# Package 'STrollR'

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<b>Title</b> Correct Standard Errors for Computing Spatial and Temporal Correlation post-estimation			
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<b>Description</b> A computationally efficient way to calculate covariance matrices that are corrected for spatial and temporal correlation.			
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bartlettSparse

Weighting Kernel

## Description

Weighting Kernel

## Usage

```
bartlettSparse(d, dmax)
```

## Arguments

d distance vector

dmax maximumm distance

## Value

vector of bartlett weights

df2stack

Convert list of dataframes with to rasterstack

## Description

Convert list of dataframes with to rasterstack

## Usage

```
df2stack(sim_i, DF)
```

## Arguments

sim\_i which simulation
DF list of dataframe

## Value

rasterstack

KNN 3

KNN

K nearest neighbours Calculate the number of neighbours within a neighbourhood.

#### **Description**

K nearest neighbours Calculate the number of neighbours within a neighbourhood.

#### Usage

```
KNN(w, h = w, type = "Moore")
```

#### **Arguments**

```
w number of neighbours wide (east-west).h number of neighbours long (north-south).type of neighbourhood; "Moore" or "VonNeumann"
```

#### Value

the number of nearest neighbours

#### **Examples**

```
KNN(4)
```

make\_spacetime\_data

Create Multivariate SpaceTime Data

## Description

Create Multivariate SpaceTime Data

```
make_spacetime_data(
  dimS,
  dimT,
  Κ,
  xpars = c(0, 1),
  theta = 1:K,
  space_groups = round(sqrt(dimS)),
  time_groups = round(sqrt(dimT)),
  error_type = "distance",
  error_scale = 1,
  xy_mat = expand.grid(Loc_X = seq(0, 1, length.out = dimS), Loc_Y = seq(1, 0, 1, length.out = dimS)
    length.out = dimS)),
  ws_mat = exp(-1 * as.matrix(dist(xy_mat))),
  wt_mat = toeplitz((2/3)^(0:(dimT - 1))),
  wst_mat = kronecker(wt_mat, ws_mat),
  verbose = F
)
```

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#### **Arguments**

dimS, dimT spatial dimension and temporal dimension Κ number of covariates xpars generate each Xvariable from uniform(xpar1,xpar2) parameter vector for Y=X%\*%theta theta time\_groups number of contiguous groups (e.g. countries, time-regimes) space\_groups, error\_type c('focal', 'distance') error\_scale variance of errors (normally distributed) matrix specifing xy locations xy\_mat ws\_mat matrix specifing correlation structer of error\_type='distance' matrix specifing correlation structer of error\_type='distance' wt\_mat wst\_mat=kronecker(wt\_mat, ws\_mat)

#### **Details**

'focal' creates errors based on  $e=u+\rho v$  'distance' creates errors base on drawing e from a multivariate normal with covariance matrix wst\_mat

#### Value

dataframe

#### **Examples**

```
DFst <- make_spacetime_data(11,6,2)
DFst[DFst$Space_ID==1,]</pre>
```

make\_space\_data

Create Multivariate Spatial Data

## Description

Create Multivariate Spatial Data

```
make_space_data(
  dimS,
  K,
  t_id = NA,
  space_groups = round(sqrt(dimS)),
  xpars = c(0, 1),
  theta = 1:K,
  error_type = "distance",
  error_scale = 1,
  sar_factor = 0.5,
```

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```
wf_mat = rbind(rep(sar_factor, 3), c(sar_factor, 0, sar_factor), rep(sar_factor, 3)),
  xy_mat = expand.grid(Loc_X = seq(0, 1, length.out = dimS), Loc_Y = seq(1, 0, 1, length.out = dimS)
    length.out = dimS)),
  ws_mat = exp(-as.matrix(dist(xy_mat)))
make_space_data.raster(
  dimS,
  Κ,
  t_id = NA,
  space_groups = round(sqrt(dimS)),
  xpars = c(0, 1),
  theta = 1:K,
  error_type = "distance",
  error_scale = 1,
  sar_factor = 0.5,
 wf_mat = rbind(rep(sar_factor, 3), c(sar_factor, 0, sar_factor), rep(sar_factor, 3)),
  xy_mat = expand.grid(Loc_X = seq(0, 1, length.out = dimS), Loc_Y = seq(1, 0,
    length.out = dimS)),
  ws_mat = exp(-as.matrix(dist(xy_mat)))
)
```

#### **Arguments**

dimS	spatial dimension
K	number of covariates
t_id	Time ID
space_groups	number of contiguous groups (e.g. countries)
xpars	generate each Xvariable from uniform(xpar1,xpar2)
theta	parameter vector for Y=X%*%theta
error_type	c('distance', 'focal', 'spherical')
error_scale	variance of errors (normally distributed)
sar_factor	parameter for error_type='focal'
wf_mat	matrix specifing weights to smooth when error_type='focal'
xy_mat	matrix specifing xy locations
ws_mat	matrix specifing correlation structer of error_type='distance'

#### **Details**

'focal' creates errors based on  $e=u+\rho v$ . 'distance' (the default) creates errors base on drawing e from a multivariate normal with covariance matrix ws\_mat.

## Value

dataframe

#### **Functions**

• make\_space\_data.raster: make\_space\_data.df uses raster functions instead of matrices (primarily for transparent debugging)

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#### **Examples**

```
library(STrollR)
DFs <- make_space_data.raster(10,2,error_type='distance')
DFs <- make_space_data.raster(10,2,error_type='focal')
DFs <- make_space_data.raster(10,2,error_type='spherical')
DFs2 <- make_space_data(10,2,error_type='distance')
DFs2 <- make_space_data(10,2,error_type='focal')
DFs2 <- make_space_data(10,2,error_type='spherical')</pre>
```

make\_time\_data

Create Multivariate Temporal Data

#### **Description**

Create Multivariate Temporal Data

#### Usage

```
make_time_data(
   dimT,
   K,
   s_id = NA,
   time_groups = round(sqrt(dimT)),
   xpars = c(0, 1),
   theta = 1:K,
   error_type = "distance",
   error_scale = 1,
   ar_factor = 2/3,
   wt_mat = toeplitz(c(1, ar_factor, ar_factor^2, rep(0, dimT - 3))),
   xy_mat = data.frame(Loc_X = NA, Loc_Y = NA)
)
```

## **Arguments**

```
dimT
                  temporal dimension
Κ
                  number of covariates
s_id
                  Spatial ID
                  number of contiguous groups (e.g. time-regimes)
time_groups
xpars
                  generate each Xvariable from uniform(xpar1,xpar2)
theta
                  parameter vector for Y=X%*%theta
                  c('focal', 'distance', 'spherical')
error_type
                  variance of errors (normally distributed)
error_scale
ar_factor
                  parameter for error_type='focal'
                  matrix specifing correlation structer of error_type='distance'
wt_mat
                  matrix specifing xy locations (defaults to NA)
xy_mat
```

#### **Details**

'focal' creates errors based on  $e=u+\rho v$ . 'distance' creates errors base on drawing e from a multivariate normal with covariance matrix wt\_mat.

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#### Value

dataframe

#### **Examples**

```
DFt <- make_time_data(6,2)</pre>
```

mkGif

Create Gifs

## Description

Create Gifs

## Usage

```
mkGif(
   DFlist,
   ti,
   ti,
   fdir = "~/Desktop/Packages/STrollR/STsim/",
   pname = "STvarX",
   ind = 1,
   vw = FALSE
)
```

#### **Arguments**

DFlist
ti number of time periods
fdir, pname directory and file
ind which simulation
vw view output

## Value

list of rasterstacks

model.frame.i

Model Frame with Regression Intercept

## Description

Model Frame with Regression Intercept

```
## S3 method for class 'i'
model.frame(reg, check_int = T)
```

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NEIGH

Calculate the weights objects

#### **Description**

Calculate the weights objects

## Usage

```
NEIGH(
  coord_sp,
  neigh = 1,
  knn = TRUE,
  adj = FALSE,
  dnn = FALSE,
  rast = FALSE
  vario = FALSE,
  tracer = TRUE,
  tr_type = "mult",
  tr_m = 20,
  tr_p = 16,
  symm = TRUE,
  symm_check = TRUE,
  SAVE = NA,
  write_gwt = F
)
```

#### **Arguments**

```
matrix of coordinates or a SpatialPoints object
coord_sp
                  number of neighbours to use in calculation
neigh
knn
                  calculate weights using knn approach
adj
                  calculate vonneumann weights (see VonNeumann)
                  dnn approach unsupported
dnn
rast
                  raster approach unsupported
                  is coord_sp a weights matrix?
vario
tracer
                  create trace matrix objects?
                  type of trace matrix
tr_type
                  trace matrix m
tr_m
tr_p
                  trace matrix p
                  make weights symmetric
symm
                  check for symmetric weights matrix
symm_check
SAVE
                  filename to save to, NA (default) returns as object
                  create GWT objects used in spdep or sphet
write_gwt
```

#### Value

filename of saved objects, or returns objects if SAVE=NA

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sim2stack

Convert simulation to rasterstack

## Description

Convert simulation to rasterstack

## Usage

```
sim2stack(e_spt, nsim, xyt)
```

## Arguments

e\_spt matrix of draws from spam.mvtnorm (each row a realization of a simulation)

xyt lattice structure
number of simulations

#### Value

list of rasterstacks

var2stack

Convert Dataframe with 1 variable to raster for one realization

## Description

Convert Dataframe with 1 variable to raster for one realization

## Usage

```
var2stack(df_i, sim_i)
```

## Arguments

df\_i dataframe

sim\_i which simulation

#### Value

raster

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varioJ

Variogram Calculation

## Description

Variogram Calculation

## Usage

```
varioJ(
  coords,
  cuttoff,
  E,
  latlon = FALSE,
  indices = FALSE,
  clean = FALSE,
  verbose = FALSE
```

## Arguments

coords coordinate matrix

cuttoff cutoff from which to calculate variogram

E vector of values (i.e. OLS residuals) associated coords

latlon coordinates are lon,lat or x,y

indices return indices? clean unused currently

verbose

## Value

data.frame of dij and (ei-ej)^2

vcovSHAC

Calculate a SHAC (Spatial Heteroskedastic and Autocorrelation Consistent) Variance Covariance Matrix

## Description

Calculate a SHAC (Spatial Heteroskedastic and Autocorrelation Consistent) Variance Covariance Matrix

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```
vcovSHAC(
  reg,
  wmat,
  method = "bruteforce_ll",
  cutoff_s,
  loc_lat,
  loc_lon,
  loc_y,
  loc_x,
  add_hc = T,
  add_cluster = F,
  add_hac = F,
  verbose = FALSE,
  manual_dist = T,
)
vcovSHACsep(
  reg,
  wmat,
  method = "bruteforce_llt",
  cutoff_s,
  cutoff_t,
  loc_lat,
  loc_lon,
  loc_y,
  loc_x,
  loc_t,
  add_hc = T,
  add_cluster = F,
  add_hac = F,
  verbose = FALSE,
  manual_dist = T,
)
vcovSTHAC(
  reg,
  wmat,
  method = "bruteforce_llt",
  cutoff_s,
  cutoff_t,
  loc_lat,
  loc_lon,
  loc_y,
  loc_x,
  loc_t,
  add_hc = T,
  add_cluster = F,
  add_hac = F,
  verbose = FALSE,
```

VonNeumann

```
manual_dist = T,
    ...
)
```

#### **Arguments**

```
an 'lm' object
reg
                  weights matrix
wmat
                  c(rolled', semirolled', 'bruteforce ll', 'bruteforce xy')
method
cutoff_s
                  include weights up to cutoff_s. Required to be in km if method='bruteforce_ll'
                  or in map-units if 'bruteforce_xy'.
loc_lat, loc_lon
                  required if method='bruteforce_ll'
loc_y, loc_x
                  required if method='bruteforce_xy'
add_hc, add_cluster, add_hac
                  logical for adding HC correction (default=T), Cluster correction (default=F),
                  HAC correction (default=F)
verbose
                  show messages
options
                  passed to sandwich
```

#### Value

covariance matrix

#### **Functions**

- vcovSHACsep: similar to vcovSTHAC, but treats spatial autocorrelation seperately within each time period
- vcovSTHAC: Space and Time HAC. Also weights the time-dimension according to bartlett kernel (i.e., weight= K(space)\*K(time)) for bartlett-kernel K.

#### **Examples**

```
library(STrollR)
DFs <- make_space_data(21,2)
reg <- lm(Y~X1+X2, data=DFs)
E <-
wmat <- weight_mat.df(DFs, cutoff_s=.5)
vcv <- vcovSHAC(reg, wmat, method='rolled')
##STrollR:::XOmegaX_rolled( as.matrix(DFs[,c('X1', 'X2')]), wmat, resid(reg))</pre>
```

VonNeumann

Compute VonNeumann Neighbours

#### **Description**

Compute VonNeumann Neighbours

```
VonNeumann(coord_sp, directions = 4)
```

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#### **Arguments**

coord\_sp SpatialPoints object or coordinate-matrix directions see adjacent

## Value

sparse weights matrix

weight\_mat

Compute Sparse Spatial Weights Matrix

## Description

Compute Sparse Spatial Weights Matrix

#### Usage

```
weight_mat(
   XY_Mat,
   latlon = FALSE,
   cutoff_s,
   cutoff_km2angles = FALSE,
   cutoff_angles2km = FALSE,
   verbose = F
)

weight_mat.df(
   DFs,
   xy_names = c("Loc_X", "Loc_Y"),
   latlon = FALSE,
   cutoff_s,
   cutoff_km2angles = FALSE,
   cutoff_angles2km = FALSE,
   verbose = F
)
```

#### **Arguments**

```
XY_Mat matrix of coordinates (either lat,lon or x,y).

latlon does XY_Mat have lat,lon coordinates?

cutoff_s use distances up cutoff_s. (if using generic map units, must specify latlon=cutoff_km2angles=cutoff_s

cutoff_km2angles,

cutoff_angles2km convert cutoff_s from km to angles or vice-versa?

verbose print messages
```

#### Value

the number of nearest neighbours

14 XOmegaX

#### **Functions**

• weight\_mat.df: weight\_mat.df is a wrapper of weight\_mat for dataframes

## **Examples**

```
weight_mat(expand.grid( list(x=1:10, y=1:10)), cutoff_s=.5)
```

XOmegaX

XOmegaX Meat Matrix Calculations

## Description

XOmegaX Meat Matrix Calculations

## Usage

```
XOmegaX_rolled(X, WMAT, E)
XOmegaX_semirolled(X, WMAT, E)
XOmegaX_bruteforce_ll(X, E, loc_lat, loc_lon, cutoff_s, manual_dist = T)
XOmegaX_bruteforce_xy(X, E, loc_y, loc_x, cutoff_s, manual_dist = T)
XOmegaX_bruteforce_llt(
  Χ,
  Ε,
  loc_lat,
  loc_lon,
  loc_t,
  cutoff_s,
  cutoff_t,
  manual\_dist = T
XOmegaX_bruteforce_xyt(
  Χ,
  Ε,
  loc_y,
  loc_x,
  loc_t,
  cutoff_s,
  cutoff_t,
  manual_dist = T
)
```

#### **Arguments**

```
X design matrix
WMAT weighting matrix (preferably sparse sparse)
```

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E vector of residuals

loc\_lat, loc\_lon spatial-coordinate vectors

cutoff\_s how far to extend bartlett kernel in spatial dimention (in km)

loc\_y, loc\_x spatial-coordinate vectors

loc\_t time-coordinate vector

cutoff\_t how far to extend bartlett kernel in time dimension (in time units)

#### **Details**

XOmegaX\_rolled Fully is for Sparse-Matrix Meat-Matrix Calculation (Whole Matrix At Once) XOmegaX\_semirolled is for Sparse-Matrix Meat-Matrix Calculation (One Row At A Time) XOmegaX\_bruteforce\_ll is for latlon data. XOmegaX\_bruteforce\_llt is for latlon and time data. XOmegaX\_bruteforce\_xy is for projected data. XOmegaX\_bruteforce\_xyt is for projected and time data. XOmegaX\_bruteforce\_\*t weights the time-dimension according to bartlett kernel

#### Value

object to be used in vcov\* functions

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