

Package ‘geocausal’

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Type Package

Title Causal Inference with Spatio-Temporal Data

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Description Spatio-temporal causal inference based on point process data.
You provide the raw data of locations and timings of treatment and outcome events, specify counterfactual scenarios, and the package estimates causal effects over specified spatial and temporal windows.
See Papadogeorgou, et al. (2022) <[doi:10.1111/rssb.12548](https://doi.org/10.1111/rssb.12548)>.

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airstrikes	<i>airstrikes</i>
------------	-------------------

Description

A subset of airstrikes data in Iraq (March to June 2007)

Usage

```
airstrikes
```

Format

A tibble with 3938 rows and 4 variables:

date Date (YYYY-MM-DD)

longitude Longitudes (decimal)

latitude Latitudes (decimal)

type Types of airstrikes (airstrikes or shows of force (SOF))

Examples

```
airstrikes
```

airstrikes_base	<i>airstrikes_base</i>
-----------------	------------------------

Description

A subset of airstrikes data in Iraq (a subset of airstrikes in 2006) that can be used to construct baseline densities

Usage

```
airstrikes_base
```

Format

A tibble with 808 rows and 3 variables:

date Date

longitude Longitudes (decimal)

latitude Latitudes (decimal)

Examples

```
airstrikes_base
```

conv_owin_into_sf	<i>Convert windows into sf objects</i>
-------------------	--

Description

'conv_owin_into_sf' takes an owin object and converts it to sf-related objects. This function is mostly an internal function of other functions.

Usage

```
conv_owin_into_sf(window)
```

Arguments

window owin object

Value

list of polygon, dataframe, sfc_POLYGON, sf, and SpatialPolygonsDataFrame objects

get_base_dens	<i>Get the baseline density</i>
---------------	---------------------------------

Description

'get_base_dens()' takes a dataframe and returns the baseline densities using Scott's rule of thumb (out-of-sample data) or fitting an inhomogeneous Poisson model (in-sample data) by regressing the in-sample data on time-invariant covariates.

Usage

```
get_base_dens(  
  window,  
  option,  
  ndim = 256,  
  out_data,  
  out_coordinates = c("longitude", "latitude"),  
  hfr,  
  dep_var,  
  indep_var,  
  ratio  
)
```

Arguments

window	owin object
option	"in" (using in-sample data) or "out" (using out-of-sample data)
ndim	the number of dimensions of grid cells (ndim^2). By default, $\text{ndim} = 256$.
out_data	dataframe (if using out-of-sample data)
out_coordinates	vector of column names of longitudes and latitudes (in this order) (if using in-sample data)
hfr	hyperframe (if using in-sample data)
dep_var	the name of the dependent variable (if using in-sample data)
indep_var	the names of time-invariant independent variables (if using in-sample data)
ratio	for random sampling of data (if using in-sample data)

Value

an im object of baseline density

get_cate	<i>Generate a Hajek estimator for heterogeneity analysis</i>
----------	--

Description

A function that returns a Hajek estimator of CATE for a spatial or spatio-temporal effect modifier

Usage

```
get_cate(
  obs,
  cf1,
  cf2,
  treat,
  pixel_count_out,
  lag,
  trunc_level = 0.95,
  time_after = TRUE,
  entire_window = NULL,
  em = NULL,
  E_mat = NULL,
  nbase = 6,
  spline_type = "ns",
  intercept = TRUE,
  eval_values = NULL,
  eval_mat = NULL,
  test_beta = NULL,
```

```

    bound = 1,
    save_weights = TRUE,
    ...
)

```

Arguments

obs	observed density
cf1	counterfactual density 1
cf2	counterfactual density 2
treat	column of a hyperframe that summarizes treatment data. In the form of 'hyperframe\$column'.
pixel_count_out	column of a hyperframe that summarizes the number of outcome events in each pixel
lag	integer that specifies lags to calculate causal estimates.
trunc_level	the level of truncation for the weights (0-1).
time_after	whether to include one unit time difference between treatment and outcome. By default = TRUE
entire_window	owin object (the entire region of interest)
em	treat column of a hyperframe that summarizes the effect modifier data. In the form of 'hyperframe\$column'. It can be NULL if E_mat is provided.
E_mat	optional covariance matrix (excluding the intercept) for the effect modifier. If provided, then the regression model will be based on this matrix. If 'intercept = TRUE', then a column of 1 will be add to 'E_mat'.
nbase	number of bases for splines
spline_type	type of splines. Either "ns" or "bs".
intercept	whether to include intercept in the regression model. Default is TRUE.
eval_values	a vector of values of the effect modifier for which CATE will be evaluated. Default is a 'seq(a,b,length.out=20)' where 'a' and 'b' are minimum and maximum values of the effect modifier.
eval_mat	evaluated spline basis (excluding the intercept) matrix at 'eval_values'. If 'intercept = TRUE', then a column of 1 will be add to 'eval_mat'.
test_beta	a vector of integers contain the indices of the coefficients that are included in the hypothesis test. By default, the null hypothesis is that all coefficient (except the intercept is 0). See details below
bound	either '1' or '2' specifying which bound estimator will be used. Default is '1'
save_weights	whether to save weights. Default is 'TRUE'
...	arguments passed onto the function

Details

‘E_mat’ should be a matrix or array of dimensions n by m where n is the product of image dimensions and number of time period, and m is ‘nbase’-‘intercept’. If you want to construct your own covariate matrix ‘E_mat’, you should use ‘get_em_vec()’ to convert the effect modifier(usually a column of a hyperframe) to a vector, and then construct the splines basis based on the vector. The covariate matrix ‘E_mat’ should not the column for intercept. The function ‘get_cate()’ will conduct a hypothesis testing on whether all the selected coefficients are 0. ‘test_beta’ is a vector of positive integers specifying the indices of the chosen beta. The coefficients (except the intercept) are indexed by ‘1,2,...,nbase-intercept’. By default, it test whether all the coefficients(except the intercept) are 0, and this is testing the the heterogeneity effect of the effect modifier.

Value

list of the following: ‘est_beta’: estimated regression coefficient ‘V_beta’: estimated asymptotic covariance matrix of regression coefficient (normalized by total time periods) ‘chisq_stat’: observed chi-square statistics for the hypothesis test ‘p.value’: observed chi-square statistics for the hypothesis test ‘specification’: information about the specification of the spline basis and the values on which the CATE is estimated ‘est_eval’: estimated CATE evaluated at chosen values ‘V_eval’: estimated asymptotic covariance matrix of the estimated CATE values (normalized by total time periods) ‘mean_effect’: Mean of the pseudo pixel effect ‘total_effect’: Mean of the pseudo effect for the window ‘entire_window’. It is equal to mean effect times the total number of pixels inside the chosen window

<code>get_cf_dens</code>	<i>Get counterfactual densities</i>
--------------------------	-------------------------------------

Description

‘get_cf_dens’ takes the target (expected) number, baseline density, and power density, and generates a hyperframe with counterfactual densities.

Usage

```
get_cf_dens(expected_number, base_dens, power_dens = NA, window)
```

Arguments

<code>expected_number</code>	the expected number of observations.
<code>base_dens</code>	baseline density (im object)
<code>power_dens</code>	power density (im object)
<code>window</code>	owin object

Details

There are two ways of generating counterfactual densities. First, users can keep the locations of observations as they are and change the expected number of observations. In this case, users do not have to set 'power_dens' and simply modify 'expected_number'. Alternatively, users can shift the locations as well. In this case, 'power_dens' must be specified. To obtain power densities, refer to [get_power_dens()].

Value

an im object of a counterfactual density

get_cf_sum_log_intens *Calculate the log counterfactual densities*

Description

A function that takes a hyperframe and returns the log counterfactual densities ie, the numerator of the equation

Usage

```
get_cf_sum_log_intens(cf_dens, treatment_data)
```

Arguments

cf_dens A counterfactual density (an im object)
treatment_data In the form of hyperframe\$column

Value

A numeric vector of sums of log densities for each time period

get_distexp *Get the expectation of treatment events with arbitrary distances*

Description

'get_distexp()' takes counterfactual densities and returns the expected number of treatment events based on distances from a user-specified focus.

Usage

```
get_distexp(
  cf_sim_results,
  entire_window,
  dist_map,
  dist_map_unit = "km",
  grayscale = FALSE,
  use_raw = FALSE
)
```

Arguments

`cf_sim_results` output of `'sim_cf_dens()'`

`entire_window` `owin` object of the entire region

`dist_map` `im` object whose cell values are the distance from a focus (e.g., city)

`dist_map_unit` either `"km"` or `"mile"`

`grayscale` logical. `'grayscale'` specifies whether to convert plot to grayscale (by default, `FALSE`).

`use_raw` logical. `'use_raw'` specifies whether to use the raw value of expectations or percentiles. By default, `'FALSE'`.

Value

A list of `ggplot` objects that summarizes how expectations change over distances from a focus (`'expectation_plot'`) and summarizes distances and areas (`'window_plot'`). Note that the second object can not necessarily be well drawn depending on how windows are defined.

<code>get_dist_focus</code>	<i>Get distance maps</i>
-----------------------------	--------------------------

Description

`'get_dist_focus()'` generates a distance map from focus locations.

Usage

```
get_dist_focus(window, lon, lat, resolution, mile = FALSE, preprocess = FALSE)
```

Arguments

`window` `owin` object

`lon` vector of longitudes

`lat` vector of latitudes

`resolution` resolution of raster objects

mile	logical. 'mile' specifies whether to return the output in miles instead of kilometers (by default, FALSE).
preprocess	logical. 'preprocess' specifies whether to first pick the potentially closest point. It is recommended to set 'preprocess = TRUE' if users need to obtain distances from many points.

Details

'get_dist_focus()' depends on 'geosphere::distVincentyEllipsoid()'. Since it calculates accurate distances considering the ellipsoid, the process sometimes becomes computationally demanding, namely when we need to obtain distances from many points. In that case, users can set 'preprocess = TRUE'. With this option, 'get_dist_focus()' calculates distances from points by first identifying the closest point using 'sf::st_nearest_feature()' with approximations. This process is more efficient than computing distances from all the points with 'geosphere::distVincentyEllipsoid()' and then obtaining the minimum of all the distances. By default, 'get_dist_focus()' returns distances in kilometers unless users set 'mile = TRUE'.

Value

an im object

get_dist_line	<i>Get distance maps from lines and polygons</i>
---------------	--

Description

'get_dist_line()' generates a distance map from lines and polygons.

Usage

```
get_dist_line(
  window,
  path_to_shapefile,
  line_data = NULL,
  mile = FALSE,
  resolution,
  preprocess = TRUE
)
```

Arguments

window	owin object
path_to_shapefile	path to shapefile
line_data	sfc_MULTILINESTRING file (If available. If not, 'get_dist_line()' creates it from a shapefile.)

mile	logical. 'mile' specifies whether to return the output in miles instead of kilometers (by default, FALSE).
resolution	resolution of raster objects
preprocess	logical. 'preprocess' specifies whether to first pick the potentially closest point. It is recommended to set 'preprocess = TRUE' if users need to obtain distances from many points.

Value

an im object

get_elev	<i>Get elevation data</i>
----------	---------------------------

Description

'get_elevation()' takes a directory that hosts shapefile and returns an owin object of altitudes.

Usage

```
get_elev(load_path, ...)
```

Arguments

load_path	path to the shp file (note: a folder)
...	other parameters passed to 'elevatr::get_elev_raster()'. The resolution argument z must be specified.

Value

an im object (unit: meters)

get_em_vec	<i>convert a list of im objects to a vector</i>
------------	---

Description

'get_em_vec()' get the vector form of a column of a hyperframe that summarizes the effect modifier data in heterogeneity analysis

Usage

```
get_em_vec(em, time_after = TRUE, lag, entire_window = NULL, ngrid = NULL)
```

Arguments

em	column of a hyperframe that summarizes effect modifier data. In the form of 'hyperframe\$column'.
time_after	whether to include one unit time difference between treatment and outcome. By default = TRUE
lag	integer that specifies lags to calculate causal estimates
entire_window	owin object (the entire region of interest). If given, then the values outside the region will be set to 'NA'.
ngrid	a number or a vector of two numbers that specify the dimension of pixels. If NULL, the pixel dimension of the original images will not be changed

Details

The function 'get_em_vec()' get the vector form of the effect modifier in the heterogeneity analysis. It is useful if you want to construct the variance matrix 'E_mat' that is passed to the function 'get_cate()'

get_est

Get causal estimates comparing two scenarios

Description

'get_est()' generates causal estimates comparing two counterfactual scenarios.

Usage

```
get_est(
  obs,
  cf1,
  cf2,
  treat,
  sm_out,
  mediation = FALSE,
  obs_med_log_sum_dens = NA,
  cf1_med_log_sum_dens = NA,
  cf2_med_log_sum_dens = NA,
  lag,
  time_after = TRUE,
  entire_window,
  use_dist,
  windows,
  dist_map,
  dist,
  trunc_level = NA,
  save_weights = TRUE
)
```

Arguments

obs	observed density
cf1	counterfactual density 1
cf2	counterfactual density 2
treat	column of a hyperframe that summarizes treatment data. In the form of 'hyperframe\$column'.
sm_out	column of a hyperframe that summarizes the smoothed outcome data
mediation	whether to perform causal mediation analysis (don't use; still in development). By default, FALSE.
obs_med_log_sum_dens	sum of log densities of mediators for the observed (don't use; still in development)
cf1_med_log_sum_dens	sum of log densities of mediators for counterfactual 1 (don't use; still in development)
cf2_med_log_sum_dens	sum of log densities of mediators for counterfactual 2 (don't use; still in development)
lag	integer that specifies lags to calculate causal estimates
time_after	whether to include one unit time difference between treatment and outcome. By default = TRUE
entire_window	owin object (the entire region of interest)
use_dist	whether to use distance-based maps. By default, TRUE
windows	a list of owin objects (if 'use_dist = FALSE')
dist_map	distance map (an im object, if 'use_dist = TRUE')
dist	distances (a numeric vector within the max distance of 'dist_map')
trunc_level	the level of truncation for the weights (0-1)
save_weights	whether to save weights

Details

The level of truncation indicates the quantile of weights at which weights are truncated. That is, if 'trunc_level = 0.95', then all weights are truncated at the 95 percentile of the weights.

Value

list of the following: 'cf1_ave_surf': average weighted surface for scenario 1 'cf2_ave_surf': average weighted surface for scenario 2 'est_cf': estimated effects of each scenario 'est_causal': estimated causal contrasts 'var_cf': variance upper bounds for each scenario 'var_causal': variance upper bounds for causal contrasts 'windows': list of owin objects

get_estimates	<i>Generate a Hajek estimator</i>
---------------	-----------------------------------

Description

A function that returns a Hajek estimator of causal contrasts

Usage

```
get_estimates(
  weighted_surf_1,
  weighted_surf_2,
  use_dist = TRUE,
  windows,
  dist_map,
  dist,
  entire_window
)
```

Arguments

weighted_surf_1	a weighted surface for scenario 1
weighted_surf_2	another weighted surface for scenario 2
use_dist	whether to use distance-based maps. By default, TRUE
windows	a list of owin objects (if 'use_dist = FALSE')
dist_map	distance map (an im object, if 'use_dist = TRUE')
dist	distances (a numeric vector within the max distance of 'dist_map')
entire_window	an owin object of the entire map

Details

'get_estimates()' is an internal function to 'get_est()' function, performing the estimation analysis after 'get_weighted_surf()' function

Value

list of Hajek estimators for each scenario ('est_haj'), causal contrasts (Hajek estimator) as a matrix ('est_tau_haj_matrix'), and causal contrast (scenario 2 - scenario 1) as a numeric vector ('est_tau_haj_cf2_vs_cf1'), along with weights, windows, and smoothed outcomes

get_hfr	<i>Create a hyperframe</i>
---------	----------------------------

Description

‘get_hfr()’ takes a dataframe with time and location variables and generates a hyperframe with point patterns. ‘get_hfr()’ is usually the first function that users employ in order to perform spatiotemporal causal inference analytic methods.

Usage

```
get_hfr(
  data,
  col,
  window,
  time_col,
  time_range,
  coordinates = c("longitude", "latitude"),
  combine = TRUE
)
```

Arguments

data	dataframe. The dataframe must have time and location variables. Location variables should be standard coordinates (i.e., longitudes and latitudes).
col	the name of the column for subtypes of events of interest
window	owin object (for more information, refer to ‘spatstat.geom::owin()’). Basically, an owin object specifies the geographical boundaries of areas of interest.
time_col	the name of the column for time variable. Note that the time variable must be integers.
time_range	numeric vector. ‘time_range’ specifies the range of the time variable (i.e., min and max of the time variable). The current version assumes that the unit of this time variable is dates.
coordinates	character vector. ‘coordinates’ specifies the names of columns for locations. By default, ‘c("longitude", "latitude")’ in this order. Note that the coordinates must be in decimal degree formats.
combine	logical. ‘combine’ tells whether to generate output for all subtypes of events combined. By default, ‘TRUE’, which means that a column of ppp objects with all subtypes combined is generated in the output.

Value

A hyperframe is generated with rows representing time and columns representing the following:

- * The first column: time variable
- * The middle columns: ppp objects (see ‘spatstat.geom::ppp()’) generated for each subtype of events of interest
- * The last column (if ‘combine = TRUE’): ppp objects with all subtypes combined. This column is named as ‘all_combined’.

Examples

```
# Data
dat <- data.frame(time = c(1, 1, 2, 2),
                  longitude = c(43.9, 44.5, 44.1, 44.0),
                  latitude = c(33.6, 32.7, 33.6, 33.5),
                  type = rep(c("treat", "out"), 2))

# Hyperframe
get_hfr(data = dat,
        col = "type",
        window = iraq_window,
        time_col = "time",
        time_range = c(1, 2),
        coordinates = c("longitude", "latitude"),
        combine = FALSE)
```

`get_hist`*Obtain histories of treatment or outcome events*

Description

`'get_hist()'` takes a hyperframe and time and columns of interest, and generates histories of events of interest.

Usage

```
get_hist(tt, Xt, Yt = NA, lag, window, x_only = TRUE)
```

Arguments

<code>tt</code>	values of the time variable of interest for which <code>'get_hist()'</code> generates histories
<code>Xt</code>	the name of a treatment column
<code>Yt</code>	the name of an outcome column
<code>lag</code>	numeric. <code>'lag'</code> specifies the number of time periods over which <code>'get_hist()'</code> aggregates treatment and outcome columns.
<code>window</code>	owin object.
<code>x_only</code>	logical. <code>'x_only'</code> specifies whether to generate only treatment history (no outcome history). By default, <code>'FALSE'</code> .

Value

list of treatment and outcome histories

Examples

```

dat_out <- insurgencies[1:100, ]
dat_out$time <- as.numeric(dat_out$date - min(dat_out$date) + 1)

# Hyperframe
dat_hfr <- get_hfr(data = dat_out,
                  col = "type",
                  window = iraq_window,
                  time_col = "time",
                  time_range = c(1, max(dat_out$time)),
                  coordinates = c("longitude", "latitude"),
                  combine = TRUE)

# Histories
lapply(1:nrow(dat_hfr), get_hist,
       Xt = dat_hfr$all_outcome,
       lag = 1, window = iraq_window)

```

get_obs_dens	<i>Generate observed densities</i>
--------------	------------------------------------

Description

‘get_obs_dens()’ takes a hyperframe and returns observed densities. The output is used as propensity scores.

Usage

```
get_obs_dens(hfr, dep_var, indep_var, ngrid = 100, window)
```

Arguments

hfr	hyperframe
dep_var	The name of the dependent variable. Since we need to obtain the observed density of treatment events, ‘dep_var’ should be the name of the treatment variable.
indep_var	vector of names of independent variables (covariates)
ngrid	the number of grid cells that is used to generate observed densities. By default = 100. Notice that as you increase ‘ngrid’, the process gets computationally demanding.
window	owin object

Details

‘get_obs_dens()’ assumes the poisson point process model and calculates observed densities for each time period. It depends on ‘spatstat.model::mppm()’. Users should note that the coefficients in the output are not directly interpretable, since they are the coefficients inside the exponential of the poisson model.

Value

list of the following: * 'indep_var': independent variables * 'coef': coefficients * 'intens_grid_cells': im object of observed densities for each time period * 'estimated_counts': the number of events that is estimated by the poisson point process model for each time period * 'sum_log_intens': the sum of log intensities for each time period

get_power_dens	<i>Get power densities</i>
----------------	----------------------------

Description

'get_power_dens()' takes the target densities and their priorities and returns a power density.

Usage

```
get_power_dens(target_dens, priorities, window)
```

Arguments

target_dens	list of target densities
priorities	vector of priorities for each of target densities
window	owin object

Value

list of an im object and a ggplot object of power densities

get_var_bound	<i>Calculate variance upper bounds</i>
---------------	--

Description

A function that calculates variance upper bounds

Usage

```
get_var_bound(estimates, bound_est = 1)
```

Arguments

estimates	an object returned from 'get_est()' function
bound_est	an integer specifying the estimator for the asymptotic variance bound. Use '1' for the first estimator and '2' for the second estimator. Default is '1'. See details.

Details

'get_var_bound()' is an internal function to 'get_estimates()' function, performing the estimation analysis after 'get_est()' function. There are two consistent estimators for the asymptotic variance bound. The first estimator (default, 'bound_est=1') performs well in scenarios with low variability, but exhibit some under coverage issue for high variability scenarios where longer time series are needed for getting robust estimator. The second estimator('bound_est=2') yields good coverage rates in simulations, but gives unreasonably large bound when there 'are extremely large IPW weights. Users should carefully consider the characteristics of their data when selecting the estimator.

Value

list of variance upper bounds for each scenario ('bound_haj') and causal contrasts ('bound_tau_haj'). Note that this function returns variance upper bounds for Hajek estimators

get_weighted_surf	<i>Generate average weighted surfaces</i>
-------------------	---

Description

A function that returns averaged weighted surfaces (both IPW and Hajek) along with weights

Usage

```
get_weighted_surf(
  obs_dens,
  cf_dens,
  mediation = FALSE,
  cate = FALSE,
  obs_med_log_sum_dens,
  cf_med_log_sum_dens,
  treatment_data,
  smoothed_outcome,
  lag,
  entire_window,
  time_after,
  truncation_level = truncation_level
)
```

Arguments

obs_dens	observed density
cf_dens	counterfactual density
mediation	whether to perform causal mediation analysis. By default, FALSE.
cate	whether to perform the heterogeneity analysis. By default, FALSE.

obs_med_log_sum_dens	sum of log densities of mediators for the observed (if 'mediation = TRUE')
cf_med_log_sum_dens	sum of log densities of mediators for counterfactual (if 'mediation = TRUE')
treatment_data	column of a hyperframe that summarizes treatment data. In the form of 'hyperframe\$column'.
smoothed_outcome	column of a hyperframe that summarizes the smoothed outcome data
lag	integer that specifies lags to calculate causal estimates
entire_window	owin object (the entire region of interest)
time_after	whether to include one unit time difference between treatment and outcome
truncation_level	the level at which the weights are truncated (see 'get_estimates()')

Details

'get_weighted_surf()' is an internal function to 'get_estimates()' function. If 'time_after' is TRUE, then this function uses treatment data and weights from lag to nrow(data)-1, and outcome data from lag+1 to nrow(data).

Value

list of an average weighted surface ('avarage_surf', an 'im' object), a Hajek average weighted surface ('average_weighted_surf_haj', an 'im' object), weights, and smoothed outcomes

get_window	<i>Generate a window</i>
------------	--------------------------

Description

'get_window()' takes a directory that hosts a shapefile and returns an owin object.

Usage

```
get_window(load_path)
```

Arguments

load_path path to the shp file

Value

owin object

imls_to_arr	<i>convert a list of im objects to a three-dimensional array</i>
-------------	--

Description

'imls_to_arr()' convert a list of im object to a 3D array

Usage

```
imls_to_arr(imls, start = 1, end = NULL, entire_window = NULL, ngrid = NULL)
```

Arguments

imls	a list of im objects (imlist)
start	the index of the first im to be converted. Default is 1.
end	the index of the last im to be converted. If not provided, then it will be set to the length of the list.
entire_window	a owin object. If given, then the values outside the region will be set to 'NA'
ngrid	an optional arugument that takes one integer or vector of two integers specifying the dimensions of the 'im' objects. If provided, the dimensions of the objects will be adjusted to 'ngrid' before the conversion to the array.

Details

'imls_to_arr()' is a internal function for 'imls_to_vec()'. By default, it returns a three-dimensional array of dimension n by m by l where n and m are the dimensions of the im objects, and l is the length of the list. All the im objects in the list need to have the same dimensions.

insurgencies	<i>insurgencies</i>
--------------	---------------------

Description

A subset of insurgencies data in Iraq (March to June 2007)

Usage

```
insurgencies
```

Format

A tibble with 68573 rows and 4 variables:

date Date (YYYY-MM-DD)

longitude Longitudes (decimal)

latitude Latitudes (decimal)

type Types of insurgencies (improvised explosive devices (IED), small arms fire (SAF), or other)

Examples

insurgencies

iraq_window	<i>iraq_window</i>
-------------	--------------------

Description

An owin object of Iraq

Usage

iraq_window

Format

A polygonal object:

type Polygonal

xrange Range (longitude)

yrange Range (latitude)

bdry Boundaries

units Units

Examples

iraq_window

pixel_count_ppp	<i>Get number of events in a pixel</i>
-----------------	--

Description

'pixel_count_ppp()' takes a column of hyperframes (ppp objects) and gets the number of events in each pixel.

Usage

```
pixel_count_ppp(
  data,
  ngrid = c(128, 128),
  W = NULL,
  weights = NULL,
  DivideByPixelArea = FALSE,
  ...
)
```

Arguments

data	the name of a hyperframe and column of interest.
ngrid	a number or a vector of two numbers specifying the pixel array dimensions. A single integer, or an integer vector of length 2 giving dimensions in the y and x directions. Default is 'c(128,128)'.
W	Optional window mask (object of class "owin") determining the pixel raster. 'data' should be in the form of "hyperframe\$column".
weights	Optional vector of weights associated with the points.
DivideByPixelArea	Logical value determining whether the resulting pixel values should be divided by the pixel area. Default value is 'False'.
...	parameters passed on to the function.

Value

im objects

Examples

```
# Time variable
dat_out <- insurgencies[1:100, ]
dat_out$time <- as.numeric(dat_out$date - min(dat_out$date) + 1)

# Hyperframe
dat_hfr <- get_hfr(data = dat_out,
                  col = "type",
                  window = iraq_window,
                  time_col = "time",
                  time_range = c(1, max(dat_out$time)),
                  coordinates = c("longitude", "latitude"),
                  combine = TRUE)

# Get the number of events for each pixel
pixel_count_ppp(data = dat_hfr$all_combined)
```

plot.cate

Plot estimated CATE

Description

Plot estimated CATE

Usage

```
## S3 method for class 'cate'
plot(
  x,
  ...,
  result = "cate",
  type = "l",
  scale = 1,
  xrange = NULL,
  main = "",
  xlab = "",
  ylim = NULL
)
```

Arguments

x	input
...	arguments passed on to the function
result	specify which values will be used for plot. Default is "cate" - If 'result' is "cate", then estimated cate values will be used - If 'result' is "beta", then the estimated regression coefficients will be used
type	The type of plot to draw. This argument will be ignored if 'result' = "beta". Default is "l". - If 'type' is "p", points with error bars will be drawn. - If 'type' is "l", lines with shaded region will be drawn. - If 'type' is a vector of strings, each element specifies the type for the corresponding 'eval_values' value.
scale	a positive number specifying the scale by which the estimates will be scaled. If provided, the estimates will be scaled by this value. Default is NULL, which means no scaling is applied.
xrange	an optional vector of two values the range of x shown.
main	title
xlab	label of x-axis
ylim	an optional vector of two values specifying the limits of y

plot.est

Plot estimates

Description

Plot estimates

Usage

```
## S3 method for class 'est'
plot(x, ..., lim = NA)
```


Arguments

x	input
...	arguments passed on to the function
lim	limits of the scale. By default, NA. To set limits manually, provide a vector of max and min

plot.hyperframe	<i>Plot estimates</i>
-----------------	-----------------------

Description

Plot estimates

Usage

```
## S3 method for class 'hyperframe'
plot(
  x,
  ...,
  col,
  time_col = "time",
  range,
  lim = NA,
  scalename = NA,
  color = c("white", "#F8DAC5FF", "#F4825AFF", "#D2204CFF", "#771F59FF"),
  combined = TRUE
)
```

Arguments

x	input
...	arguments passed on to the function
col	the name/s of a column of interest. To specify multiple columns, users should list column names as a character vector.
time_col	The name of the column of time variable. By default, "time". Note that the time variable must be integers.
range	vector that specifies the range of time variable (e.g., 'c("2007-01-01", "2007-01-31")')
lim	limits of the scale. By default, NA. To set limits manually, provide a vector or max and min
scalename	the name of the scale (for images only)
color	the color scale. By default, "white", "#F8DAC5FF", "#F4825AFF", "#D2204CFF", and "#771F59FF".
combined	logical. 'combined' specifies whether to combine all the point processes to one plot. This argument applies only to the case when users specify one column with multiple time periods. By default = TRUE

plot.im	<i>Plot im</i>
---------	----------------

Description

Plot im

Usage

```
## S3 method for class 'im'
plot(
  x,
  ...,
  main = "Image object",
  scalename = "Density",
  grayscale = "FALSE",
  color = c("white", "#F8DAC5FF", "#F4825AFF", "#D2204CFF", "#771F59FF"),
  lim = NA
)
```

Arguments

x	input
...	arguments passed on to the function
main	title
scalename	the name of the scale (for images only)
grayscale	whether to use grayscale. By default, FALSE.
color	the color scale. By default, "white", "#F8DAC5FF", "#F4825AFF", "#D2204CFF", and "#771F59FF".
lim	limits of the scale. By default, NA. To set limits manually, provide a vector or max and min

plot.obs	<i>Plot observed densities</i>
----------	--------------------------------

Description

Plot observed densities

Usage

```
## S3 method for class 'obs'
plot(x, ..., dens_2 = NA, dens_3 = NA, actual_data = NA, time_unit = NA)
```

Arguments

x	input
...	arguments passed on to the function
dens_2	density 2 (if any). By default, 'NA'.
dens_3	density 3 (if any). By default, 'NA'.
actual_data	actual data in the form of 'hyperframe\$column'.
time_unit	x-axis label of the output

plot.weights	<i>Plot weights</i>
--------------	---------------------

Description

Plot weights

Usage

```
## S3 method for class 'weights'
plot(x, ..., type_weights = "standardized", binwidth = NULL)
```

Arguments

x	input
...	arguments passed on to the function
type_weights	the type of weights to plot. - If 'plot_weights' is 'standardized', histogram of standardized weights will be generated. - If 'plot_weights' is 'unstandardized', histogram of unstandardized weights will be generated. Default is 'standardized'.
binwidth	bin width of the histogram. Default is NULL

predict_obs_dens	<i>Perform out-of-sample prediction</i>
------------------	---

Description

'predict_obs_dens()' performs out-of-sample prediction (separating data into training and test sets). It assumes that training and test sets have the same window.

Usage

```
predict_obs_dens(hfr, ratio, dep_var, indep_var, ngrid = 100, window)
```

Arguments

hfr	hyperframe
ratio	numeric. ratio between training and test sets
dep_var	dependent variables
indep_var	independent variables
ngrid	the number of grids. By default, '100'.
window	owin object

Value

list of the following: * 'indep_var': independent variables * 'coef': coefficients * 'intens_grid_cells': im object of observed densities for each time period * 'estimated_counts': the number of events that is estimated by the poisson point process model for each time period * 'sum_log_intens': the sum of log intensities for each time period * 'training_row_max': the max row ID of the training set

sim_cf_dens	<i>Simulate counterfactual densities</i>
-------------	--

Description

'sim_cf_dens()' takes a list of power densities and returns simulated counterfactual densities.

Usage

```
sim_cf_dens(
  expected_number,
  base_dens,
  power_sim_results,
  window,
  grayscale = FALSE,
  color = c("white", "#F8DAC5FF", "#F4825AFF", "#D2204CFF", "#771F59FF")
)
```

Arguments

expected_number	the expected number of observations
base_dens	the baseline density (im object)
power_sim_results	the results obtained by 'simulate_power_density()'
window	owin object
grayscale	logical. 'grayscale' specifies whether to convert plot to grayscale (by default, FALSE).
color	the color scale. By default, "white", "#F8DAC5FF", "#F4825AFF", "#D2204CFF", and "#771F59FF".

Value

list of counterfactual densities, a ggplot, and priorities

sim_power_dens	<i>Simulate power densities</i>
----------------	---------------------------------

Description

A function that takes the target densities and their priorities and returns a power density image over a range of parameters

Usage

```
sim_power_dens(
  target_dens,
  dens_manip,
  priorities,
  priorities_manip,
  window,
  color = c("white", "#F8DAC5FF", "#F4825AFF", "#D2204CFF", "#771F59FF"),
  grayscale = FALSE
)
```

Arguments

target_dens	list of target densities. This should always be a list, even if there is only one target density.
dens_manip	a target density for which we manipulate the value of priorities
priorities	numeric. ‘priorities’ specifies the priority for the target density that we do not manipulate.
priorities_manip	vector of priorities for the density that we manipulate.
window	owin object
color	the color scale. By default, "white", "#F8DAC5FF", "#F4825AFF", "#D2204CFF", and "#771F59FF".
grayscale	logical. ‘grayscale’ specifies whether to convert plot to grayscale (by default, FALSE).

Value

list of densities, plot, and priorities

 smooth_ppp

Smooth outcome events

Description

'smooth_ppp()' takes a column of hyperframes (ppp objects) and smoothes them.

Usage

```
smooth_ppp(data, method, sampling = NA)
```

Arguments

data	the name of a hyperframe and column of interest. 'data' should be in the form of "hyperframe\$column".
method	methods for smoothing ppp objects. Either "mclust" or "abramson". See details.
sampling	numeric between 0 and 1. 'sampling' determines the proportion of data to use for initialization. By default, NA (meaning that it uses all data without sampling).

Details

To smooth ppp objects, users can choose either the Gaussian mixture model ('method = "mclust"') or Abramson's adaptive smoothing ('method = "abramson"'). The Gaussian mixture model is essentially the method that performs model-based clustering of all the observed points. In this package, we employ the EII model (equal volume, round shape (spherical covariance)). This means that we model observed points by several Gaussian densities with the same, round shape. This is why this model is called fixed-bandwidth smoothing. This is a simple model to smooth observed points, yet given that analyzing spatiotemporal data is often computationally demanding, it is often the best place to start (and end). Sometimes this process can also take time, which is why an option for 'init' is included in this function.

Another, more precise, method for smoothing outcomes is adaptive smoothing ('method = "abram"'). This method allows users to vary bandwidths based on 'Abramson (1982)'. Essentially, this model assumes that the bandwidth is inversely proportional to the square root of the target densities. Since the bandwidth is adaptive, the estimation is usually more precise than the Gaussian mixture model. However, the caveat is that this method is often extremely computationally demanding.

Value

im objects

summary.cate	<i>Summarize results</i>
--------------	--------------------------

Description

‘summary’ functions take the output and summarize it.

Usage

```
## S3 method for class 'cate'
summary(object, ..., significance_level = 0.05)
```

Arguments

object	an output object
...	arguments passed on to the function
significance_level	Numeric scalar between 0 and 1, inclusive, representing the significance level for the chi-square test. The test is used to determine whether at least one of the coefficients (except the intercept) is not equal to 0. Default is 0.05

Details

Currently, observed densities (class: obs), estimates (class: est) and heterogeneity estimates (class: cate) are supported by this function.

summary.est	<i>Summarize results</i>
-------------	--------------------------

Description

‘summary’ functions take the output and summarize it.

Usage

```
## S3 method for class 'est'
summary(object, ...)
```

Arguments

object	an output object
...	arguments passed on to the function

Details

Currently, observed densities (class: obs) and estimates (class: est) are supported by this function.

`summary.obs`*Summarize results*

Description

'summary' functions take the output and summarize it.

Usage

```
## S3 method for class 'obs'  
summary(object, ...)
```

Arguments

<code>object</code>	an output object
<code>...</code>	arguments passed on to the function

Details

Currently, observed densities (class: obs) and estimates (class: est) are supported by this function.

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