0.3	pkgload_1.0.2
[13]	evaluate_0.14	lattice_0.20-38	memoise_1.1.0	tibble_2.1.3
[17]	pkgconfig_2.0.2	rlang_0.4.0	Matrix_1.2-17	foreach_1.4.4
[21]	cli_1.1.0	rstudioapi_0.10	commonmark_1.7	yaml_2.2.0
[25]	parallel_3.6.1	expm_0.999-4	xfun_0.8	knitr_1.23
[29]	withr_2.1.2	stringr_1.4.0	dplyr_0.8.1	roxygen2_6.1.1
[33]	xml2_1.2.0	gtools_3.8.1	desc_1.2.0	fs_1.3.1
[37]	devtools_2.1.0	grid_3.6.1	rprojroot_1.3-2	tidyselect_0.2.5
[41]	glue_1.3.1	R6_2.4.0	processx_3.3.1	rmarkdown_1.13
[45]	sessioninfo_1.1.1	callr_3.2.0	purrr_0.3.2	magrittr_1.5
[49]	htmltools_0.3.6	codetools_0.2-16	backports_1.1.4	ps_1.3.0
[53]	usethis_1.5.0	assertthat_0.2.1	numDeriv_2016.8-1.1	optimx_2018-7.10
[57]	stringi 1.4.3	doParallel 1.0.14	crayon 1.3.4	