

Package ‘disaggregation’

August 22, 2025

Type Package

Title Disaggregation Modelling

Version 0.4.1

Description Fits disaggregation regression models using 'TMB' ('Template Model Builder'). When the response data are aggregated to polygon level but the predictor variables are at a higher resolution, these models can be useful. Regression models with spatial random fields. The package is described in detail in Nandi et al. (2023) <[doi:10.18637/jss.v106.i11](https://doi.org/10.18637/jss.v106.i11)>.

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Encoding UTF-8

RoxygenNote 7.3.2

Imports splancs, Matrix, stats, TMB, dplyr, ggplot2, cowplot, rSPDE, sparseMVN, fmesher, tidyterra, terra, sf, utils

Suggests testthat, knitr, rmarkdown, SpatialEpi

LinkingTo TMB, RcppEigen

SystemRequirements GNU make

VignetteBuilder knitr

NeedsCompilation yes

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Repository CRAN

Date/Publication 2025-08-22 10:30:02 UTC

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as.disag_data	<i>Function to fit the disaggregation model</i>
---------------	---

Description

Function to fit the disaggregation model

Usage

```
as.disag_data(
  polygon_shapefile,
  shapefile_names,
  covariate_rasters,
  polygon_data,
  covariate_data,
  aggregation_pixels,
  coords_for_fit,
  coords_for_prediction,
  start_end_index,
  mesh = NULL,
  coordsForFit = NULL,
  coordsForPrediction = NULL,
  startendindex = NULL
)
```

Arguments

polygon_shapefile	sf object containing the response data
shapefile_names	List of 2: polygon id variable name and response variable name from x
covariate_rasters	SpatRaster of covariates
polygon_data	data.frame with two columns: polygon id and response
covariate_data	data.frame with cell id, polygon id and covariate columns
aggregation_pixels	vector with value of aggregation raster at each pixel
coords_for_fit	coordinates of the covariate data points within the polygons in x
coords_for_prediction	coordinates of the covariate data points in the whole raster extent
start_end_index	matrix containing the start and end index for each polygon
mesh	inla.mesh object to use in the fit
coordsForFit	Deprecated.
coordsForPrediction	Deprecated.
startendindex	Deprecated.

Value

A list is returned of class `disag_data`. The functions `summary`, `print` and `plot` can be used on `disag_data`. The list of class `disag_data` contains:

x	The sf object used as an input.
covariate_rasters	The SpatRaster used as an input.
polygon_data	A data frame with columns of <i>area_id</i> , <i>response</i> and <i>N</i> (sample size: all NAs unless using binomial data). Each row represents a polygon.
covariate_data	A data frame with columns of <i>area_id</i> , <i>cell_id</i> and one for each covariate in <i>covariate_rasters</i> . Each row represents a pixel in a polygon.
aggregation_pixels	An array with the value of the aggregation raster for each pixel in the same order as the rows of <i>covariate_data</i> .
coords_for_fit	A matrix with two columns of x, y coordinates of pixels within the polygons. Used to make the spatial field.
coords_for_prediction	A matrix with two columns of x, y coordinates of pixels in the whole Raster. Used to make predictions.
start_end_index	A matrix with two columns containing the start and end index of the pixels within each polygon.
mesh	A INLA mesh to be used for the spatial field of the disaggregation model.

build_mesh	<i>Build mesh for disaggregaton model</i>
------------	---

Description

build_mesh function takes a sf object and mesh arguments to build an appropriate mesh for the spatial field.

Usage

```
build_mesh(shapes, mesh_args = NULL, mesh.args = NULL)
```

Arguments

shapes	sf covering the region under investigation.
mesh_args	list of parameters that control the mesh structure. <i>convex</i> , <i>concave</i> and <i>resolution</i> , to control the boundary of the inner mesh, and <i>max.edge</i> , <i>cutoff</i> and <i>offset</i> , to control the mesh itself, with the parameters having the same meaning as in the INLA functions <i>inla.convex.hull</i> and <i>inla.mesh.2d</i> . <i>cut</i> has been deprecated - use <i>cutoff</i> instead.
mesh.args	Deprecated.

Details

The mesh is created by finding a tight boundary around the polygon data, and creating a fine mesh within the boundary and a coarser mesh outside. This speeds up computation time by only having a very fine mesh within the area of interest and having a small region outside with a coarser mesh to avoid edge effects.

Six mesh parameters can be specified as arguments: *convex*, *concave* and *resolution*, to control the boundary of the inner mesh, and *max.edge*, *cutoff* and *offset*, to control the mesh itself, with the names meaning the same as used by the fmesher functions *fm_nonconvex_hull_inla* and *fm_mesh_2d*.

Defaults are: `pars <- list(convex = -0.01, concave = -0.5, resolution = 300, max.edge = c(3.0, 8), cutoff = 0.4, offset = c(1, 15))`.

Value

An inla.mesh object

Examples

```
## Not run:
polygons <- list()
for(i in 1:14) {
  row <- ceiling(i/10)
  col <- ifelse(i %% 10 != 0, i %% 10, 10)
  xmin = 2*(col - 1); xmax = 2*col; ymin = 2*(row - 1); ymax = 2*row
}
```

```

polygons[[i]] <- list(cbind(c(xmin, xmax, xmax, xmin, xmin),
                           c(ymax, ymax, ymin, ymin, ymax)))
}

polys <- lapply(polygons, sf::st_polygon)
response_df <- data.frame(area_id = 1:100,
                          response = runif(100, min = 0, max = 10))
spdf <- sf::st_sf(polys, response_df)

my_mesh <- build_mesh(spdf)

## End(Not run)

```

disag_model

Fit the disaggregation model

Description

disag_model function takes a *disag_data* object created by [prepare_data](#) and performs a Bayesian disaggregation fit.

Usage

```

disag_model(
  data,
  priors = NULL,
  family = "gaussian",
  link = "identity",
  iterations = 100,
  field = TRUE,
  iid = TRUE,
  hess_control_parscale = NULL,
  hess_control_ndeps = 1e-04,
  silent = TRUE
)

```

Arguments

data	disag_data object returned by prepare_data function that contains all the necessary objects for the model fitting
priors	list of prior values: <ul style="list-style-type: none"> • priormean_intercept • priorsd_intercept • priormean_slope • priorsd_slope • prior_rho_min

	<ul style="list-style-type: none"> • prior_rho_prob • prior_sigma_max • prior_sigma_prob • prior_iideffect_sd_max • prior_iideffect_sd_prob
family	likelihood function: <i>gaussian</i> , <i>binomial</i> or <i>poisson</i>
link	link function: <i>logit</i> , <i>log</i> or <i>identity</i>
iterations	number of iterations to run the optimisation for
field	logical. Flag the spatial field on or off
iid	logical. Flag the iid effect on or off
hess_control_parscale	Argument to scale parameters during the calculation of the Hessian. Must be the same length as the number of parameters. See optimHess for details.
hess_control_ndeps	Argument to control step sizes during the calculation of the Hessian. Either length 1 (same step size applied to all parameters) or the same length as the number of parameters. Default is 1e-3, try setting a smaller value if you get NaNs in the standard error of the parameters. See optimHess for details.
silent	logical. Suppress verbose output.

Details

The model definition

The disaggregation model makes predictions at the pixel level:

$$\text{link}(\text{pred}_i) = \beta_0 + \beta X + GP(s_i) + u_i$$

And then aggregates these predictions to the polygon level using the weighted sum (via the aggregation raster, agg_i):

$$\text{cases}_j = \sum_{i \in j} \text{pred}_i \times \text{agg}_i$$

$$\text{rate}_j = \frac{\sum_{i \in j} \text{pred}_i \times \text{agg}_i}{\sum_{i \in j} \text{agg}_i}$$

The different likelihood correspond to slightly different models (y_j is the response count data):

- Gaussian: If σ is the dispersion of the pixel data, σ_j is the dispersion of the polygon data, where $\sigma_j = \sigma \sqrt{\sum \text{agg}_i^2 / \sum \text{agg}_i}$

$$\text{dnorm}(y_j / \sum \text{agg}_i, \text{rate}_j, \sigma_j)$$

- predicts incidence rate.

- Binomial: For a survey in polygon j , y_j is the number positive and N_j is the number tested.

$$\text{dbinom}(y_j, N_j, \text{rate}_j)$$

- predicts prevalence rate.

- Poisson:

$$dpois(y_j, cases_j)$$

- predicts incidence count.

Specify priors for the regression parameters, field and iid effect as a single list. Hyperpriors for the field are given as penalised complexity priors you specify ρ_{min} and ρ_{prob} for the range of the field where $P(\rho < \rho_{min}) = \rho_{prob}$, and σ_{min} and σ_{prob} for the variation of the field where $P(\sigma > \sigma_{min}) = \sigma_{prob}$. Also, specify pc priors for the iid effect

The *family* and *link* arguments are used to specify the likelihood and link function respectively. The likelihood function can be one of *gaussian*, *poisson* or *binomial*. The link function can be one of *logit*, *log* or *identity*. These are specified as strings.

The field and iid effect can be turned on or off via the *field* and *iid* logical flags. Both are default TRUE.

The *iterations* argument specifies the maximum number of iterations the model can run for to find an optimal point.

The *silent* argument can be used to publish/suppress verbose output. Default TRUE.

Value

A list is returned of class `disag_model`. The functions *summary*, *print* and *plot* can be used on `disag_model`. The list of class `disag_model` contains:

<code>obj</code>	The TMB model object returned by <code>MakeADFun</code> .
<code>opt</code>	The optimized model object returned by <code>nlminb</code> .
<code>sd_out</code>	The TMB object returned by <code>sdreport</code> .
<code>data</code>	The <code>disag_data</code> object used as an input to the model.
<code>model_setup</code>	A list of information on the model setup. Likelihood function (<i>family</i>), link function (<i>link</i>), logical: whether a field was used (<i>field</i>) and logical: whether an iid effect was used (<i>iid</i>).

References

Nanda et al. (2023) disaggregation: An R Package for Bayesian Spatial Disaggregation Modeling. <doi:10.18637/jss.v106.i11>

Examples

```
## Not run:
polygons <- list()
n_polygon_per_side <- 10
n_polygons <- n_polygon_per_side * n_polygon_per_side
n_pixels_per_side <- n_polygon_per_side * 2

for(i in 1:n_polygons) {
  row <- ceiling(i/n_polygon_per_side)
  col <- ifelse(i %% n_polygon_per_side != 0, i %% n_polygon_per_side, n_polygon_per_side)
  xmin = 2*(col - 1); xmax = 2*col; ymin = 2*(row - 1); ymax = 2*row
  polygons[[i]] <- list(cbind(c(xmin, xmax, xmax, xmin, xmin),
```

```

        c(ymax, ymax, ymin, ymin, ymax)))
    }

    polys <- lapply(polygons,sf::st_polygon)
    N <- floor(runif(n_polygons, min = 1, max = 100))
    response_df <- data.frame(area_id = 1:n_polygons, response = runif(n_polygons, min = 0, max = 1000))

    spdf <- sf::st_sf(response_df, geometry = polys)

    # Create raster stack
    r <- terra::rast(ncol=n_pixels_per_side, nrow=n_pixels_per_side)
    terra::ext(r) <- terra::ext(spdf)
    r[] <- sapply(1:terra::ncell(r), function(x){
    rnorm(1, ifelse(x %% n_pixels_per_side != 0, x %% n_pixels_per_side, n_pixels_per_side), 3)})
    r2 <- terra::rast(ncol=n_pixels_per_side, nrow=n_pixels_per_side)
    terra::ext(r2) <- terra::ext(spdf)
    r2[] <- sapply(1:terra::ncell(r), function(x) rnorm(1, ceiling(x/n_pixels_per_side), 3))
    cov_stack <- c(r, r2)
    names(cov_stack) <- c('layer1', 'layer2')

    test_data <- prepare_data(polygon_shapefile = spdf,
                             covariate_rasters = cov_stack)

    result <- disag_model(test_data, iterations = 2)

    ## End(Not run)

```

dummy

Roxygen commands

Description

Roxygen commands

Usage

dummy()

getCovariateRasters

Get a SpatRaster of covariates from a folder containing .tif files

Description

Looks in a specified folder for raster files. Returns a multi-layered SpatRaster of the rasters cropped to the extent specified by the shape parameter.

Usage

```
getCovariateRasters(directory, file_pattern = ".tif$", shape)
```

Arguments

directory Filepath to the directory containing the rasters.
file_pattern Pattern the filenames must match. Default is all files ending in .tif .
shape An object with an extent that the rasters will be cropped to.

Value

A multi-layered SpatRaster of the raster files in the directory

Examples

```
## Not run:
  getCovariateRasters('/home/rasters', '.tif$', shape)

## End(Not run)
```

<code>getPolygonData</code>	<i>Extract polygon id and response data into a data.frame from a sf object</i>
-----------------------------	--

Description

Returns a data.frame with a row for each polygon in the sf object and columns: area_id, response and N, containing the id of the polygon, the values of the response for that polygon, and the sample size respectively. If the data is not survey data (the sample size does not exist), this column will contain NAs.

Usage

```
getPolygonData(
  shape,
  id_var = "area_id",
  response_var = "response",
  sample_size_var = NULL
)
```

Arguments

shape A sf object containing response data.
id_var Name of column in shape object with the polygon id. Default 'area_id'.
response_var Name of column in shape object with the response data. Default 'response'.
sample_size_var For survey data, name of column in sf object (if it exists) with the sample size data. Default NULL.

Value

A data.frame with a row for each polygon in the sf object and columns: area_id, response and N, containing the id of the polygon, the values of the response for that polygon, and the sample size respectively. If the data is not survey data (the sample size does not exist), this column will contain NAs.

Examples

```
{
polygons <- list()
for(i in 1:100) {
  row <- ceiling(i/10)
  col <- ifelse(i %% 10 != 0, i %% 10, 10)
  xmin = 2*(col - 1); xmax = 2*col; ymin = 2*(row - 1); ymax = 2*row
  polygons[[i]] <- list(cbind(c(xmin, xmax, xmax, xmin, xmin),
                             c(ymax, ymax, ymin, ymin, ymax)))
}

polys <- lapply(polygons,sf::st_polygon)
response_df <- data.frame(area_id = 1:100, response = runif(100, min = 0, max = 10))
spdf <- sf::st_sf(response_df, geometry = polys)

  getPolygonData(spdf, id_var = 'area_id', response_var = 'response')
}
```

getStartendindex

Function to match pixels to their corresponding polygon

Description

From the covariate data and polygon data, the function matches the polygon id between the two to find which pixels from the covariate data are contained in each of the polygons.

Usage

```
getStartendindex(covariates, polygon_data, id_var = "area_id")
```

Arguments

covariates	data.frame with each covariate as a column and an id column.
polygon_data	data.frame with polygon id and response data.
id_var	string with the name of the column in the covariate data.frame containing the polygon id.

Details

Takes a data.frame containing the covariate data with a polygon id column and one column for each covariate, and another data.frame containing polygon data with a polygon id, response and sample size column (as returned by getPolygonData function).

Returns a matrix with two columns and one row for each polygon. The first column is the index of the first row in covariate data that corresponds to that polygon, the second column is the index of the last row in covariate data that corresponds to that polygon.

Value

A matrix with two columns and one row for each polygon. The first column is the index of the first row in covariate data that corresponds to that polygon, the second column is the index of the last row in covariate data that corresponds to that polygon.

Examples

```
{
  covs <- data.frame(area_id = c(1, 1, 1, 2, 2, 3, 3, 3, 3), response = c(3, 9, 5, 2, 3, 6, 7, 3, 5))
  response <- data.frame(area_id = c(1, 2, 3), response = c(4, 7, 2), N = c(NA, NA, NA))
  getStartendindex(covs, response, 'area_id')
}
```

make_model_object

Create the TMB model object for the disaggregation model

Description

make_model_object function takes a *disag_data* object created by [prepare_data](#) and creates a TMB model object to be used in fitting.

Usage

```
make_model_object(
  data,
  priors = NULL,
  family = "gaussian",
  link = "identity",
  field = TRUE,
  iid = TRUE,
  silent = TRUE
)
```

Arguments

data	disag_data object returned by <code>prepare_data</code> function that contains all the necessary objects for the model fitting
priors	list of prior values
family	likelihood function: <i>gaussian</i> , <i>binomial</i> or <i>poisson</i>
link	link function: <i>logit</i> , <i>log</i> or <i>identity</i>
field	logical. Flag the spatial field on or off
iid	logical. Flag the iid effect on or off
silent	logical. Suppress verbose output.

Details**The model definition**

The disaggregation model make predictions at the pixel level:

$$\text{link}(\text{pred}_i) = \beta_0 + \beta X + GP(s_i) + u_i$$

And then aggregates these predictions to the polygon level using the weighted sum (via the aggregation raster, agg_i):

$$\text{cases}_j = \sum_{i \in j} \text{pred}_i \times \text{agg}_i$$

$$\text{rate}_j = \frac{\sum_{i \in j} \text{pred}_i \times \text{agg}_i}{\sum_{i \in j} \text{agg}_i}$$

The different likelihood correspond to slightly different models (y_j is the response count data):

- Gaussian: If σ is the dispersion of the pixel data, σ_j is the dispersion of the polygon data, where $\sigma_j = \sigma \sqrt{\sum \text{agg}_i^2 / \sum \text{agg}_i}$

$$\text{dnorm}(y_j / \sum \text{agg}_i, \text{rate}_j, \sigma_j)$$

- predicts incidence rate.

- Binomial: For a survey in polygon j , y_j is the number positive and N_j is the number tested.

$$\text{dbinom}(y_j, N_j, \text{rate}_j)$$

- predicts prevalence rate.

- Poisson:

$$\text{dpois}(y_j, \text{cases}_j)$$

- predicts incidence count.

Specify priors for the regression parameters, field and iid effect as a single named list. Hyperpriors for the field are given as penalised complexity priors you specify ρ_{min} and ρ_{prob} for the range of the field where $P(\rho < \rho_{min}) = \rho_{prob}$, and σ_{min} and σ_{prob} for the variation of the field where $P(\sigma > \sigma_{min}) = \sigma_{prob}$. Also, specify pc priors for the iid effect.

The precise names and default values for these priors are:

- `priormean_intercept`: 0
- `priorsd_intercept`: 10.0
- `priormean_slope`: 0.0
- `priorsd_slope`: 0.5
- `prior_rho_min`: A third the length of the diagonal of the bounding box.
- `prior_rho_prob`: 0.1
- `prior_sigma_max`: `sd(response/mean(response))`
- `prior_sigma_prob`: 0.1
- `prior_iideffect_sd_max`: 0.1
- `prior_iideffect_sd_prob`: 0.01

The *family* and *link* arguments are used to specify the likelihood and link function respectively. The likelihood function can be one of *gaussian*, *poisson* or *binomial*. The link function can be one of *logit*, *log* or *identity*. These are specified as strings.

The *field* and *iid* effect can be turned on or off via the *field* and *iid* logical flags. Both are default TRUE.

The *iterations* argument specifies the maximum number of iterations the model can run for to find an optimal point.

The *silent* argument can be used to publish/supress verbose output. Default TRUE.

Value

The TMB model object returned by [MakeADFun](#).

Examples

```
## Not run:
polygons <- list()
n_polygon_per_side <- 10
n_polygons <- n_polygon_per_side * n_polygon_per_side
n_pixels_per_side <- n_polygon_per_side * 2

for(i in 1:n_polygons) {
  row <- ceiling(i/n_polygon_per_side)
  col <- ifelse(i %% n_polygon_per_side != 0, i %% n_polygon_per_side, n_polygon_per_side)
  xmin = 2*(col - 1); xmax = 2*col; ymin = 2*(row - 1); ymax = 2*row
  polygons[[i]] <- list(cbind(c(xmin, xmax, xmax, xmin, xmin),
                             c(ymax, ymax, ymin, ymin, ymax)))
}

polys <- lapply(polygons,sf::st_polygon)
N <- floor(runif(n_polygons, min = 1, max = 100))
response_df <- data.frame(area_id = 1:n_polygons, response = runif(n_polygons, min = 0, max = 1000))

spdf <- sf::st_sf(response_df, geometry = polys)

# Create raster stack
```

```

r <- terra::rast(ncol=n_pixels_per_side, nrow=n_pixels_per_side)
terra::ext(r) <- terra::ext(spdf)
r[] <- sapply(1:terra::ncell(r), function(x){
  rnorm(1, ifelse(x %% n_pixels_per_side != 0, x %% n_pixels_per_side, n_pixels_per_side), 3)})
r2 <- terra::rast(ncol=n_pixels_per_side, nrow=n_pixels_per_side)
terra::ext(r2) <- terra::ext(spdf)
r2[] <- sapply(1:terra::ncell(r), function(x) rnorm(1, ceiling(x/n_pixels_per_side), 3))
cov_stack <- c(r, r2)
names(cov_stack) <- c('layer1', 'layer2')

test_data <- prepare_data(polygon_shapefile = spdf,
                          covariate_rasters = cov_stack)

result <- make_model_object(test_data)

## End(Not run)

```

plot.disag_data

Plot input data for disaggregation

Description

Plotting function for class *disag_data* (the input data for disaggregation).

Usage

```

## S3 method for class 'disag_data'
plot(x, which = c(1, 2, 3), ...)

```

Arguments

x	Object of class <i>disag_data</i> to be plotted.
which	If a subset of plots is required, specify a subset of the numbers 1:3
...	Further arguments to <i>plot</i> function.

Details

Produces three plots: polygon response data, covariate rasters and INLA mesh.

Value

A list of three plots: the polygon plot (ggplot), covariate plot (splot) and INLA mesh plot (ggplot)

plot.disag_model	<i>Plot results of fitted model</i>
------------------	-------------------------------------

Description

Plotting function for class *disag_model* (the result of the disaggregation fitting).

Usage

```
## S3 method for class 'disag_model'  
plot(x, include_iid = FALSE, ...)
```

Arguments

x	Object of class <i>disag_model</i> to be plotted.
include_iid	logical. Whether or not to include predictions that include the IID effect.
...	Further arguments to <i>plot</i> function.

Details

Produces two plots: results of the fixed effects and in-sample observed vs predicted plot.

Value

A list of two ggplot plots: results of the fixed effects and an in-sample observed vs predicted plot

plot.disag_prediction	<i>Plot mean and uncertainty predictions from the disaggregation model results</i>
-----------------------	--

Description

Plotting function for class *disag_prediction* (the mean and uncertainty predictions of the disaggregation fitting).

Usage

```
## S3 method for class 'disag_prediction'  
plot(x, ...)
```

Arguments

x	Object of class <i>disag_prediction</i> to be plotted.
...	Further arguments to <i>plot</i> function.

Details

Produces raster plots of the mean prediction, and the lower and upper confidence intervals.

Value

A list of plots of rasters from the prediction: mean prediction, lower CI and upper CI.

`plot_disag_model_data` *Convert results of the model ready for plotting*

Description

Convert results of the model ready for plotting

Usage

```
plot_disag_model_data(x)
```

Arguments

`x` Object of class *disag_model* to be plotted.

Value

A list that contains:

<code>posteriors</code>	A data.frame of posteriors
<code>data</code>	A data.frame of observed and predicted data
<code>title</code>	The title of the observed vs. predicted plot

`predict.disag_model` *Predict mean and uncertainty from the disaggregation model result*

Description

predict.disag_model function takes a *disag_model* object created by *disaggregation::disag_model* and predicts mean and uncertainty maps.

Usage

```
## S3 method for class 'disag_model'
predict(
  object,
  new_data = NULL,
  predict_iid = FALSE,
  N = 100,
  CI = 0.95,
  newdata = NULL,
  ...
)
```

Arguments

object	disag_model object returned by disag_model function.
new_data	If NULL, predictions are made using the data in model_output. If this is a raster stack or brick, predictions will be made over this data.
predict_iid	logical. If TRUE, any polygon iid effect from the model will be used in the prediction. Default FALSE.
N	Number of realisations. Default: 100.
CI	Credible interval to be calculated from the realisations. Default: 0.95.
newdata	Deprecated.
...	Further arguments passed to or from other methods.

Details

To predict over a different spatial extent to that used in the model, a `SpatRaster` covering the region to make predictions over is passed to the argument `new_data`. If this is not given predictions are made over the data used in the fit.

The `predict_iid` logical flag should be set to TRUE if the results of the iid effect from the model are to be used in the prediction.

For the uncertainty calculations, the number of the realisations and the size of the credible interval to be calculated are given by the arguments `N` and `CI` respectively.

Value

An object of class `disag_prediction` which consists of a list of two objects:

mean_prediction

List of:

- `prediction` Raster of mean predictions based.
- `field` Raster of the field component of the linear predictor.
- `iid` Raster of the iid component of the linear predictor.
- `covariates` Raster of the covariate component of the linear predictor.

uncertainty_prediction:

List of:

- *realisations* SpatRaster of realisations of predictions. Number of realisations defined by argument *N*.
- *predictions_ci* SpatRaster of the upper and lower credible intervals. Defined by argument *CI*.

Examples

```
## Not run:
predict(fit_result)

## End(Not run)
```

predict_model	<i>Function to predict mean from the model result</i>
---------------	---

Description

predict_model function takes a *disag_model* object created by *disaggregation::disag_model* and predicts mean maps.

Usage

```
predict_model(
  model_output,
  new_data = NULL,
  predict_iid = FALSE,
  newdata = NULL
)
```

Arguments

model_output	disag_model object returned by disag_model function
new_data	If NULL, predictions are made using the data in model_output. If this is a SpatRaster, predictions will be made over this data. Default NULL.
predict_iid	If TRUE, any polygon iid effect from the model will be used in the prediction. Default FALSE.
newdata	Deprecated.

Details

Function returns rasters of the mean predictions as well as the covariate and field contributions to the linear predictor.

To predict over a different spatial extent to that used in the model, a SpatRaster covering the region to make predictions over is passed to the argument *new_data*. If this is not given predictions are made over the data used in the fit.

The *predict_iid* logical flag should be set to TRUE if the results of the iid effect from the model are to be used in the prediction.

Value

The mean prediction, which is a list of:

- *prediction* Raster of mean predictions based.
- *field* Raster of the field component of the linear predictor.
- *iid* Raster of the iid component of the linear predictor.
- *covariates* Raster of the covariate component of the linear predictor.

Examples

```
## Not run:
predict_model(result)

## End(Not run)
```

predict_uncertainty *Function to predict uncertainty from the model result*

Description

predict_uncertainty function takes a *disag_model* object created by *disaggregation::disag_model* and predicts upper and lower credible interval maps.

Usage

```
predict_uncertainty(
  model_output,
  new_data = NULL,
  predict_iid = FALSE,
  N = 100,
  CI = 0.95,
  newdata = NULL
)
```

Arguments

<code>model_output</code>	<i>disag_model</i> object returned by <i>disag_model</i> function.
<code>new_data</code>	If NULL, predictions are made using the data in <code>model_output</code> . If this is a raster stack or brick, predictions will be made over this data. Default NULL.
<code>predict_iid</code>	If TRUE, any polygon iid effect from the model will be used in the prediction. Default FALSE.
<code>N</code>	number of realisations. Default: 100.
<code>CI</code>	credible interval. Default: 0.95.
<code>newdata</code>	Deprecated.

Details

Function returns a `SpatRaster` of the realisations as well as the upper and lower credible interval rasters.

To predict over a different spatial extent to that used in the model, a `SpatRaster` covering the region to make predictions over is passed to the argument `new_data`. If this is not given predictions are made over the data used in the fit.

The `predict_iid` logical flag should be set to `TRUE` if the results of the iid effect from the model are to be used in the prediction.

The number of the realisations and the size of the credible interval to be calculated. are given by the arguments `N` and `CI` respectively.

Value

The uncertainty prediction, which is a list of:

- `realisations` `SpatRaster` of realisations of predictions. Number of realisations defined by argument `N`.
- `predictions_ci` `SpatRaster` of the upper and lower credible intervals. Defined by argument `CI`.

Examples

```
## Not run:
predict_uncertainty(result)

## End(Not run)
```

```
prepare_data
```

```
Prepare data for disaggregation modelling
```

Description

`prepare_data` function is used to extract all the data required for fitting a disaggregation model. Designed to be used in the `disaggregation::disag_model` function.

Usage

```
prepare_data(
  polygon_shapefile,
  covariate_rasters,
  aggregation_raster = NULL,
  id_var = "area_id",
  response_var = "response",
  sample_size_var = NULL,
  mesh_args = NULL,
  na_action = FALSE,
```

```

    make_mesh = TRUE,
    mesh.args = NULL,
    na.action = NULL,
    makeMesh = NULL,
    ncores = NULL
  )

```

Arguments

polygon_shapefile	sf object containing at least three columns: one with the geometried, one with the id for the polygons (<i>id_var</i>) and one with the response count data (<i>response_var</i>); for binomial data, i.e survey data, it can also contain a sample size column (<i>sample_size_var</i>).
covariate_rasters	SpatRaster of covariate rasters to be used in the model.
aggregation_raster	SpatRaster to aggregate pixel level predictions to polygon level e.g. population to aggregate prevalence. If this is not supplied a uniform raster will be used.
id_var	Name of column in sf object with the polygon id.
response_var	Name of column in sf object with the response data.
sample_size_var	For survey data, name of column in sf object (if it exists) with the sample size data.
mesh_args	list of parameters that control the mesh structure with the same names as used by INLA.
na_action	logical. If TRUE, NAs in response will be removed, covariate NAs will be given the median value, aggregation NAs will be set to zero. Default FALSE (NAs in response or covariate data within the polygons will give errors).
make_mesh	logical. If TRUE, build INLA mesh, takes some time. Default TRUE.
mesh.args	Deprecated.
na.action	Deprecated.
makeMesh	Deprecated.
ncores	Deprecated.

Details

Takes a sf object with the response data and a SpatRaster of covariates.

Extract the values of the covariates (as well as the aggregation raster, if given) at each pixel within the polygons (*parallelExtract* function). This is done in parallel and *n.cores* argument is used to set the number of cores to use for covariate extraction. This can be the number of covariates used in the model.

The aggregation raster defines how the pixels within each polygon are aggregated. The disaggregation model performs a weighted sum of the pixel prediction, weighted by the pixel values in the

aggregation raster. For disease incidence rate you use the population raster to aggregate pixel incidence rate by summing the number of cases (rate weighted by population). If no aggregation raster is provided a uniform distribution is assumed, i.e. the pixel predictions are aggregated to polygon level by summing the pixel values.

Makes a matrix that contains the start and end pixel index for each polygon. Builds an INLA mesh to use for the spatial field (*getStartendindex* function).

The *mesh.args* argument allows you to supply a list of INLA mesh parameters to control the mesh used for the spatial field (*build_mesh* function).

The *na.action* flag is automatically off. If there are any NAs in the response or covariate data within the polygons the *prepare_data* method will error. Ideally the NAs in the data would be dealt with beforehand, however, setting *na.action = TRUE* will automatically deal with NAs. It removes any polygons that have NAs as a response, sets any aggregation pixels with NA to zero and sets covariate NAs pixels to the median value for the that covariate.

Value

A list is returned of class *disag_data*. The functions *summary*, *print* and *plot* can be used on *disag_data*. The list of class *disag_data* contains:

<code>x</code>	The sf object used as an input.
<code>covariate_rasters</code>	The SpatRaster used as an input.
<code>polygon_data</code>	A data frame with columns of <i>area_id</i> , <i>response</i> and <i>N</i> (sample size: all NAs unless using binomial data). Each row represents a polygon.
<code>covariate_data</code>	A data frame with columns of <i>area_id</i> , <i>cell_id</i> and one for each covariate in <i>covariate_rasters</i> . Each row represents a pixel in a polygon.
<code>aggregation_pixels</code>	An array with the value of the aggregation raster for each pixel in the same order as the rows of <i>covariate_data</i> .
<code>coords_for_fit</code>	A matrix with two columns of x, y coordinates of pixels within the polygons. Used to make the spatial field.
<code>coords_for_prediction</code>	A matrix with two columns of x, y coordinates of pixels in the whole Raster. Used to make predictions.
<code>start_end_index</code>	A matrix with two columns containing the start and end index of the pixels within each polygon.
<code>mesh</code>	A INLA mesh to be used for the spatial field of the disaggregation model.

Examples

```
polygons <- list()
for(i in 1:100) {
  row <- ceiling(i/10)
  col <- ifelse(i %% 10 != 0, i %% 10, 10)
  xmin = 2*(col - 1); xmax = 2*col; ymin = 2*(row - 1); ymax = 2*row
  polygons[[i]] <- list(cbind(c(xmin, xmax, xmax, xmin, xmin),
```

```

    c(ymax, ymax, ymin, ymin, ymax))
  }

  polys <- lapply(polygons,sf::st_polygon)
  response_df <- data.frame(area_id = 1:100, response = runif(100, min = 0, max = 10))
  spdf <- sf::st_sf(response_df,geometry=polys)

  r <- terra::rast(nrow=20,ncol=20)
  terra::ext(r) <- terra::ext(spdf)
  r[] <- sapply(1:terra::ncell(r), function(x) rnorm(1, ifelse(x %% 20 != 0, x %% 20, 20), 3))

  r2 <- terra::rast(nrow=20,ncol=20)
  terra::ext(r2) <- terra::ext(spdf)
  r2[] <- sapply(1:terra::ncell(r), function(x) rnorm(1, ceiling(x/10), 3))
  cov_rasters <- c(r, r2)

  test_data <- prepare_data(polygon_shapefile = spdf,
                           covariate_rasters = cov_rasters)

```

print.disag_data

Print function for disaggregation input data

Description

Function that prints the input data from the disaggregation model.

Usage

```

## S3 method for class 'disag_data'
print(x, ...)

```

Arguments

x	Object returned from prepare_data.
...	Further arguments to <i>print</i> function.

Details

Prints the number of polyons and pixels, the number of pixels in the largest and smallest polygons and summaries of the covariates.

print.disag_model *Print function for disaggregation fit result.*

Description

Function that prints the result of the fit from the disaggregation model.

Usage

```
## S3 method for class 'disag_model'  
print(x, ...)
```

Arguments

x Object returned from `disag_model`.
... Further arguments to *print* function.

Details

Prints the negative log likelihood, model parameters and calculates metrics from in-sample performance.

print.disag_prediction *Print function for disaggregation prediction*

Description

Function that prints the prediction from the disaggregation model.

Usage

```
## S3 method for class 'disag_prediction'  
print(x, ...)
```

Arguments

x Object returned from `predict.disag_model`.
... Further arguments to *print* function.

Details

Prints the number of polygons and pixels, the number of pixels in the largest and smallest polygons and summaries of the covariates.

summary.disag_data *Summary function for disaggregation input data*

Description

Function that summarizes the input data from the disaggregation model.

Usage

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

Arguments

object Object returned from prepare_data.
... Further arguments to *summary* function.

Details

Prints the number of polyons and pixels, the number of pixels in the largest and smallest polygons and summaries of the covariates.

Value

A list of the number of polyons, the number of covariates and summaries of the covariates.

summary.disag_model *Summary function for disaggregation fit result*

Description

Function that summarises the result of the fit from the disaggregation model.

Usage

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

Arguments

object Object returned from disag_model.
... Further arguments to *summary* function.

Details

Prints the negative log likelihood, model parameters and calculates metrics from in-sample performance.

Value

A list of the model parameters, negative log likelihood and metrics from in-sample performance.

```
summary.disag_prediction
```

Summary function for disaggregation prediction

Description

Function that summarizes the prediction from the disaggregation model.

Usage

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

Arguments

object	Object returned from predict.disag_model
...	Further arguments to <i>summary</i> function.

Details

Prints the number of polygons and pixels, the number of pixels in the largest and smallest polygons and summaries of the covariates.

Value

A list of the number of polygons, the number of covariates and summaries of the covariates.

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