

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 meiotic 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 perennials, 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

NOTE In this tutorial we are reading methylome files from the methimpute package:

You can find more information here: [Methimpute package](#)

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)
```

	filename	generation	lineage
1:	data/methylome_Col_G0-merged.txt	G0	
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
1:	G3_26_r1	CG	0.2542201
2:	G3_87_r1	CG	0.2522355
3:	G3_87_r2	CG	0.2524761
4:	G31_109_r1	CG	0.2482041
5:	G31_109_r2	CG	0.2654014
6:	G31_119_r1	CG	0.2623544

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	3	26
2:	G3_87_r1	3	87
3:	G3_87_r2	3	87
4:	G31_109_r1	31	109
5:	G31_109_r2	31	109
6:	G31_119_r1	31	119

2.5 File containing lineage branch points

Sample file

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

	BP	Generation	Lineage
1:	1	0	none
2:	2	2	87
3:	3	30	109
4:	4	30	119
5:	5	30	29
6:	6	30	39

3 Germline epimutations

Models ABneutral, ABselectMM and ABselectUU can be used to estimate the rate of spontaneous epimutations from pedigree-based high-throughput DNA methylation data. The models are generally designed for pedigree data arising from selfing diploid species.

3.1 Calculate divergence times

Divergence time (Δt) is calculated as follows: $\Delta t = t_1 + t_2 - 2 \cdot t_0$, where t_1 is the time of sample 1 (in generations), t_2 is the time of sample 2 (in generations) and t_0 is the time (in generations) of the most recent common founder of samples 1 and 2.

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

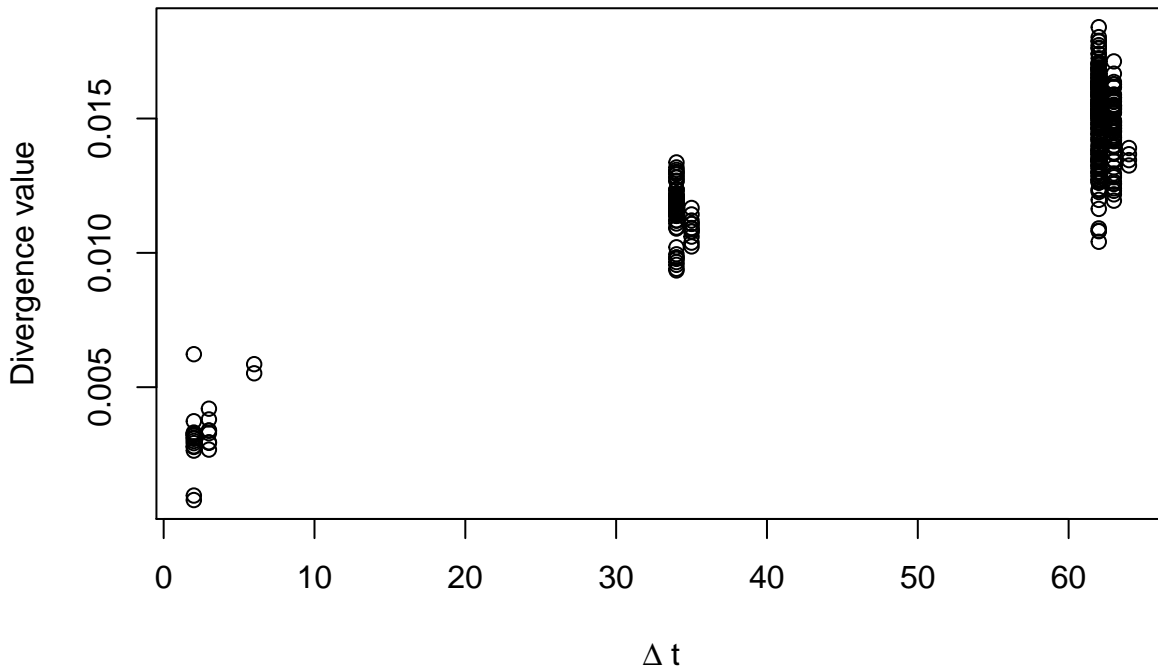
```
props.name <- read.table(system.file("extdata/dm", "AB-methprop-CG-0.99.csv",
  package = "AlphaBeta"), sep = "\t", header = TRUE)
sample.info <- read.table(system.file("extdata/dm", "sampleInfo.csv",
  package = "AlphaBeta"), sep = "\t", header = TRUE)
branch.points <- read.table(system.file("extdata/dm", "branchPoints.csv",
  package = "AlphaBeta"), sep = "\t", header = TRUE)
dmatrix <- read.table(system.file("extdata/dm", "AB-dMatrix-CG-0.99.csv",
  package = "AlphaBeta"), sep = "\t", header = TRUE)
context <- "CG"
```

calculate divergence times of the pedigree:

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

	time0	time1	time2	D.value
[1,]	0	3	3	0.005516667
[2,]	0	3	3	0.005856857
[3,]	0	3	31	0.011792749
[4,]	0	3	31	0.009345341
[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:

```
outliers <- "none"
dmatrix <- dmatrix[which(dmatrix[, 1] != outliers), ]
dmatrix <- dmatrix[which(dmatrix[, 2] != outliers), ]
pedigree <- pedigree[c(as.numeric(rownames(dmatrix))), ]

props <- props.name[which(as.character(props.name[, 2]) == context), ]
```

```
]
props <- props.name[which(!is.element(props.name[, 1], outliers) ==
  TRUE), ]
```

Calculate initial proportions of unmethylated cytosines after removal of outliers:

```
p0uu_in <- 1 - mean(as.numeric(as.character(props[, 3])))
p0uu_in
```

```
[1] 0.7435074
```

3.2 Run Models

3.2.1 Run Model with no selection (ABneutral)

This model assumes that heritable gains and losses in cytosine methylation are selectively neutral.

```
# output directory
output.data.dir <- paste0(getwd(), "/")

output <- ABneutral(pedigree.data = pedigree, p0uu = p0uu_in,
  eqp = p0uu_in, eqp.weight = 1, Nstarts = 4, out.dir = output.data.dir,
  out.name = "CG_global_estimates_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
estimates.flagged	20	data.frame	list
pedigree	2457	-none-	numeric
settings	2	data.frame	list
model	1	-none-	character
for.fit.plot	315	-none-	numeric

```
head(output$pedigree)
```

	time0	time1	time2	div.obs	delta.t	div.pred	residual
[1,]	0	3	3	0.005516667	6	0.007859192	-2.342525e-03
[2,]	0	3	3	0.005856857	6	0.007859192	-2.002335e-03
[3,]	0	3	31	0.011792749	34	0.011423787	3.689615e-04
[4,]	0	3	31	0.009345341	34	0.011423787	-2.078446e-03
[5,]	0	3	31	0.010905316	34	0.011423787	-5.184719e-04
[6,]	0	3	31	0.011464732	34	0.011423787	4.094473e-05

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

This model assumes that heritable losses of cytosine methylation are under negative selection. The selection parameter is estimated.

```
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")
```

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

```
summary(output)
```

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

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

This model assumes that heritable gains of cytosine methylation are under negative selection. The selection parameter is estimated.

```
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)
```

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

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

This is the null model of no accumulation.

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

```
summary(output)
```

	Length	Class	Mode
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

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

out <- FtestRSS(pedigree.select = file1, pedigree.null = file2)

out$Ftest
```

	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",
  package = "AlphaBeta")

out <- FtestRSS(pedigree.select = file1, pedigree.null = file2)

out$Ftest
```

	RSS_F	RSS_R	df_F	df_R	Fvalue	pvalue
	6.507729e-04	4.124786e-03	3.460000e+02	3.500000e+02	4.617618e+02	2.662626e-137

3.3.3 Testing ABselectUU vs. ABneutral

```
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)

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

3.4 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(), "/")

Boutput <- BOOTmodel(pedigree.data = inputModel, Nboot = 4, out.dir = output.data.dir,
  out.name = "Boot_CG_global_estimates_ABneutral")
```

```

Bootstrap iteration: 0.25
Bootstrap iteration: 0.5
Bootstrap iteration: 0.75
Bootstrap iteration: 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	8.498549e-06	9.942815e-05	0.0001172198
beta	2.471870e-05	2.886639e-04	0.0003404124
beta/alpha	3.861497e-04	2.903241e+00	2.9040451810
weight	5.155835e-03	1.964948e-02	0.0309148665
intercept	4.833266e-05	2.171803e-03	0.0022764591
PrMMinf	5.058372e-05	2.557959e-01	0.2559018183
PrUMinf	5.048680e-05	5.910658e-04	0.0006967596
PrUUnif	1.384886e-07	7.435071e-01	0.7435073898

4 Somatic epimutations

Models ABneutralSOMA, ABselectMMSOMA and ABselectUUSOMA can be used to estimate the rate of spontaneous epimutations from pedigree-based high-throughput DNA methylation data. The models are generally designed for pedigree data arising from clonally or asexually propagated diploid species. The models can also be applied to long-lived perennials, such as trees, using leaf methylomes and coring data as input. In this case, the tree branching structure is treated as an intra-organismal pedigree (or phylogeny) of somatic lineages.

4.1 Loading data and generation of pedigree

```

props.name <- read.table(system.file("extdata/soma/", "AB-methprop-CG-0.99.csv",
  package = "AlphaBeta"), sep = "\t", header = TRUE, stringsAsFactors = FALSE)
sample.info <- read.table(system.file("extdata/soma/", "sampleInfo.csv",
  package = "AlphaBeta"), sep = "\t", header = TRUE, stringsAsFactors = FALSE)
dmatrix <- read.table(system.file("extdata/soma/", "AB-dMatrix-CG-0.99.csv",
  package = "AlphaBeta"), sep = "\t", header = TRUE, stringsAsFactors = FALSE)

```

4.2 Generate pedigree from the input files

```

pedigree.out <- makePHYLO(tall = 330, pedigree = dmatrix, sample.info = sample.info)
pedigree.out <- pedigree.out[[1]]
head(pedigree.out)

```

	time0	time1	time2	D.value
[1,]	0	297	287	0.003796614
[2,]	0	297	324	0.003974756
[3,]	0	327	287	0.003995156
[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
```

```
[1] 0.2564926
```

4.4 Run Models

4.4.1 Run Model with no selection (ABneutralSOMA)

This model assumes that somatically heritable gains and losses in cytosine methylation are selectively neutral.

```
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
```

```
summary(outneutral)
```

	Length	Class	Mode
estimates	20	data.frame	list
estimates.flagged	20	data.frame	list
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
[1,]	0	297	287	0.003796614	584	0.004010233	-2.136193e-04
[2,]	0	297	324	0.003974756	621	0.004125074	-1.503178e-04
[3,]	0	327	287	0.003995156	614	0.004103354	-1.081976e-04
[4,]	0	297	287	0.004040671	584	0.004010233	3.043771e-05
[5,]	0	328	287	0.004046553	615	0.004106457	-5.990430e-05
[6,]	0	328	287	0.004048672	615	0.004106457	-5.778530e-05

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

This model assumes that somatically heritable losses of cytosine methylation are under negative selection. The selection parameter is estimated.

```
outselectMM <- ABselectMMSOMA(pedigree.data = pedigree.out, p0uu = p0uu_in,
  eqp = p0uu_in, eqp.weight = 0.001, Nstarts = 5, out.dir = output.data.dir,
  out.name = "ABselectMMSOMA_CG_estimates")
```

```
Progress: 0.2
Progress: 0.4
Progress: 0.6
Progress: 0.8
Progress: 1
```

```
summary(outselectMM)
```

	Length	Class	Mode
estimates	22	data.frame	list
estimates.flagged	22	data.frame	list
pedigree	196	-none-	numeric
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 (ABselectUUSOMA)

This model assumes that somatically heritable gains of cytosine methylation are under negative selection. The selection parameter is estimated.

```
outselectUU <- ABselectUUSOMA(pedigree.data = pedigree.out, p0uu = p0uu_in,  
  eqp = p0uu_in, eqp.weight = 0.001, Nstarts = 5, out.dir = output.data.dir,  
  out.name = "ABselectUUSOMA_CG_estimates")
```

```
Progress: 0.2  
Progress: 0.4  
Progress: 0.6  
Progress: 0.8  
Progress: 1
```

```
summary(outselectUU)
```

	Length	Class	Mode
estimates	22	data.frame	list
estimates.flagged	22	data.frame	list
pedigree	196	-none-	numeric
settings	2	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/libopenblas-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	prettyunits_1.0.2	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	