

Package ‘redist’

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Title Markov Chain Monte Carlo Methods for Redistricting Simulation

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Description Enables researchers to sample redistricting plans from a pre-specified target distribution using a Markov Chain Monte Carlo algorithm. The package allows for the implementation of various constraints in the redistricting process such as geographic compactness and population parity requirements. The algorithm also can be used in combination with efficient simulation methods such as simulated and parallel tempering algorithms. Tools for analysis such as inverse probability reweighting and plotting functionality are included. The package implements methods described in Fifield, Higgins, Imai and Tarr (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo," working paper available at <<http://http://imai.princeton.edu/research/files/redist.pdf>>.

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redist-package	<i>R Package for the MCMC Redistricting Simulator</i>
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Description

redist implements methods developed by Fifield, Higgins, Imai and Tarr (2015) to randomly sample congressional redistricting plans using Markov Chain Monte Carlo methods. The current version of this package implements the basic simulator and offers several improvements to improve the performance of the algorithm and to incorporate substantive constraints found in American congressional redistricting. First, it allows users to draw plans that are nearly equal in population. Second, users can apply constraints that increase the geographic compactness of redistricting plans. Third, it implements several tempering techniques to help efficiently explore the full distribution of redistricting plans. Finally, it allows users to generate standard diagnostics from the Markov Chain Monte Carlo literature in order to examine the performance of the simulations.

Details

Package: redist
 Type: Package
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References

- Barbu, Adrian and Song-Chun Zhu. (2005) "Generalizing Swendsen-Wang to Sampling Arbitrary Posterior Probabilities." IEEE Transactions on Pattern Analysis and Machine Intelligence.
- Fifield, Benjamin, Michael Higgins, Kosuke Imai and Alexander Tarr. (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo." *Working Paper*. Available at <http://imai.princeton.edu/research/files/redist.pdf>.
- Swendsen, Robert and Jian-Sheng Wang. (1987) "Nonuniversal Critical Dynamics in Monte Carlo Simulations." Physical Review Letters.

algdat.p10	<i>All Partitions of 25 Precincts into 3 Congressional Districts (10% population constraint)</i>
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Description

This data set contains demographic and geographic information about 25 contiguous precincts in the state of Florida. The data lists all possible partitions of the 25 precincts into three contiguous congressional districts, conditional on the congressional districts falling within 10% of population parity.

Usage

```
data("algdat.p10")
```

Format

A list with five entries.

`adjlist` An adjacency list for the 25 precincts.

`cdmat` A matrix containing every partition of the 25 precincts into three contiguous congressional districts, with no population constraint.

`precinct.data` A matrix containing demographic information for each of the 25 precincts.

`segregation.index` A matrix containing the dissimilarity index of segregation (Massey and Denton 1987) for each congressional district map in `cdmat`.

`distancemat` A square matrix containing the squared distance between the centroids of any two precincts.

References

- Fifield, Benjamin, Michael Higgins, Kosuke Imai and Alexander Tarr. (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo." Working Paper. Available at <http://imai.princeton.edu/research/files/redist.pdf>.
- Massey, Douglas and Nancy Denton. (1987) "The Dimensions of Social Segregation". Social Forces.

Examples

```
## Not run:
data(algdat.p10)

## End(Not run)
```

algdat.p20

All Partitions of 25 Precincts into 3 Congressional Districts (20% Population Constraint)

Description

This data set contains demographic and geographic information about 25 contiguous precincts in the state of Florida. The data lists all possible partitions of the 25 precincts into three contiguous congressional districts, conditional on the congressional districts falling within 20% of population parity.

Usage

```
data("algdat.p20")
```

Format

A list with five entries.

`adjlist` An adjacency list for the 25 precincts.

`cdmat` A matrix containing every partition of the 25 precincts into three contiguous congressional districts, with no population constraint.

`precinct.data` A matrix containing demographic information for each of the 25 precincts.

`segregation.index` A matrix containing the dissimilarity index of segregation (Massey and Denton 1987) for each congressional district map in `cdmat`.

`distancemat` A square matrix containing the squared distance between the centroids of any two precincts.

References

Fifield, Benjamin, Michael Higgins, Kosuke Imai and Alexander Tarr. (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo." Working Paper. Available at <http://imai.princeton.edu/research/files/redist.pdf>.

Massey, Douglas and Nancy Denton. (1987) "The Dimensions of Social Segregation". Social Forces.

Examples

```
## Not run:
data(algdat.p20)

## End(Not run)
```

algdat.pfull	<i>All Partitions of 25 Precincts into 3 Congressional Districts (No Population Constraint)</i>
--------------	---

Description

This data set contains demographic and geographic information about 25 contiguous precincts in the state of Florida. The data lists all possible partitions of the 25 precincts into three contiguous congressional districts.

Usage

```
data("algdat.pfull")
```

Format

A list with five entries.

`adjlist` An adjacency list for the 25 precincts.

`cdmat` A matrix containing every partition of the 25 precincts into three contiguous congressional districts, with no population constraint.

`precinct.data` A matrix containing demographic information for each of the 25 precincts.

`segregation.index` A matrix containing the dissimilarity index of segregation (Massey and Denton 1987) for each congressional district map in `cdmat`.

`distancemat` A square matrix containing the squared distance between the centroids of any two precincts.

References

Fifield, Benjamin, Michael Higgins, Kosuke Imai and Alexander Tarr. (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo." Working Paper. Available at <http://imai.princeton.edu/research/files/redist.pdf>.

Massey, Douglas and Nancy Denton. (1987) "The Dimensions of Social Segregation". Social Forces.

Examples

```
## Not run:  
data(algdat.pfull)  
  
## End(Not run)
```

redist.diagplot *Diagnostic plotting functionality for MCMC redistricting.*

Description

redist.diagplot generates several common MCMC diagnostic plots.

Usage

```
redist.diagplot(sumstat,  
                plot = c("trace", "autocorr", "densplot",  
                        "mean", "gelmanrubin"),  
                logit = FALSE,  
                savename = NULL)
```

Arguments

sumstat	A vector, list, mcmc or mcmc.list object containing a summary statistic of choice.
plot	The type of diagnostic plot to generate: one of "trace", "autocorr", "densplot", "mean", "gelmanrubin". If plot = "gelmanrubin", the input sumstat must be of class mcmc.list or list.
logit	Flag for whether to apply the logistic transformation for the summary statistic. The default is FALSE.
savename	Filename to save the plot. Default is NULL.

Details

This function allows users to generate several standard diagnostic plots from the MCMC literature, as implemented by Plummer et. al (2006). Diagnostic plots implemented include trace plots, autocorrelation plots, density plots, running means, and Gelman-Rubin convergence diagnostics (Gelman & Rubin 1992).

Value

Returns a plot of file type .pdf.

References

Fifield, Benjamin, Michael Higgins, Kosuke Imai and Alexander Tarr. (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo." Working Paper. Available at <http://imai.princeton.edu/research/files/redist.pdf>.

Gelman, Andrew and Donald Rubin. (1992) "Inference from iterative simulations using multiple sequences (with discussion)." *Statistical Science*.

Plummer, Martin, Nicky Best, Kate Cowles and Karen Vines. (2006) "CODA: Convergence Diagnosis and Output Analysis for MCMC." *R News*.

Examples

```
## Not run:
data(algdat.pfull)

## Get an initial partition
set.seed(1)
initcids <- algdat.pfull$cdmat[,sample(1:ncol(algdat.pfull$cdmat), 1)]

## 25 precinct, three districts - no pop constraint ##
alg_253 <- redist.mcmc(adjobj = algdat.pfull$adjlist,
                      popvec = algdat.pfull$precinct.data$pop,
                      initcids = initcids,
                      nsims = 10000)

## Get Republican Dissimilarity Index from simulations
rep_dmi_253 <- redist.segcalc(alg_253,
                             algdat.pfull$precinct.data$repvote,
                             algdat.pfull$precinct.data$pop)

## Generate diagnostic plots
redist.diagplot(rep_dmi_253, plot = "trace")
redist.diagplot(rep_dmi_253, plot = "autocorr")
redist.diagplot(rep_dmi_253, plot = "densplot")
redist.diagplot(rep_dmi_253, plot = "mean")

## End(Not run)
```

redist.ipw

Inverse probability reweighting for MCMC Redistricting

Description

redist.ipw properly weights and resamples simulated redistricting plans so that the set of simulated plans resemble a random sample from the underlying distribution. redist.ipw is used to correct the sample when population parity, geographic compactness, or other constraints are implemented.

Usage

```
redist.ipw(algout,
           resampleconstraint = c("pop", "compact", "segregation", "similar"),
           targetbeta,
           targetpop = NULL,
           temper = 0)
```

Arguments

<code>algout</code>	An object of class "redist".
<code>resampleconstraint</code>	The constraint implemented in the simulations: one of "pop", "compact", "segregation", or "similar".
<code>targetbeta</code>	The target value of the constraint.
<code>targetpop</code>	The desired level of population parity. <code>targetpop = 0.01</code> means that the desired distance from population parity is 1%. The default is NULL.
<code>temper</code>	A flag for whether simulated tempering was used to improve the mixing of the Markov Chain. The default is 1.

Details

This function allows users to resample redistricting plans using inverse probability weighting techniques described in Rubin (1987). This techniques reweights and resamples redistricting plans so that the resulting sample is representative of a random sample from the uniform distribution.

Value

`redist.ipw` returns an object of class "redist". The object `redist` is a list that contains the following components (the inclusion of some components is dependent on whether tempering techniques are used):

<code>partitions</code>	Matrix of congressional district assignments generated by the algorithm. Each row corresponds to a geographic unit, and each column corresponds to a simulation.
<code>distance_parity</code>	Vector containing the maximum distance from parity for a particular simulated redistricting plan.
<code>mhdecisions</code>	A vector specifying whether a proposed redistricting plan was accepted (1) or rejected (0) in a given iteration.
<code>mhprob</code>	A vector containing the Metropolis-Hastings acceptance probability for each iteration of the algorithm.
<code>pparam</code>	A vector containing the draw of the p parameter for each simulation, which dictates the number of swaps attempted.
<code>constraint_pop</code>	A vector containing the value of the population constraint for each accepted redistricting plan.
<code>constraint_compact</code>	A vector containing the value of the compactness constraint for each accepted redistricting plan.
<code>constraint_segregation</code>	A vector containing the value of the segregation constraint for each accepted redistricting plan.
<code>constraint_similar</code>	A vector containing the value of the similarity constraint for each accepted redistricting plan.

- `beta_sequence` A vector containing the value of beta for each iteration of the algorithm. Returned when tempering is being used.
- `mhdecisions_beta` A vector specifying whether a proposed beta value was accepted (1) or rejected (0) in a given iteration of the algorithm. Returned when tempering is being used.
- `mhprob_beta` A vector containing the Metropolis-Hastings acceptance probability for each iteration of the algorithm. Returned when tempering is being used.

References

Fifield, Benjamin, Michael Higgins, Kosuke Imai and Alexander Tarr. (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo." Working Paper. Available at <http://imai.princeton.edu/research/files/redist.pdf>.

Rubin, Donald. (1987) "Comment: A Noniterative Sampling/Importance Resampling Alternative to the Data Augmentation Algorithm for Creating a Few Imputations when Fractions of Missing Information are Modest: the SIR Algorithm." *Journal of the American Statistical Association*.

See Also

[redist.mcmc](#) to simulate redistricting plans using Markov Chain Monte Carlo techniques.

Examples

```
## Not run:
data(algdat.p20)

## Code to run the simulations in Figure 4 of Fifield, Higgins,
## Imai and Tarr (2015)

## Get an initial partition
set.seed(1)
initcds <- algdat.p20$cdmat[,sample(1:ncol(algdat.p20$cdmat), 1)]

## Vector of beta weights
betaweights <- rep(NA, 10); for(i in 1:10){betaweights[i] <- 4^i}

## Run simulations - tempering population constraint
alg_253_20_st <- redist.mcmc(adjobj = algdat.p20$adjlist,
                           popvec = algdat.p20$precinct.data$pop,
                           initcds = initcds,
                           nsims = 10000,
                           betapop = -5.4,
                           betaweights = betaweights,
                           temperbetapop = 1)

## Resample using inverse probability weighting.
## Target distance from parity is 20
alg_253_20_st <- redist.ipw(alg_253_20_st,
                           resampleconstraint = "pop",
                           targetbeta = -5.4,
                           targetpop = .2,
```

```
temper = 1)
```

```
## End(Not run)
```

```
redist.mcmc
```

```
MCMC Redistricting Simulator
```

Description

redist.mcmc is used to simulate Congressional redistricting plans using Markov Chain Monte Carlo methods.

Usage

```
redist.mcmc(adjobj, popvec, nsims, ndists = NULL, initcds = NULL, loopscompleted = 0,
            nloop = 1, nthin = 1, eprob = 0.05, lambda = 0,
            popcons = NULL, grouppopvec = NULL, ssdmat = NULL,
            betacompact = 0, betapop = 0,
            betaseg = 0, betasimilar = 0,
            temperbetacompact = 0, temperbetapop = 0,
            temperbetaseg = 0, temperbetasimilar = 0,
            betaseq = "powerlaw", betaseqlength = 10,
            betaweights = NULL,
            adjswaps = TRUE, rngseed = NULL, maxiterrsg = 5000,
            savename = NULL, verbose = TRUE)
```

Arguments

adjobj	An adjacency matrix, list, or object of class "SpatialPolygonsDataFrame."
popvec	A vector containing the populations of each geographic unit.
nsims	The number of simulations run before a save point.
ndists	The number of congressional districts. The default is NULL.
initcds	A vector containing the congressional district labels of each geographic unit. The default is NULL. If not provided, random and contiguous congressional district assignments will be generated using redist.rsg.
loopscompleted	Number of save points reached by the algorithm. The default is 0.
nloop	The total number of save points for the algorithm. The default is 1. Note that the total number of simulations run will be nsims * nloop.
nthin	The amount by which to thin the Markov Chain. The default is 1.
eprob	The probability of keeping an edge connected. The default is 0.05.
lambda	The parameter determining the number of swaps to attempt each iteration of the algorithm. The number of swaps each iteration is equal to Pois(lambda) + 1. The default is 0.

popcons	The strength of the hard population constraint. popcons = 0.05 means that any proposed swap that brings a district more than 5% away from population parity will be rejected. The default is NULL.
grouppopvec	A vector of populations for some sub-group of interest. The default is NULL.
ssdmat	A matrix of squared distances between geographic units. The default is NULL.
betacompact	Strength of the constraint to impose geographic compactness. Values of betacompact less than zero impose more compactness. The default is 0 (no constraint).
betapop	Strength of the constraint to impose population parity. Values of betapop less than zero bring the simulated districts closer to population parity. The default is 0 (no constraint).
betaseg	Strength of the constraint to "pack" a particular subgroup into a single district. Values of betaseg greater than zero will create districts with higher concentrations of the subgroup. The default is 0 (no constraint).
betasimilar	Strength of the constraint to simulate redistricting similar to a specified plan. Values of betasimilar less than zero will create districts more similar to the specified plan. The default is 0 (no constraint).
temperbetacompact	Flag to apply simulated tempering to the betacompact constraint. The default is 0 (no tempering).
temperbetapop	Flag to apply simulated tempering to the betacompact constraint. The default is 0 (no tempering).
temperbetaseg	Flag to apply simulated tempering to the betaseg constraint. The default is 0 (no tempering).
temperbetasimilar	Flag to apply simulated tempering to the betasimilar constraint. The default is 0 (no tempering).
betaseq	Sequence of beta values for tempering. The default is powerlaw (see Fifield et al (2015) for details).
betaseqlength	Length of beta sequence desired for tempering. The default is 10.
betaweights	Sequence of weights for different values of beta. Allows the user to upweight certain values of beta over others. The default is NULL (equal weighting).
adjswaps	Flag to restrict swaps of beta so that only values adjacent to current constraint are proposed. The default is TRUE.
rngseed	Allows the user to set the seed for the simulations. Default is NULL.
maxiterrsg	Maximum number of iterations for random seed-and-grow algorithm to generate starting values. Default is 5000.
savename	Filename to save simulations. Default is NULL.
verbose	Whether to print initialization statement. Default is TRUE.

Details

This function allows users to simulate redistricting plans using Markov Chain Monte Carlo methods. Several constraints corresponding to substantive requirements in the redistricting process are implemented, including population parity and geographic compactness. In addition, the function includes multiple-swap and simulated tempering functionality to improve the mixing of the Markov Chain.

Value

redist.mcmc returns an object of class "redist". The object redist is a list that contains the following components (the inclusion of some components is dependent on whether tempering techniques are used):

partitions	Matrix of congressional district assignments generated by the algorithm. Each row corresponds to a geographic unit, and each column corresponds to a simulation.
distance_parity	Vector containing the maximum distance from parity for a particular simulated redistricting plan.
mhdecisions	A vector specifying whether a proposed redistricting plan was accepted (1) or rejected (0) in a given iteration.
mhprob	A vector containing the Metropolis-Hastings acceptance probability for each iteration of the algorithm.
pparam	A vector containing the draw of the p parameter for each simulation, which dictates the number of swaps attempted.
constraint_pop	A vector containing the value of the population constraint for each accepted redistricting plan.
constraint_compact	A vector containing the value of the compactness constraint for each accepted redistricting plan.
constraint_segregation	A vector containing the value of the segregation constraint for each accepted redistricting plan.
constraint_similar	A vector containing the value of the similarity constraint for each accepted redistricting plan.
beta_sequence	A vector containing the value of beta for each iteration of the algorithm. Returned when tempering is being used.
mhdecisions_beta	A vector specifying whether a proposed beta value was accepted (1) or rejected (0) in a given iteration of the algorithm. Returned when tempering is being used.
mhprob_beta	A vector containing the Metropolis-Hastings acceptance probability for each iteration of the algorithm. Returned when tempering is being used.

References

Fifield, Benjamin, Michael Higgins, Kosuke Imai and Alexander Tarr. (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo." Working Paper. Available at <http://imai.princeton.edu/research/files/redist.pdf>.

See Also

[redist.ipw](#) for inverse probability weighting functionality.

Examples

```
## Not run:
data(algdat.pfull)

## Code to run the simulations in Figure 4 in Fifield, Higgins, Imai and Tarr (2015)

## Get an initial partition
set.seed(1)
initcids <- algdat.pfull$cdmat[,sample(1:ncol(algdat.pfull$cdmat), 1)]

## Run the algorithm
alg_253 <- redist.mcmc(adjobj = algdat.pfull$adjlist,
                      popvec = algdat.pfull$precinct.data$pop,
                      initcids = initcids,
                      nsims = 10000)

## End(Not run)
```

redist.rsg

*Redistricting via Random Seed and Grow Algorithm***Description**

redist.rsg generates redistricting plans using a random seed a grow algorithm. This is the non-compact districting algorithm described in Chen and Rodden (2013). The algorithm can provide start values for the other redistricting routines in this package.

Usage

```
redist.rsg(adj.list, population, ndists, thresh, verbose = TRUE, maxiter=5000)
```

Arguments

adj.list	list of length N, where N is the number of precincts. Each list element is an integer vector indicating which precincts that precinct is adjacent to. It is assumed that precinct numbers start at 0.
population	numeric vector of list N, where N is the number of precincts. Each element lists the population total of the corresponding precinct, and is used to enforce population constraints.
ndists	integer, the number of districts we want to partition the precincts into.
thresh	numeric, indicating how close district population targets have to be to the target population before algorithm converges. thresh=0.05 for example means that all districts must be between 0.95 and 1.05 times the size of target.pop in population size.
verbose	boolean, indicating whether the time to run the algorithm is printed.

`maxiter` integer, indicating maximum number of iterations to attempt before convergence to population constraint fails. If it fails once, it will use a different set of start values and try again. If it fails again, `redist.rsg()` returns an object of all NAs, indicating that use of more iterations may be advised.

Value

list, containing three objects containing the completed redistricting plan.

- `district_membership` A vector of length N, indicating the district membership of each precinct.
- `district_list` A list of length Ndistrict. Each list contains a vector of the precincts in the respective district.
- `district_pop` A vector of length Ndistrict, containing the population totals of the respective districts.

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James Lo, <jameslo@princeton.edu>

Alexander Tarr, Department of Electrical Engineering, Princeton University <atarr@princeton.edu>

References

Jowei Chen and Jonathan Rodden (2013) “Unintentional Gerrymandering: Political Geography and Electoral Bias in Legislatures.” *Quarterly Journal of Political Science*. 8(3): 239-269.

Examples

```
## Not run:

### Real data example from test set

data("algdat.pfull")
res <- redist.rsg(algdat.pfull$adjlist, algdat.pfull$precinct.data$pop, 3, 0.05)

### Example that generates test data from a square map with equal population districts
### Number of precincts is Nrows*Ncols
### getTest() outputs an adjacency list out of specified rows and columns

genTest <- function(Nrows,Ncols){

NN <- Nrows * Ncols
```

```

geog <- matrix(NA,nrow=Nrows+2, ncol=Ncols+2)
geog[2:(Nrows+1), 2:(Ncols+1)] <- 0:(NN-1)

adj.list <- vector("list", NN)

for(i in 2:(Nrows+1)){
  for(j in 2:(Ncols+1)){
    adj.list[[ geog[i,j] + 1 ]] <- c(geog[i-1,j],geog[i+1,j],geog[i,j-1],geog[i,j+1])
  }
}
adj.list <- lapply(adj.list, na.omit)
adj.list <- lapply(adj.list, as.numeric)
return(adj.list)
}

### Generate a 100x100 precinct map and redistrict it into 10 districts
adj.list <- genTest(100,100)
population <- rep(300,length(adj.list))
tmp <- redist.rsg(adj.list, population, 10, 0.05)

## End(Not run)

```

redist.segcalc	<i>Segregation index calculation for MCMC redistricting.</i>
----------------	--

Description

redist.segcalc calculates the dissimilarity index of segregation (see Massey & Denton 1987 for more details) for a specified subgroup under any redistricting plan.

Usage

```

redist.segcalc(algout,
              grouppop,
              fullpop)

```

Arguments

algout	A matrix of congressional district assignments or a redist object.
grouppop	A vector of populations for some sub-group of interest.
fullpop	A vector containing the populations of each geographic unit.

Value

redist.segcalc returns a vector where each entry is the dissimilarity index of segregation (Massey & Denton 1987) for each redistricting plan in algout.

References

Fifield, Benjamin, Michael Higgins, Kosuke Imai and Alexander Tarr. (2015) "A New Automated Redistricting Simulator Using Markov Chain Monte Carlo." Working Paper. Available at <http://imai.princeton.edu/research/files/redist.pdf>.

Massey, Douglas and Nancy Denton. (1987) "The Dimensions of Social Segregation". Social Forces.

Examples

```
## Not run:
data(algdat.pfull)

## Code to run the simulations in Figure 4 of Fifield, Higgins,
## Imai and Tarr (2015)

## Get an initial partition
set.seed(1)
initcdfs <- algdat.pfull$cdmat[,sample(1:ncol(algdat.pfull$cdmat), 1)]

## Run simulations
alg_253 <- redist.mcmc(adjobj = algdat.pfull$adjlist,
                     popvec = algdat.pfull$precinct.data$pop,
                     initcdfs = initcdfs,
                     nsims = 10000)

## Get Republican Dissimilarity Index from simulations
rep_dmi_253 <- redist.segcalc(alg_253,
                             algdat.pfull$precinct.data$repvote,
                             algdat.pfull$precinct.data$pop)

## End(Not run)
```

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