Package: rMVP (via r-universe)

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Type Package

Title Memory-Efficient, Visualize-Enhanced, Parallel-Accelerated GWAS Tool

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Description A memory-efficient, visualize-enhanced, parallel-accelerated Genome-Wide Association Study (GWAS) tool. It can (1) effectively process large data, (2) rapidly evaluate population structure, (3) efficiently estimate variance components several algorithms, (4) implement parallel-accelerated association tests of markers three methods, (5) globally efficient design on GWAS process computing, (6) enhance visualization of related information. 'rMVP' contains three models GLM (Alkes Price (2006) <DOI:10.1038/ng1847>), MLM (Jianming Yu (2006) <DOI:10.1038/ng1702>) and FarmCPU (Xiaolei Liu (2016) <doi:10.1371/journal.pgen.1005767>); variance components estimation methods EMMAX (Hyunmin Kang (2008) <DOI:10.1534/genetics.107.080101>;), FaSTLMM (method: Christoph Lippert (2011) <DOI:10.1038/nmeth.1681>, R implementation from 'GAPIT2': You Tang and Xiaolei Liu (2016) <DOI:10.1371/journal.pone.0107684> and 'SUPER': Qishan Wang and Feng Tian (2014) <DOI:10.1371/journal.pone.0107684>), and HE regression (Xiang Zhou (2017) <DOI:10.1214/17-AOAS1052>).

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URL https://github.com/xiaolei-lab/rMVP

BugReports https://github.com/xiaolei-lab/rMVP/issues

Imports utils, stats, methods, graphics, grDevices, MASS, bigmemory, RhpcBLASctl

Depends R (>= 3.3)

LinkingTo Rcpp, RcppArmadillo, RcppEigen, RcppProgress, BH, bigmemory

NeedsCompilation yes

Suggests knitr, testthat, rmarkdown

RoxygenNote 7.3.2

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MVP *MVP, Memory-efficient, Visualization-enhanced, Parallel-accelerated*

Description

Object 1: To perform GWAS using General Linear Model (GLM), Mixed Linear Model (MLM), and FarmCPU model Object 2: To calculate kinship among individuals using Varaden method Object 3: Estimate variance components using EMMA, FaST-LMM, and HE regression Object 4: Generate high-quality figures

Usage

```
MVP(
  phe,
  geno,
  map,
  K = NULL,
  nPC.GLM = NULL,
  nPC.MLM = NULL,
  nPC.FarmCPU = NULL,
  CV.GLM = NULL,
  CV.MLM = NULL,
  CV.FarmCPU = NULL,
  REML = NULL,
  maxLine = 10000,
  ncpus = detectCores(logical = FALSE),
  vc.method = c("BRENT", "EMMA", "HE"),
  method = c("GLM", "MLM", "FarmCPU"),
  maf = NULL,
  p.threshold = NA,
  QTN.threshold = 0.01,
  method.bin = "static",
  bin.size = c(5e+05, 5e+06, 5e+07),
  bin.selection = seq(10, 100, 10),
  maxLoop = 10,
  permutation.threshold = FALSE,
  permutation.rep = 100,
  memo = NULL,
  outpath = getwd(),
 col = c("#4197d8", "#f8c120", "#413496", "#495226", "#d60b6f", "#e66519", "#d581b7",
    "#83d3ad", "#7c162c", "#26755d"),
  file.output = TRUE,
  file.type = "jpg",
  dpi = 300,
  threshold = 0.05,
  verbose = TRUE
)
```

Arguments

phe	phenotype, n * 2 matrix, n is sample size
geno	genotype, either m by n or n by m is supportable, m is marker size, n is population size
map	SNP map information, SNP name, Chr, Pos
K	Kinship, Covariance matrix(n * n) for random effects, must be positive semi-definite
nPC.GLM	number of PCs added as fixed effects in GLM
nPC.MLM	number of PCs added as fixed effects in MLM
nPC.FarmCPU	number of PCs added as fixed effects in FarmCPU
CV.GLM	covariates added in GLM
CV.MLM	covariates added in MLM
CV.FarmCPU	covariates added in FarmCPU
REML	a list contains ve and vg
maxLine	the number of markers handled at a time, smaller value would reduce the memory cost
ncpus	number of cpus used for parallel
vc.method	methods for estimating variance component("EMMA" or "HE" or "BRENT")
method	the GWAS model, "GLM", "MLM", and "FarmCPU", models can be selected simutaneously, i.e. c("GLM", "MLM", "FarmCPU")
maf	the threshold of minor allele frequency to filter SNPs in analysis
p.threshold	if all p values generated in the first iteration are bigger than p.threshold, FarmCPU stops
QTN.threshold	in second and later iterations, only SNPs with lower p-values than QTN.threshold have chances to be selected as pseudo QTNs
method.bin	'static' or 'FaST-LMM'
bin.size	window size in genome
bin.selection	a vector, how many windows selected
maxLoop	maximum number of iterations
permutation.threshold	if use a permutation cutoff or not (bonferroni cutoff)
permutation.rep	number of permutation replicates
memo	Character. A text marker on output files
outpath	the path of the output files
col	for color of points in each chromosome on manhattan plot
file.output	whether to output files or not
file.type	figure formats, "jpg", "tiff"
dpi	resolution for output figures
threshold	a cutoff line on manhattan plot, 0.05/marker size
verbose	whether to print detail.

Value

Output: MVP.return$map - SNP map information, SNP name, Chr, Pos Output: MVP.return$glm.results - p-values obtained by GLM method Output: MVP.return$mlm.results - p-values obtained by MLM method Output: MVP.return$farmcpu.results - p-values obtained by FarmCPU method

Author(s)

Lilin Yin, Haohao Zhang, and Xiaolei Liu

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")
phenotype <- read.table(phePath, header=TRUE)
print(dim(phenotype))
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
print(dim(genotype))
mapPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.map", package = "rMVP")
map <- read.table(mapPath , head = TRUE)

opts <- options(rMVP.OutputLog2File = FALSE)

mvp <- MVP(phe=phenotype, geno=genotype, map=map, maxLoop=3,
  method=c("GLM", "MLM", "FarmCPU"), file.output=FALSE, ncpus=1)
str(mvp)

options(opts)
```

MVP.BRENT.Vg.Ve *MVP.BRENT.Vg.Ve variance component estimation using the BRENT method*

Description

MVP.BRENT.Vg.Ve variance component estimation using the BRENT method

Usage

```
MVP.BRENT.Vg.Ve(y, X, eigenK, verbose = FALSE)
```

Arguments

y	phenotype
X	covariate matrix, the first column is 1s
eigenK	eigen of Kinship matrix
verbose	whether to print detail.

Value

vg, ve, and delta

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")
phenotype <- read.table(phePath, header=TRUE)
print(dim(phenotype))
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
print(dim(genotype))

eigenK <- eigen(MVP.K.VanRaden(genotype, cpu=1))
vc <- MVP.BRENT.Vg.Ve(y=phenotype[,2], X=matrix(1, nrow(phenotype)), eigenK=eigenK)
print(vc)
```

`MVP.Data` *MVP.Data: To prepare data for MVP package*

Description

MVP.Data: To prepare data for MVP package

Usage

```
MVP.Data(
  fileMVP = NULL,
  fileVCF = NULL,
  fileHMP = NULL,
  fileBed = NULL,
  fileNum = NULL,
  fileMap = NULL,
  filePhe = NULL,
  fileInd = NULL,
  fileKin = NULL,
  filePC = NULL,
  out = "mvp",
  sep.num = "\t",
  auto_transpose = TRUE,
  sep.map = "\t",
  sep.phe = "\t",
  sep.kin = "\t",
  sep.pc = "\t",
  type.geno = "char",
  pheno_cols = NULL,
  SNP.impute = "Major",
```

```
  maxLine = 10000,
  pcs.keep = 5,
  verbose = TRUE,
  ncpus = NULL,
  ...
)
```

Arguments

fileMVP	Genotype in MVP format
fileVCF	Genotype in VCF format
fileHMP	Genotype in hapmap format
fileBed	Genotype in PLINK binary format
fileNum	Genotype in numeric format; pure 0, 1, 2 matrix; m * n or n * m, m is marker size, n is sample size
fileMap	SNP map information, there are three columns, including SNP_ID, Chromosome, and Position
filePhe	Phenotype, the first column is taxa name, the subsequent columns are traits
fileInd	Individual name file
fileKin	Kinship that represents relationship among individuals, n * n matrix, n is sample size
filePC	Principal components, n*npc, n is sample size, npc is number of top columns of principal components
out	prefix of output file name
sep.num	seperator for numeric file.
auto_transpose	Whether to automatically transpose numeric genotypes, the default is True, which will identify the most one of the rows or columns as a marker, If set to False, the row represents the marker and the column represents the individual.
sep.map	seperator for map file.
sep.phe	seperator for phenotype file.
sep.kin	seperator for Kinship file.
sep.pc	seperator for PC file.
type.geno	type parameter in bigmemory, genotype data. The default is char, it is highly recommended *NOT* to modify this parameter.
pheno_cols	Extract which columns of the phenotype file (including individual IDs)
SNP.impute	"Left", "Middle", "Right", or NULL for skip impute.
maxLine	the number of markers handled at a time, smaller value would reduce the memory cost
pcs.keep	how many PCs to keep
verbose	whether to print detail.
ncpus	The number of threads used, if NULL, (logical core number - 1) is automatically used
...	Compatible with DEPRECATED parameters.

Value

NULL Output files: genotype.desc, genotype.bin: genotype file in bigmemory format phenotype.phe: ordered phenotype file, same taxa order with genotype file map.map: SNP information k.desc, k.bin: Kinship matrix in bigmemory format pc.desc, pc.bin: PC matrix in bigmemory format Requirement: fileHMP, fileBed, and fileNum can not input at the same time

Examples

```
bfilePath <- file.path(system.file("extdata", "02_bfile", package = "rMVP"), "mvp")
opts <- options(rMVP.OutputLog2File = FALSE)

MVP.Data(fileBed=bfilePath, out=tempfile("outfile"), ncpus=1)

options(opts)
```

`MVP.Data.Bfile2MVP` *MVP.Data.Bfile2MVP: To transform plink binary data to MVP package Author: Haohao Zhang Build date: Sep 12, 2018*

Description

MVP.Data.Bfile2MVP: To transform plink binary data to MVP package Author: Haohao Zhang Build date: Sep 12, 2018

Usage

```
MVP.Data.Bfile2MVP(
  bfile,
  out = "mvp",
  maxLine = 10000,
  type.geno = "char",
  threads = 0,
  verbose = TRUE
)
```

Arguments

`bfile`	Genotype in binary format (.bed, .bim, .fam)
`out`	the name of output file
`maxLine`	the number of markers handled at a time, smaller value would reduce the memory cost
`type.geno`	the type of genotype elements
`threads`	number of thread for transforming
`verbose`	whether to print the reminder

Value

number of individuals and markers. Output files: genotype.desc, genotype.bin: genotype file in bigmemory format phenotype.phe: ordered phenotype file, same taxa order with genotype file map.map: SNP information

Examples

```
bfilePath <- file.path(system.file("extdata", "02_bfile", package = "rMVP"), "mvp")

MVP.Data.Bfile2MVP(bfilePath, tempfile("outfile"), threads=1)
```

`MVP.Data.Hapmap2MVP` *MVP.Data.Hapmap2MVP: To transform Hapmap data to MVP package Author: Haohao Zhang Build date: Sep 12, 2018*

Description

MVP.Data.Hapmap2MVP: To transform Hapmap data to MVP package Author: Haohao Zhang Build date: Sep 12, 2018

Usage

```
MVP.Data.Hapmap2MVP(
  hmp_file,
  out = "mvp",
  maxLine = 10000,
  type.geno = "char",
  threads = 1,
  verbose = TRUE
)
```

Arguments

`hmp_file`	Genotype in Hapmap format
`out`	the name of output file
`maxLine`	the number of markers handled at a time, smaller value would reduce the memory cost
`type.geno`	the type of genotype elements
`threads`	number of thread for transforming
`verbose`	whether to print the reminder

Value

number of individuals and markers. Output files: genotype.desc, genotype.bin: genotype file in bigmemory format phenotype.phe: ordered phenotype file, same taxa order with genotype file map.map: SNP information

Examples

```
hapmapPath <- system.file("extdata", "03_hapmap", "mvp.hmp.txt", package = "rMVP")

MVP.Data.Hapmap2MVP(hapmapPath, tempfile("outfile"), threads=1)
```

`MVP.Data.impute` *MVP.Data.impute: To impute the missing genotype Author: Haohao Zhang Build date: Sep 12, 2018*

Description

MVP.Data.impute: To impute the missing genotype Author: Haohao Zhang Build date: Sep 12, 2018

Usage

```
MVP.Data.impute(
  mvp_prefix,
  out = NULL,
  mrk_bycol = TRUE,
  method = "Major",
  ncpus = NULL,
  verbose = TRUE
)
```

Arguments

`mvp_prefix`	the prefix of mvp file
`out`	the prefix of output file
`mrk_bycol`	whether the markers are stored by columns in genotype (i.e. genotype is a n by m matrix)
`method`	'Major', 'Minor', "Middle"
`ncpus`	number of threads for imputing
`verbose`	whether to print the reminder

Value

NULL Output files: imputed genotype file

Examples

```
mvpPath <- file.path(system.file("extdata", "05_mvp", package = "rMVP"), "mvp")

MVP.Data.impute(mvpPath, tempfile("outfile"), ncpus=1)
```

`MVP.Data.Kin` *Kinship*

Description

Kinship

Usage

```
MVP.Data.Kin(
  fileKin = TRUE,
  mvp_prefix = "mvp",
  out = NULL,
  maxLine = 10000,
  mrk_bycol = TRUE,
  sep = "\t",
  cpu = 1,
  verbose = TRUE
)
```

Arguments

`fileKin`	Kinship that represents relationship among individuals, n * n matrix, n is sample size
`mvp_prefix`	Prefix for mvp format files
`out`	prefix of output file name
`maxLine`	the number of markers handled at a time, smaller value would reduce the memory cost
`mrk_bycol`	whether the markers are stored by columns in genotype (i.e. genotype is a n by m matrix)
`sep`	seperator for Kinship file.
`cpu`	the number of cpu
`verbose`	whether to print detail.

Value

Output file: <out>.kin.bin <out>.kin.desc

Examples

```
geno <- file.path(system.file("extdata", "06_mvp-impute", package = "rMVP"), "mvp.imp")

MVP.Data.Kin(TRUE, mvp_prefix=geno, out=tempfile("outfile"), cpu=1)
```

`MVP.Data.Map` *MVP.Data.Map: To check map file Author: Haohao Zhang Build date: Sep 12, 2018*

Description

MVP.Data.Map: To check map file Author: Haohao Zhang Build date: Sep 12, 2018

Usage

```
MVP.Data.Map(
  map,
  out = "mvp",
  cols = 1:5,
  header = TRUE,
  sep = "\t",
  verbose = TRUE
)
```

Arguments

`map`	the name of map file or map object(data.frame or matrix)
`out`	the name of output file
`cols`	selected columns
`header`	whether the file contains header
`sep`	seperator of the file
`verbose`	whether to print detail.

Value

Output file: <out>.map

Examples

```
mapPath <- system.file("extdata", "05_mvp", "mvp.geno.map", package = "rMVP")

MVP.Data.Map(mapPath, tempfile("outfile"))
```

`MVP.Data.MVP2Bfile` *MVP.Data.MVP2Bfile: To transform MVP data to binary format Author: Haohao Zhang Build date: Sep 12, 2018*

Description

MVP.Data.MVP2Bfile: To transform MVP data to binary format Author: Haohao Zhang Build date: Sep 12, 2018

Usage

```
MVP.Data.MVP2Bfile(
  bigmat,
  map,
  pheno = NULL,
  out = "mvp.plink",
  threads = 1,
  verbose = TRUE
)
```

Arguments

`bigmat`	Genotype in bigmatrix format (0,1,2)
`map`	the map file
`pheno`	the phenotype file
`out`	the name of output file
`threads`	number of thread for transforming
`verbose`	whether to print the reminder

Value

NULL Output files: .bed, .bim, .fam

Examples

```
bigmat <- as.big.matrix(matrix(1:6, 3, 2))
map <- matrix(c("rs1", "rs2", "rs3", 1, 1, 1, 10, 20, 30), 3, 3)

MVP.Data.MVP2Bfile(bigmat, map, out=tempfile("outfile"), threads=1)
```

`MVP.Data.Numeric2MVP` *MVP.Data.Numeric2MVP: To transform Numeric data to MVP package Author: Haohao Zhang Build date: Sep 12, 2018*

Description

MVP.Data.Numeric2MVP: To transform Numeric data to MVP package Author: Haohao Zhang Build date: Sep 12, 2018

Usage

```
MVP.Data.Numeric2MVP(
  num_file,
  map_file,
  out = "mvp",
  maxLine = 10000,
  row_names = FALSE,
  col_names = FALSE,
  type.geno = "char",
  auto_transpose = TRUE,
  verbose = TRUE
)
```

Arguments

`num_file`	Genotype in Numeric format (0,1,2)
`map_file`	Genotype map file, SNP_name, Chr, Pos
`out`	the name of output file
`maxLine`	the number of markers handled at a time, smaller value would reduce the memory cost
`row_names`	whether the numeric genotype has row names
`col_names`	whether the numeric genotype has column names
`type.geno`	the type of genotype elements
`auto_transpose`	whether to detecte the row and column
`verbose`	whether to print the reminder

Value

number of individuals and markers. Output files: genotype.desc, genotype.bin: genotype file in bigmemory format phenotype.phe: ordered phenotype file, same taxa order with genotype file map.map: SNP information

Examples

```
numericPath <- system.file("extdata", "04_numeric", "mvp.num", package = "rMVP")
mapPath <- system.file("extdata", "04_numeric", "mvp.map", package = "rMVP")
MVP.Data.Numeric2MVP(numericPath, mapPath, tempfile("outfile"))
```

`MVP.Data.PC` *Principal component analysis*

Description

Principal component analysis

Usage

```
MVP.Data.PC(
  filePC = TRUE,
  mvp_prefix = "mvp",
  K = NULL,
  out = NULL,
  pcs.keep = 5,
  maxLine = 10000,
  mrk_bycol = TRUE,
  sep = "\t",
  cpu = 1,
  verbose = TRUE
)
```

Arguments

`filePC`	Principal components, n*npc, n is sample size, npc is number of top columns of principal components
`mvp_prefix`	Prefix for mvp format files
`K`	Kinship matrix
`out`	prefix of output file name
`pcs.keep`	how many PCs to keep
`maxLine`	the number of markers handled at a time, smaller value would reduce the memory cost
`mrk_bycol`	whether the markers are stored by columns in genotype (i.e. genotype is a n by m matrix)
`sep`	seperator for PC file.
`cpu`	the number of cpu
`verbose`	whether to print detail.

Value

Output file: <out>.pc.bin <out>.pc.desc

Examples

```
geno <- file.path(system.file("extdata", "06_mvp-impute", package = "rMVP"), "mvp.imp")

MVP.Data.PC(TRUE, mvp_prefix=geno, out=tempfile("outfile"), cpu=1)
```

`MVP.Data.Pheno` *MVP.Data.Pheno: To clean up phenotype file Author: Haohao Zhang Build date: Sep 12, 2018*

Description

MVP.Data.Pheno: To clean up phenotype file Author: Haohao Zhang Build date: Sep 12, 2018

Usage

```
MVP.Data.Pheno(
  pheno_file,
  out = "mvp",
  cols = NULL,
  header = TRUE,
  sep = "\t",
  missing = c(NA, "NA", "-9", 9999),
  verbose = TRUE
)
```

Arguments

`pheno_file`	the name of phenotype file
`out`	the name of output file
`cols`	selected columns
`header`	whether the file contains header
`sep`	seperator of the file
`missing`	the missing value
`verbose`	whether to print detail.

Value

NULL Output files: cleaned phenotype file

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")

MVP.Data.Pheno(phePath, out=tempfile("outfile"))
```

`MVP.Data.VCF2MVP` *MVP.Data.VCF2MVP: To transform vcf data to MVP package Author: Haohao Zhang Build date: Sep 12, 2018*

Description

Accept the | or / separated markers, any variant sites that are not 0 or 1 will be considered NA.

Usage

```
MVP.Data.VCF2MVP(
  vcf_file,
  out = "mvp",
  maxLine = 10000,
  type.geno = "char",
  threads = 1,
  verbose = TRUE
)
```

Arguments

`vcf_file`	Genotype in VCF format
`out`	the name of output file
`maxLine`	the number of markers handled at a time, smaller value would reduce the memory cost
`type.geno`	the type of genotype elements
`threads`	number of thread for transforming
`verbose`	whether to print the reminder

Value

number of individuals and markers. Output files: genotype.desc, genotype.bin: genotype file in bigmemory format phenotype.phe: ordered phenotype file, same taxa order with genotype file map.map: SNP information

Examples

```
vcfPath <- system.file("extdata", "01_vcf", "mvp.vcf", package = "rMVP")

MVP.Data.VCF2MVP(vcfPath, tempfile("outfile"), threads=1)
```

`MVP.EMMA.Vg.Ve` *Estimate variance components using EMMA*

Description

Build date: August 30, 2016 Last update: January 27, 2017

Usage

```
MVP.EMMA.Vg.Ve(y, X, K, ngrids = 100, llim = -10, ulim = 10, esp = 1e-10)
```

Arguments

`y`	phenotype, n * 2
`X`	covariate matrix, the first column is 1s
`K`	Kinship matrix
`ngrids`	parameters for estimating vg and ve
`llim`	parameters for estimating vg and ve
`ulim`	parameters for estimating vg and ve
`esp`	parameters for estimating vg and ve

Value

Output: REML - maximum log likelihood Output: delta - exp(root) Output: ve - residual variance Output: vg - genetic variance

Author(s)

EMMA (Kang et. al. Genetics, 2008), Modified only for speed up by Xiaolei Liu and Lilin Yin

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")
phenotype <- read.table(phePath, header=TRUE)
print(dim(phenotype))
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
print(dim(genotype))

K <- MVP.K.VanRaden(genotype, cpu=1)
vc <- MVP.EMMA.Vg.Ve(y=phenotype[,2], X=matrix(1, nrow(phenotype)), K=K)
print(vc)
```

MVP.FarmCPU *Perform GWAS using FarmCPU method*

Description

Date build: Febuary 24, 2013 Last update: May 25, 2017 Requirement: Y, GD, and CV should have same taxa order. GD and GM should have the same order on SNPs

Usage

```
MVP.FarmCPU(
  phe,
  geno,
  map,
  CV = NULL,
  ind_idx = NULL,
  mrk_idx = NULL,
  P = NULL,
  method.sub = "reward",
  method.sub.final = "reward",
  method.bin = c("EMMA", "static", "FaST-LMM"),
  bin.size = c(5e+05, 5e+06, 5e+07),
  bin.selection = seq(10, 100, 10),
  memo = "MVP.FarmCPU",
  Prior = NULL,
  ncpus = 2,
  maxLoop = 10,
  maxLine = 5000,
  threshold.output = 0.01,
  converge = 1,
  iteration.output = FALSE,
  p.threshold = NA,
  QTN.threshold = 0.01,
  bound = NULL,
  verbose = TRUE
)
```

Arguments

phe	phenotype, n by t matrix, n is sample size, t is number of phenotypes
geno	genotype, either m by n or n by m is supportable, m is marker size, n is population size. This is Pure Genotype Data Matrix(GD). THERE IS NO COLUMN FOR TAXA.
map	SNP map information, m by 3 matrix, m is marker size, the three columns are SNP_ID, Chr, and Pos
CV	covariates, n by c matrix, n is sample size, c is number of covariates

ind_idx	the index of effective genotyped individuals
mrk_idx	the index of effective markers used in analysis
P	start p values for all SNPs
method.sub	method used in substitution process, five options: 'penalty', 'reward', 'mean', 'median', or 'onsite'
method.sub.final	method used in substitution process, five options: 'penalty', 'reward', 'mean', 'median', or 'onsite'
method.bin	method for selecting the most appropriate bins, three options: 'static', 'EMMA' or 'FaST-LMM'
bin.size	bin sizes for all iterations, a vector, the bin size is always from large to small
bin.selection	number of selected bins in each iteration, a vector
memo	a marker on output file name
Prior	prior information, four columns, which are SNP_ID, Chr, Pos, P-value
ncpus	number of threads used for parallele computation
maxLoop	maximum number of iterations
maxLine	the number of markers handled at a time, smaller value would reduce the memory cost
threshold.output	only the GWAS results with p-values lower than threshold.output will be output
converge	a number, 0 to 1, if selected pseudo QTNs in the last and the second last iterations have a certain probality (the probability is converge) of overlap, the loop will stop
iteration.output	whether to output results of all iterations
p.threshold	if all p values generated in the first iteration are bigger than p.threshold, FarmCPU stops
QTN.threshold	in second and later iterations, only SNPs with lower p-values than QTN.threshold have chances to be selected as pseudo QTNs
bound	maximum number of SNPs selected as pseudo QTNs in each iteration
verbose	whether to print detail.

Value

a m by 4 results matrix, m is marker size, the four columns are SNP_ID, Chr, Pos, and p-value

Author(s)

Xiaolei Liu and Zhiwu Zhang

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")
phenotype <- read.table(phePath, header=TRUE)
idx <- !is.na(phenotype[, 2])
phenotype <- phenotype[idx, ]
print(dim(phenotype))
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
genotype <- deepcopy(genotype, rows=idx)
print(dim(genotype))
mapPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.map", package = "rMVP")
map <- read.table(mapPath , head = TRUE)

farmcpu <- MVP.FarmCPU(phe=phenotype,geno=genotype,map=map,maxLoop=2,method.bin="static")
str(farmcpu)
```

`MVP.FaSTLMM.LL` *Evaluation of the maximum likelihood using FaST-LMM method*

Description

Last update: January 11, 2017

Usage

```
MVP.FaSTLMM.LL(pheno, snp.pool, X0 = NULL, ncpus = 2)
```

Arguments

`pheno`	a two-column phenotype matrix
`snp.pool`	matrix for pseudo QTNs
`X0`	covariates matrix
`ncpus`	number of threads used for parallel computation

Value

Output: beta - beta effect Output: delta - delta value Output: LL - log-likelihood Output: vg - genetic variance Output: ve - residual variance

Author(s)

Xiaolei Liu (modified)

MVP.GLM *To perform GWAS with GLM and MLM model and get the P value of SNPs*

Description

Build date: Aug 30, 2016 Last update: May 25, 2017

Usage

```
MVP.GLM(
  phe,
  geno,
  CV = NULL,
  ind_idx = NULL,
  mrk_idx = NULL,
  mrk_bycol = TRUE,
  maxLine = 5000,
  cpu = 1,
  verbose = TRUE
)
```

Arguments

phe	phenotype, n * 2 matrix
geno	genotype, either m by n or n by m is supportable, m is marker size, n is population size
CV	Covariance, design matrix(n * x) for the fixed effects
ind_idx	the index of effective genotyped individuals
mrk_idx	the index of effective markers used in analysis
mrk_bycol	whether the markers are stored by columns in genotype (i.e. M is a n by m matrix)
maxLine	the number of markers handled at a time, smaller value would reduce the memory cost
cpu	number of cpus used for parallel computation
verbose	whether to print detail.

Value

m * 2 matrix, the first column is the SNP effect, the second column is the P values

Author(s)

Lilin Yin and Xiaolei Liu

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")
phenotype <- read.table(phePath, header=TRUE)
idx <- !is.na(phenotype[, 2])
phenotype <- phenotype[idx, ]
print(dim(phenotype))
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
genotype <- deepcopy(genotype, rows=idx)
print(dim(genotype))

glm <- MVP.GLM(phe=phenotype, geno=genotype, cpu=1)
str(glm)
```

`MVP.HE.Vg.Ve` *To estimate variance component using HE regression*

Description

Build date: Feb 2, 2017 Last update: Feb 2, 2019

Usage

```
MVP.HE.Vg.Ve(y, X, K)
```

Arguments

y	phenotype
X	genotype
K	kinship matrix

Value

vg, ve, and delta

Author(s)

Translated from C++(GEMMA, Xiang Zhou) to R by: Haohao Zhang

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")
phenotype <- read.table(phePath, header=TRUE)
print(dim(phenotype))
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
print(dim(genotype))
```

```
K <- MVP.K.VanRaden(genotype, cpu=1)
vc <- MVP.HE.Vg.Ve(y=phenotype[,2], X=matrix(1, nrow(phenotype)), K=K)
print(vc)
```

`MVP.Hist` *Phenotype distribution histogram*

Description

Phenotype distribution histogram

Usage

```
MVP.Hist(
  phe,
  col = c("dodgerblue4", "olivedrab4", "violetred", "darkgoldenrod1", "purple4"),
  breakNum = 15,
  memo = NULL,
  outpath = getwd(),
  test.method = "auto",
  file.type = "pdf",
  file.output = TRUE,
  dpi = 300
)
```

Arguments

`phe`	phenotype data
`col`	The color vector of the histogram. If the number of colors is less than break.n, the color will be reused. If the number of colors is greater than break.n, only the previous break.n colors will be used.
`breakNum`	the number of cells for the histogram. The default value is 15.
`memo`	Character. A text marker on output files
`outpath`	Effective only when file.output = TRUE, determines the path of the output file
`test.method`	The method used to test the normal distribution. The options are "auto", "Shapiro-Wilk", "Kolmogorov-Smirnov", and NULL. When set to "auto", "Shapiro- Wilk" method, "Kolmogorov-Smirnov" method will be used when it is greater than 5000, and it will not be tested when set to NULL.
`file.type`	A string or NULL is used to determine the type of output file. Can be "jpg", "pdf", "tiff". If it is NULL, it will use `dev.new()` to create a new graphics device in the current environment, which may be RStudioGD or the default device of the system.
`file.output`	Logical value. If TRUE, the figures will be generated.
`dpi`	The resolution of the image, specifying how many pixels per inch.

Value

Output file: MVP.Phe_Distribution.<trait>.<type>

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")
phe <- read.table(phePath, header=TRUE)

MVP.Hist(phe, file.output = FALSE)
```

MVP.K.VanRaden *Calculate Kinship matrix by VanRaden method*

Description

Calculate Kinship matrix by VanRaden method

Usage

```
MVP.K.VanRaden(
  M,
  maxLine = 5000,
  ind_idx = NULL,
  mrk_idx = NULL,
  mrk_freq = NULL,
  mrk_bycol = TRUE,
  cpu = 1,
  verbose = TRUE,
  checkNA = TRUE
)
```

Arguments

M	genotype, either m by n or n by m is supportable, m is marker size, n is population size
maxLine	the number of markers handled at a time, smaller value would reduce the memory cost
ind_idx	the index of effective genotyped individuals used in analysis
mrk_idx	the index of effective markers used in analysis
mrk_freq	the prior calculated major allele frequency (not MAF) for all markers used in analysis
mrk_bycol	whether the markers are stored by columns in genotype (i.e. M is a n by m matrix)
cpu	the number of cpu
verbose	whether to print detail.
checkNA	whether to check NA in genotype.

Value

K, n * n matrix

Examples

```
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
print(dim(genotype))

K <- MVP.K.VanRaden(genotype, cpu=1)
```

`MVP.MLM` *To perform GWAS with GLM and MLM model and get the P value of SNPs*

Description

To perform GWAS with GLM and MLM model and get the P value of SNPs

Usage

```
MVP.MLM(
  phe,
  geno,
  K = NULL,
  eigenK = NULL,
  CV = NULL,
  ind_idx = NULL,
  mrk_idx = NULL,
  mrk_bycol = TRUE,
  REML = NULL,
  maxLine = 5000,
  cpu = 1,
  vc.method = c("BRENT", "EMMA", "HE"),
  verbose = TRUE
)
```

Arguments

`phe`	phenotype, n * 2 matrix
`geno`	genotype, either m by n or n by m is supportable, m is marker size, n is population size
`K`	Kinship, Covariance matrix(n * n) for random effects; must be positive semi-definite
`eigenK`	list of eigen Kinship

CV	covariates
ind_idx	the index of effective genotyped individuals
mrk_idx	the index of effective markers used in analysis
mrk_bycol	whether the markers are stored by columns in genotype (i.e. M is a n by m matrix)
REML	a list that contains ve and vg
maxLine	the number of markers handled at a time, smaller value would reduce the memory cost
cpu	number of cpus used for parallel computation
vc.method	the methods for estimating variance component("emma" or "he" or "brent")
verbose	whether to print detail.

Value

results: a m * 2 matrix, the first column is the SNP effect, the second column is the P values

Author(s)

Lilin Yin and Xiaolei Liu

Examples

```
phePath <- system.file("extdata", "07_other", "mvp.phe", package = "rMVP")
phenotype <- read.table(phePath, header=TRUE)
idx <- !is.na(phenotype[, 2])
phenotype <- phenotype[idx, ]
print(dim(phenotype))
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
genotype <- deepcopy(genotype, rows=idx)
print(dim(genotype))
K <- MVP.K.VanRaden(genotype, cpu=1)

mlm <- MVP.MLM(phe=phenotype, geno=genotype, K=K, cpu=1)
str(mlm)
```

MVP.PCA *Principal Component Analysis*

Description

Principal Component Analysis

Usage

```
MVP.PCA(
  M = NULL,
  K = NULL,
  maxLine = 10000,
  ind_idx = NULL,
  mrk_idx = NULL,
  mrk_bycol = TRUE,
  pcs.keep = 5,
  cpu = 1,
  verbose = TRUE
)
```

Arguments

M	genotype, either m by n or n by m is supportable, m is marker size, n is population size
K	kinship matrix
maxLine	the number of markers handled at a time, smaller value would reduce the memory cost
ind_idx	the index of effective genotyped individuals used in analysis
mrk_idx	the index of effective markers used in analysis
mrk_bycol	whether the markers are stored by columns in genotype (i.e. M is a n by m matrix)
pcs.keep	maximum number of PCs for output
cpu	the number of cpu
verbose	whether to print detail.

Value

Output: PCs - a n * npc matrix of top number of PCs, n is population size and npc is @param pcs.keep

Examples

```
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
genotype <- attach.big.matrix(genoPath)
print(dim(genotype))

pca <- MVP.PCA(M=genotype, cpu=1)
str(pca)
```

MVP.PCAplot *PCA Plot*

Description

PCA Plot

Usage

```
MVP.PCAplot(
  PCA,
  memo = "MVP",
  col = NULL,
  pch = NULL,
  class = NULL,
  legend.pos = "topright",
  Ncluster = 1,
  plot3D = FALSE,
  file.type = "pdf",
  dpi = 300,
  box = FALSE,
  file.output = TRUE,
  outpath = getwd(),
  verbose = TRUE
)
```

Arguments

PCA	Principal component analysis result, 2-column matrix
memo	the prefix of the output image file.
col	colors for each cluster
pch	Either an integer specifying a symbol or a single character to be used as the default in plotting points. See points for possible values and their interpretation. Note that only integers and single-character strings can be set as a graphics parameter (and not NA nor NULL).
class	the class of all individuals, for example: "breed", "location"
legend.pos	position of legend. default is "topright"
Ncluster	cluster number
plot3D	(DEPRECATED)if TRUE, plot PC figure in 3D format, it can be only used in windows and mac operation system, "rgl" package should be installed beforehead
file.type	Character. Options are jpg, pdf, and tiff
dpi	Number. Dots per inch for .jpg and .tiff files
box	Logical value. If TRUE, the border line of Manhattan plot will be added

file.output	Logical value. If TRUE, the figures will be generated.
outpath	Effective only when file.output = TRUE, determines the path of the output file
verbose	whether to print detail.

Value

Output file: MVP.PCA_2D.<type>

Examples

```
genoPath <- system.file("extdata", "06_mvp-impute", "mvp.imp.geno.desc", package = "rMVP")
geno <- attach.big.matrix(genoPath)
pca <- MVP.PCA(M=geno, cpu=1)

MVP.PCAplot(PCA=pca, Ncluster=3, class=NULL,
    col=c("red", "green", "yellow"), file.output=FALSE, pch=19)
```

MVP.Report *MVP.Report*

Description

MVP.Report

Usage

```
MVP.Report(
  MVP,
  col = c("#4197d8", "#f8c120", "#413496", "#495226", "#d60b6f", "#e66519", "#d581b7",
    "#83d3ad", "#7c162c", "#26755d"),
  bin.size = 1e+06,
  bin.range = NULL,
  pch = 19,
  band = 1,
  H = 1.5,
  ylim = NULL,
  cex.axis = 1,
  lwd.axis = 1.5,
  cex.lab = 1.5,
  plot.type = "b",
  multracks = FALSE,
  cex = c(0.5, 1, 1),
  r = 0.3,
  xlab = "Chromosome",
  ylab = expression(-log[10](italic(p))),
```

```
  xaxs = "i",
  yaxs = "r",
  outward = FALSE,
  threshold = NULL,
  threshold.col = "red",
  threshold.lwd = 1,
  threshold.lty = 2,
  amplify = FALSE,
  signal.cex = 1.5,
  signal.pch = 19,
  signal.col = "red",
  signal.line = 1,
  highlight = NULL,
  highlight.cex = 1.5,
  highlight.pch = 19,
  highlight.col = "green",
  chr.labels = NULL,
  chr.den.col = "black",
  cir.band = 1,
  cir.chr = TRUE,
  cir.chr.h = 1.5,
  cir.legend = TRUE,
  cir.legend.cex = 0.6,
  cir.legend.col = "black",
  LOG10 = TRUE,
  box = FALSE,
  conf.int = TRUE,
  file.output = TRUE,
  outpath = getwd(),
  file.type = "jpg",
  dpi = 300,
  height = NULL,
  width = NULL,
  memo = "",
  verbose = TRUE
)
```

Arguments

MVP	a dataframe or list, at least four columns. The first column is the name of SNP, the second column is the chromosome of SNP, the third column is the position of SNP, and the remaining columns are the P-value of each trait(Note:each trait a column).
col	a vector or a matrix, if "col" is a vector, each circle use the same colors, it means that the same chromosome is drewed in the same color, the colors are not fixed, one, two, three or more colors can be used, if the length of the "col" is shorter than the length the chromosome, then colors will be applied circularly. If "col" is a matrix, the row is the number of circles(traits), the columns are the colors

	that users want to use for different circles, each circle can be plotted in different number of colors, the missing value can be replaced by NA. For example: col=matrix(c("grey30","grey60",NA,"red","blue","green","orange",NA,NA),3,3,byrow=T).
bin.size	the size of bin for SNP_density plot.
bin.range	a vector, c(min, max). The min/max value of legend of SNP_density plot, the bin whose SNP number is smaller/bigger than 'bin.range' will be use the same color.
pch	a number, the type for the points or for traits of multi-traits Manhattan plot, is the same with "pch" in <plot>.
band	a number, the space between chromosomes, the default is 1(if the band equals to 0, then there would be no space between chromosomes).
H	a number, the height for each circle, each circle represents a trait, the default is 1.
ylim	a vector, the range of Y-axis when plotting the two type of Manhattan plots, is the same with "ylim" in <plot>.
cex.axis	a number, controls the size of ticks' numbers of X/Y-axis and the size of labels of circle plot.
lwd.axis	a number, controls the width of X/Y-axis lines.
cex.lab	a number, controls the size of labels of X/Y-axis.
plot.type	a character or vector, only "d", "c", "m", "q" or "b" can be used. if plot.type="d", SNP density will be plotted; if plot.type="c", only circle-Manhattan plot will be plotted; if plot.type="m",only Manhattan plot will be plotted; if plot.type="q",only Q-Q plot will be plotted;if plot.type="b", both circle-Manhattan, Manhattan and Q-Q plots will be plotted; if plot.type=c("m","q"), Both Manhattan and Q-Q plots will be plotted.
multracks	a logical,if multracks=FALSE, all Manhattan plots will be drew in separated files, if it is TRUE, all Manhattan plots will be plotted in only one file.
cex	a number or a vector, the size for the points, is the same with "size" in <plot>, and if it is a vector, the first number controls the size of points in circle plot(the default is 0.5), the second number controls the size of points in Manhattan plot(the default is 1), the third number controls the size of points in Q-Q plot(the default is 1)
r	a number, the radius for the circle(the inside radius), the default is 1.
xlab	a character, the labels for x axis.
ylab	a character, the labels for y axis.
xaxs	a character, The style of axis interval calculation to be used for the x-axis. Possible values are "r", "i", "e", "s", "d". The styles are generally controlled by the range of data or xlim, if given.
yaxs	a character, The style of axis interval calculation to be used for the y-axis. See xaxs above..
outward	logical, if outward=TRUE,then all points will be plotted from inside to outside for circular Manhattan plot.

threshold	a number or vector, the significant threshold. For example, Bonfferoni adjustment method: threshold=0.01/nrow(Pmap). More than one significant line can be added on the plots, if threshold=0 or NULL, then the threshold line will not be added.
threshold.col	a character or vector, the colour for the line of threshold levels.
threshold.lwd	a number or vector, the width for the line of threshold levels.
threshold.lty	a number or vector, the type for the line of threshold levels.
amplify	logical, CMplot can amplify the significant points, if amplify=T, then the points bigger than the minimal significant level will be amplified, the default: amplify=TRUE.
signal.cex	a number, if amplify=TRUE, users can set the size of significant points.
signal.pch	a number, if amplify=TRUE, users can set the shape of significant points.
signal.col	a character, if amplify=TRUE, users can set the colour of significant points, if signal.col=NULL, then the colors of significant points will not be changed.
signal.line	a number, the width of the lines of significant SNPs cross the circle.
highlight	a vector, names of SNPs which need to be highlighted.
highlight.cex	a number or vector, the size of points for SNPs which need to be highlighted.
highlight.pch	a number or vector, the pch of points for SNPs which need to be highlighted.
highlight.col	a number or vector, the col of points for SNPs which need to be highlighted.
chr.labels	a vector, the labels for the chromosomes of density plot and circle-Manhattan plot.
chr.den.col	a character or vector or NULL, the colour for the SNP density. If the length of parameter 'chr.den.col' is bigger than 1, SNP density that counts the number of SNP within given size('bin.size') will be plotted around the circle. If chr.den.col=NULL, the density bar will not be attached on the bottom of manhattan plot.
cir.band	a number, the space between circles, the default is 1.
cir.chr	logical, a boundary that represents chromosomes will be plotted on the periphery of a circle, the default is TRUE.
cir.chr.h	a number, the width for the boundary, if cir.chr=FALSE, then this parameter will be useless.
cir.legend	logical, whether to add the legend of each circle.
cir.legend.cex	a number, the size of the number of legend.
cir.legend.col	a character, the color of the axis of legend.
LOG10	logical, whether to change the p-value into log10(p-value).
box	logical, this function draws a box around the current Manhattan plot.
conf.int	logical, whether to plot confidence interval on QQ-plot.
file.output	a logical, users can choose whether to output the plot results.
outpath	Only when file.output = TRUE, determines the path of the output file
file.type	a character, users can choose the different output formats of plot, so for, "jpg", "pdf", "tiff" can be selected by users.

dpi	a number, the picture resolution for .jpg and .tiff files. The default is 300.
height	the height of output files.
width	the width of output files.
memo	add a character to the output file name.
verbose	whether to print the reminder.

Value

Output files

Examples

```
data(pig60K, package = "rMVP")

MVP.Report(pig60K[,c(1:3, 5)], plot.type="m",
    threshold=0.05/nrow(pig60K), file.output=FALSE)
```

`MVP.Report.Density` *SNP Density*

Description

SNP Density

Usage

```
MVP.Report.Density(
  Pmap,
  col = c("darkgreen", "yellow", "red"),
  dpi = 300,
  outpath = getwd(),
  memo = "MVP",
  bin.size = 1e+06,
  bin.max = NULL,
  file.type = "jpg",
  file.output = TRUE,
  verbose = TRUE
)
```

Arguments

Pmap	P value Map
col	The color vector
dpi	Number. Dots per inch for .jpg and .tiff files

outpath	Only when file.output = TRUE, determines the path of the output file
memo	Character. A text marker on output files
bin.size	the window size for counting SNP number
bin.max	maximum SNP number, for winows, which has more SNPs than bin.max, will be painted in same color
file.type	format of output figure
file.output	Whether to output the file
verbose	whether to print detail.

Value

Output file: <memo>.SNP_Density.<type>

Examples

```
data(pig60K, package = "rMVP")

MVP.Report.Density(pig60K, file.output=FALSE)
```

MVP.Report.QQplot *QQ Plot*

Description

QQ Plot

Usage

```
MVP.Report.QQplot(
  P.values,
  taxa_name,
  col = c("blue"),
  cex = 0.5,
  threshold = NULL,
  amplify = TRUE,
  signal.col = "red",
  signal.pch = 19,
  signal.cex = 0.8,
  conf.int = TRUE,
  cex.axis = 1,
  conf.int.col = "grey",
  threshold.col = "red",
  outpath = getwd(),
  file.type = "jpg",
  memo = "MVP",
```

```
  box = TRUE,
  dpi = 300,
  file.output = TRUE,
  verbose = TRUE
)
```

Arguments

P.values	P values
taxa_name	The identifier of the phenotype will be used to generate a portion of the image file name. If the title parameter is NULL, it will also be part of the title.
col	default color is "blue"
cex	A numerical value giving the amount by which plotting text and symbols should be magnified relative to the default. This starts as 1 when a device is opened, and is reset when the layout is changed, e.g. by setting mfrow. see par.
threshold	Number or Vector. The cutoff line on Manhattan plot, e.g. Bonfferoni correction. More than one significant line can be added onto one figure. If threshold=0 or NULL, the threshold line will not be added.
amplify	Logical value. If TRUE, the points that passed the threshold line will be highlighted
signal.col	Character. If "amplify" is TRUE, "signal.col" is used to set the color of significant points, if "signal.col" is NULL, the colors of significant points will not be changed
signal.pch	Number. If "amplify" is TRUE, users can set the type of significant points
signal.cex	Number. If "amplify" is TRUE, "signal.cex" is used to set the size of significant points
conf.int	Whether to draw a confidence interval
cex.axis	a number, controls the size of numbers of X-axis and the size of labels of circle plot.
conf.int.col	a character, the color of the confidence interval on QQ-plot.
threshold.col	Character or Vector. The colors of threshold lines
outpath	Only when file.output = TRUE, determines the path of the output file
file.type	A string or NULL is used to determine the type of output file. Can be "jpg", "pdf", "tiff". If it is NULL, it will use dev.new() to create a new graphics device in the current environment, which may be RStudioGD or the default device of the system.
memo	the prefix of the output image file.
box	A Boolean value that controls whether to draw a box around QQplot.
dpi	a number, the picture element for .jpg and .tiff files. The default is 300.
file.output	Logical value. If TRUE, the figures will be generated.
verbose	whether to print detail.

Value

Output file: <memo>.QQplot.<taxa_name>.<type>

Examples

```
data(pig60K, package = "rMVP")

MVP.Report(pig60K[1:10000,], plot.type="q", file.output=FALSE)
```

`MVP.Version` *Print MVP Banner*

Description

Build date: Aug 30, 2017 Last update: Dec 12, 2018

Usage

```
MVP.Version(width = 65, verbose = TRUE)
```

Arguments

`width`	the width of the message
`verbose`	whether to print detail.

Value

version number.

Author(s)

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Examples

```
MVP.Version()
```

pig60K *Genotyped by pig 60k chip*

Description

This dataset gives the results of Genome-wide association study of 3 traits, individuals were genotyped by pig 60K chip.

Usage

```
data(pig60K)
```

Format

A dataframe containing 3 traits' Pvalue

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