# Package 'POINT'

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Type Package

Title Protein Structure Guided Local Test
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<b>Description</b> Provides an implementation of a rare variant association test that utilizes protein tertiary structure to increase signal and to identify likely causal variants. Performs structure-guided collapsing, which leads to local tests that borrow information from neighboring variants on a protein and that provide association information on a variant-specific level. For details of the implemented method see West, R. M., Lu, W., Rotroff, D. M., Kuenemann, M., Chang, S-M., Wagner M. J., Buse, J. B., Motsinger-Reif, A., Fourches, D., and Tzeng, J-Y. (2019) <doi:10.1371 journal.pcbi.1006722="">.</doi:10.1371>
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Protein Structure Guided Local Test

#### **Description**

A rare variant association test that utilizes protein tertiary structure to increase signal and to identify likely causal variants. Performs structure-guided collapsing, which leads to local tests that borrow information from neighboring variants on a protein and that provide association information on a variant-specific level.

### Usage

```
point(
   yy,
   X,
   snp,
   proteinCoord,
   ...,
   trait = "binomial",
   cValues = c(0, 0.1, 0.2, 0.3, 0.4, 0.5),
   weighted = TRUE,
   weight = NULL,
   kernel = "linear",
   d = NULL,
   pvMethod = "davies",
   nperturb = 1000,
   verbose = TRUE
)
```

#### **Arguments**

уу	numeric vector; phenotype values.
Χ	numeric matrix; non-genetic covariates.
snp	numeric matrix; genotype snp matrix (count of minor alleles). Matrix cannot contain missing values.
proteinCoord	numeric matrix; columns correspond to 3 dimensional coordinates $(x,y,z)$ of each variant in the protein tertiary structure.
• • •	optional additional arguments for p-value methods CompQuadForm::davies and CompQuadForm::liu. \\
trait	character; type of phenotype data. Must be one of {'gaussian','binomial'} quantitative or case control data, respectively.
cValues	numeric vector; c values from which to choose the optimal neighborhood size for borrowing significant information.
weighted	logical; whether or not to weight the local kernel test using (non-distance based) weights.

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weight	numeric vector (optional) If NULL and weighted is TRUE (1.0-MAF)^24. Ignored if weighted is FALSE.
kernel	character; type of local kernel to use; Must be one of {'burden', 'linear', 'polynomial'}.
d	numeric; If kernel = 'poly', d is the order of the polynomial kernel.
pvMethod	character; method of calculating the p-value of each single marker test for fixed c values. Must be one of {'davies', 'liu'}.
nperturb	numeric, number of perturbations/resamples (perturbed test statistics) to calculate p-value of minP statistic.
verbose	logical; generate progress screen prints.

#### Value

Returns a matrix the rows of which correspond to individual markers. Columns correspond to:

- (1) minP statistic;
- (2) local kernel test p-value;
- (3) optimal scale value from input cValues;
- (4) minor allele frequency; and
- (5) single variant score test p-value.

#### **Examples**

```
# number of subjects
nsubj <- 1000
# number of markers
nm <- 5
# generate coordinates for proteins
protein <- cbind( stats::rnorm(n = nm, mean = 17.6, sd = 6.6),</pre>
                   stats::rnorm(n = nm, mean = 1.6, sd = 13.6),
                  stats::rnorm(n = nm, mean = 22.9, sd = 10.4)
# generate snp matrix
snp \leftarrow matrix(data = rbinom(n = nsubj*nm, size = 1, p = 0.02),
              nrow = nsubj, ncol = nm)
colnames(snp) = paste0("m",1:nm)
# generate binmoial response
MAF <- colMeans(x = snp)/2
causal <- numeric(nm)</pre>
causal[c(2,4)] <- 1.0
betaG <- 0.4*abs(log10(x = MAF))*causal
#no non-genetic covariates
X <- NULL
mu <- -0.05 + snp %*% betaG
pryy <- exp(mu)/(1+exp(mu))</pre>
```

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