Package 'trafficCAR'

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Type Package

Title Bayesian CAR Models for Road-Segment Traffic

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Description Tools for simulating and modeling traffic flow on road networks using spatial conditional autoregressive (CAR) models. The package represents road systems as graphs derived from 'OpenStreetMap' data <https://www.openstreetmap.org/> and supports network-based spatial dependence, basic preprocessing, and visualization for spatial traffic analysis.

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Encoding UTF-8

LazyData true

LazyDataCompression xz

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Depends R (>= 3.5)

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Contents

`augment_roads` *Augment roads with predicted traffic quantities*

Description

Adds predicted traffic outcomes (e.g. speed or volume) and relative congestion measures to a road network.

Usage

```
augment_roads(fit, roads, probs = c(0.025, 0.975), keep_geometry = TRUE)
```

Arguments

`fit`	a ‘traffic_fit‘ object from ‘fit_traffic()‘.
`roads`	an sf object or data.frame with a segment id column.
`probs`	length-2 numeric for equal-tail intervals.
`keep_geometry`	logical; if FALSE drops sf geometry.

Value

roads with added columns: predicted_mean, predicted_lo, predicted_hi, relative_congestion

---

`build_adjacency` *Build segment adjacency from segment geometries*

Description

Constructs an undirected adjacency matrix ‘A‘ where segments are neighbors if they share a node (endpoint). Intended to be used after ‘roads_to_segments()‘.

Usage

```
build_adjacency(segments, crs_m = 3857, tol = 0, verbose = FALSE)
```

Arguments

`segments`	An ‘sf‘ with LINESTRING geometries and (optionally) ‘seg_id‘.
`crs_m`	Metric CRS used when ‘segments‘ is lon/lat (for robust node keys). Default 3857.
`tol`	Nonnegative numeric tolerance for snapping node coordinates (in meters if projected). If 0, uses exact coordinates. Default 0.
`verbose`	Logical; emit simple messages. Default FALSE.

Details

Isolates (degree 0) are kept (all-zero rows/cols). Connected components are returned for ICAR sum-to-zero centering per component.

Value

A list with:

A Sparse symmetric adjacency matrix (‘dgCMatrix‘).

components Integer vector component id (length n). Isolates are their own components.

isolates Logical vector (length n).

`build_network` *Build a road network graph from sf LINESTRING data*

Description

Build a road network graph from sf LINESTRING data

Usage

```
build_network(
  roads_sf,
  crs_out = 3857,
  node_intersections = FALSE,
  snap_tol = 0,
  simplify = TRUE
)
```

Arguments

`roads_sf`	An sf object with LINESTRING geometry
`crs_out`	Integer EPSG code for projected CRS
`node_intersections`	Logical; if TRUE, "node" the linework by splitting at interior intersections/junctions (via ‘sf::st_union()‘), so that crossings and T-junctions become graph nodes even when they are not endpoints. This may increase the number of edge segments.
`snap_tol`	Nonnegative numeric; optional snapping tolerance (in projected CRS units) used to merge nearly identical endpoints. Use 0 to disable.
`simplify`	Logical; if TRUE, remove self-loops and parallel edges.

Value

A list with components:

- roads: cleaned sf object
- nodes: sf POINT object with node_id
- edges: sf LINESTRING object with from, to, length
- graph: igraph object
- A: sparse adjacency matrix

`car_precision` *CAR precision matrix from an adjacency matrix*

Description

Constructs the precision matrix for an intrinsic CAR (ICAR) or proper CAR model:

$$Q = \tau(D - \rho A), \quad D = \mathrm{diag}(A\mathbf{1}).$$

Usage

```
car_precision(
  A,
  type = c("icar", "proper"),
  rho = 0.99,
  tau = 1,
  symmetrize = FALSE,
  check = TRUE
)
```

Arguments

`A`	Square adjacency/weight matrix (base matrix or a ‘Matrix‘ sparse type). Diagonal entries are ignored (set to 0).
`type`	Either ‘"icar"‘ or ‘"proper"‘.
`rho`	Spatial dependence parameter for proper CAR. Ignored for ICAR.
`tau`	Positive scalar precision multiplier.
`symmetrize`	If ‘TRUE‘, replaces ‘A‘ by ‘(A + t(A))/2‘ before construction.
`check`	If ‘TRUE‘, performs basic validation and warnings.

Details

For ICAR, set ‘type = "icar"‘ (internally uses $\rho = 1$). For proper CAR, set ‘type = "proper"‘ and choose ‘rho‘ so that $D - \rho A$ is positive definite (no automatic spectral checks are performed).

Value

A symmetric sparse precision matrix ‘Q‘ (class ‘"dsCMatrix"‘).

Examples

```
A <- matrix(0, 4, 4)
A[1,2] <- A[2,1] <- 1
A[2,3] <- A[3,2] <- 1
A[3,4] <- A[4,3] <- 1
Q_icar <- car_precision(A, type = "icar", tau = 1)
Q_prop <- car_precision(A, type = "proper", rho = 0.9, tau = 2)
```

`fetch_osm_roads` *Fetch road geometries from OpenStreetMap*

Description

Convenience wrapper around ‘osmdata‘ to download road geometries and return an ‘sf‘ object that can be passed into ‘roads_to_segments()‘.

Usage

```
fetch_osm_roads(
  place,
  key = "highway",
  value = NULL,
  extra_tags = NULL,
  layer = c("osm_lines", "osm_multilines", "osm_polygons", "osm_multipolygons"),
  quiet = TRUE,
  ...
)
```

Arguments

`place`	A character place name (passed to ‘osmdata::getbb()‘) or a bounding box object accepted by ‘osmdata::opq()‘.
`key`	OSM feature key to query. Default is ‘"highway"‘.
`value`	Optional character vector of OSM feature values. For example, ‘c("primary", "secondary")‘.
`extra_tags`	Optional named list of additional tags passed to ‘osmdata::add_osm_feature()‘.
`layer`	Which OSM layer to return. Defaults to ‘"osm_lines"‘.
`quiet`	Logical; suppress osmdata messages. Default TRUE.
`...`	Additional arguments passed to ‘osmdata::osmdata_sf()‘.

Value

An ‘sf‘ object with road geometries.

`fit_car` *Fit Gaussian CAR / ICAR regression via Gibbs sampling*

Description

Fits a Gaussian regression with a CAR/ICAR latent effect: $y = X\beta + x + \epsilon$ with $\epsilon \sim N(0, \sigma^2 I)$ and $x \sim N(0, Q^{-1})$, where $Q = \tau(D - \rho A)$ and $D = diag(A1)$. For ICAR, $\rho = 1$.

Usage

```
fit_car(
  y,
  A,
  X = NULL,
  type = c("icar", "proper"),
  rho = 0.99,
  tau = 1,
  n_iter = 2000,
  burn_in = floor(n_iter/2),
  thin = 1,
  beta_init = NULL,
  x_init = NULL,
  sigma2_init = NULL,
  b0 = NULL,
  B0 = NULL,
  a0 = 2,
  b0_sigma = 1,
  center_icar = TRUE,
  verbose = FALSE
)
```

Arguments

`y`	Numeric response vector of length `n`.
`A`	Square `n` x `n` adjacency/weight matrix (base matrix or `Matrix`). Diagonal entries are ignored.
`X`	Optional `n` x `p` design matrix. If `NULL`, no regression is fit.
`type`	Either `"icar"` or `"proper"`.
`rho`	Spatial dependence parameter for proper CAR. Ignored for ICAR.
`tau`	Positive scalar precision multiplier.
`n_iter`	Total MCMC iterations.
`burn_in`	Number of initial iterations to discard.
`thin`	Keep every `thin`-th draw after burn-in.
`beta_init`	Optional initial β (length `p`).

x_init	Optional initial latent field x (length n).
sigma2_init	Optional initial σ^2 (positive scalar).
b0	Prior mean for β (length p). Default is zero vector.
B0	Prior covariance for β (p x p). Default is large diagonal.
a0	Shape parameter for inverse-gamma prior on σ^2.
b0_sigma	Scale parameter for inverse-gamma prior on σ^2.
center_icar	Logical; if TRUE and type="icar", center x to sum-to-zero within each connected component.
verbose	Logical; print coarse progress updates.

Details

The sampler updates x, β (if X is provided), and σ^2 using Gibbs steps.

Value

A list of class "trafficCAR_fit" with elements:

draws List with MCMC draws x, beta, sigma2.

keep Iteration indices that were saved.

type, rho, tau Model hyperparameters used.

y = X beta + x + epsilon

---

fit_traffic *Fit a Gaussian CAR traffic model (speed or travel time)*

Description

This is a thin wrapper around fit_car() that: 1) preprocesses the outcome (log/per-distance options), 2) fits the Gaussian CAR model, 3) returns a traffic-flavored object that can be augmented back onto roads.

Usage

```
fit_traffic(
  data,
  roads = NULL,
  A = NULL,
  segment_id_col = "segment_id",
  outcome = c("speed", "travel_time"),
  outcome_col = NULL,
  distance_col = NULL,
  per_distance = FALSE,
  transform = c("log", "identity"),
  X = NULL,
  ...
)
```

Arguments

data	data.frame with at least ‘segment_id‘ and the outcome column.
roads	optional; an sf object or similar that contains adjacency info already used by your ‘fit_car()‘ pipeline (depends on your package design).
A	adjacency matrix or object accepted by ‘fit_car()‘ (recommended explicit).
segment_id_col	character; column in ‘data‘ used to join back to roads.
outcome	character; one of "speed" or "travel_time".
outcome_col	optional character; if NULL uses ‘outcome‘.
distance_col	optional character; used only for travel_time when per_distance=TRUE.
per_distance	logical; only for travel_time.
transform	character; "log" or "identity".
X	optional design matrix; if NULL, uses intercept-only.
...	passed to ‘fit_car()‘ (e.g., type="proper"/"icar", rho, priors, n_iter, burn, etc.)

Value

An object of class ‘traffic_fit‘ containing the underlying fit and transform metadata.

---

icar_sum_to_zero *Apply sum-to-zero constraint to ICAR precision*

Description

Projects onto the subspace orthogonal to the constant vector.

Usage

```
icar_sum_to_zero(Q)
```

Arguments

Q	ICAR precision matrix

Value

Constrained precision matrix

`intrinsic_car_precision` *Intrinsic CAR (ICAR) precision matrix*

Description

Constructs the intrinsic CAR precision matrix

$$Q = \tau\, s(D - A),$$

where s is a scaling constant chosen so that the geometric mean of the marginal variances equals 1.

Usage

```
intrinsic_car_precision(
  A,
  tau = 1,
  scale = TRUE,
  symmetrize = FALSE,
  check = TRUE
)
```

Arguments

`A`	Square adjacency/weight matrix.
`tau`	Positive scalar precision multiplier.
`scale`	Logical; if ‘TRUE‘, applies Besag scaling.
`symmetrize`	If ‘TRUE‘, replaces ‘A‘ by ‘(A + t(A))/2‘.
`check`	If ‘TRUE‘, performs basic validation and warnings.

Details

The resulting precision matrix is singular with rank deficiency equal to the number of connected components.

Value

A symmetric sparse precision matrix (‘"dsCMatrix"‘).

References

Sørbye, S. H. and Rue, H. (2014). Scaling intrinsic Gaussian Markov random field priors.

map_roads_interactive *Interactive map of road-segment traffic measures*

Description

Displays standard traffic quantities such as predicted speed, predicted volume, or relative congestion on an interactive map.

Usage

```
map_roads_interactive(
  sf_aug,
  value = c("predicted_speed", "predicted_volume", "relative_congestion"),
  engine = "leaflet"
)
```

Arguments

sf_aug	An ‘sf‘ object returned by ‘augment_roads()‘.
value	Character scalar. One of: ‘"predicted_speed"‘, ‘"predicted_volume"‘, ‘"relative_congestion"‘.
engine	Currently only ‘"leaflet"‘ is supported.

Value

A leaflet widget.

map_roads_interactive_layers *Interactive map with multiple standard traffic layers*

Description

Interactive map with multiple standard traffic layers

Usage

```
map_roads_interactive_layers(
  sf_aug,
  values = c("predicted_speed", "relative_congestion")
)
```

Arguments

sf_aug	sf object with road geometries
values	Character vector of traffic measures to include.

Value

leaflet widget

`moran_residuals` *Moran's I for trafficCAR residuals*

Description

Computes Moran's I statistic for model residuals using the model adjacency.

Usage

```
moran_residuals(
  fit,
  type = c("raw", "structured", "unstructured"),
  nsim = 199,
  method = c("analytic", "permutation")
)
```

Arguments

`fit`	A 'traffic_fit' object.
`type`	Residual type: "raw" or "unstructured".
`nsim`	Number of permutations for permutation test.
`method`	"analytic" or "permutation".

Value

An object of class 'traffic_moran'.

`plot_mcmc_diagnostics` *MCMC diagnostic plots*

Description

MCMC diagnostic plots

Usage

```
plot_mcmc_diagnostics(fit)
```

Arguments

`fit`	traffic_fit

Value

A list with components:

plot A 'ggplot' object of diagnostic summaries.

summary A data frame with columns 'parameter' and 'ess', giving the effective sample size for each parameter..

---

plot_observed_fitted *Plot observed vs predicted traffic values*

Description

Plot observed vs predicted traffic values

Usage

```
plot_observed_fitted(fit, data)
```

Arguments

fit	traffic_fit
data	data.frame

Value

ggplot

---

plot_predicted *Plot predicted traffic outcome on road network*

Description

Plot predicted traffic outcome on road network

Usage

```
plot_predicted(fit, roads)
```

Arguments

fit	traffic_fit
roads	sf with segment_id

Value

ggplot

plot_relative_congestion *Plot relative congestion on road network*

Description

Shows systematic deviations after accounting for covariates.

Usage

```
plot_relative_congestion(fit, roads)
```

Arguments

fit	traffic_fit
roads	sf

Value

ggplot

plot_roads_static *Static map of road-segment traffic measures*

Description

Static map of road-segment traffic measures

Usage

```
plot_roads_static(
  sf_aug,
  value = c("predicted_speed", "predicted_volume", "relative_congestion")
)
```

Arguments

sf_aug	sf object with road geometries
value	One of "predicted_speed", "predicted_volume", or "relative_congestion"

Value

ggplot object

plot_traffic_map *Quick map helper for augmented roads*

Description

Plots road geometries colored by an augmented numeric column (e.g., posterior mean predictions or relative congestion).

Usage

```
plot_traffic_map(roads_aug, fill = c("predicted_mean", "relative_congestion"))
```

Arguments

roads_aug	An ‘sf‘ object returned by [augment_roads()].
fill	Character scalar. Which column of ‘roads_aug‘ to map. One of ‘"predicted_mean"‘ or ‘"relative_congestion"‘.

Value

An invisible copy of ‘roads_aug‘, returned as an ‘sf‘ object with the augmented columns. The function is called for its plotting side effect.

ppc_summary *Posterior predictive checks for trafficCAR fits*

Description

Computes simple posterior predictive checks comparing observed statistics to replicated data: mean, variance, and tail probabilities.

Usage

```
ppc_summary(fit, stats = c("mean", "var", "tail"), probs = c(0.05, 0.95))
```

Arguments

fit	A ‘traffic_fit‘ object.
stats	Statistics to compute: "mean", "var", "tail".
probs	Tail probabilities for "tail" statistic.

Value

An object of class ‘traffic_ppc‘.

residuals.traffic_fit *Residuals for trafficCAR fits*

Description

Computes raw, structured (spatial), or unstructured residuals from a fitted trafficCAR model.

Usage

```
## S3 method for class 'traffic_fit'
residuals(object, type = c("raw", "structured", "unstructured"), ...)
```

Arguments

object	A ‘traffic_fit‘ object.
type	Residual type: "raw", "structured", or "unstructured".
...	Unused.

Value

Numeric vector of residuals.

roads_datasets *Example road network datasets*

Description

A collection of example road network datasets provided as ‘sf‘ objects. These datasets are intended for demonstration, testing, and benchmarking.

Format

An object of class sf.

Details

Each dataset contains LINESTRING road geometries suitable for use with roads_to_segments() and related functions.

Included datasets:

roads_small Small example road network.

roads_cstat Road network for College Station, TX.

`roads_to_segments` *Convert road geometries to modeling segments*

Description

Takes an ‘sf‘ object of LINESTRING/MULTILINESTRING road geometries and returns a segment-level ‘sf‘ with stable segment IDs and metric lengths.

Usage

```
roads_to_segments(
  roads,
  crs_m = 3857,
  keep_attrs = NULL,
  drop_zero = TRUE,
  split_at_intersections = FALSE,
  verbose = FALSE
)
```

Arguments

`roads`	An ‘sf‘ object with LINESTRING or MULTILINESTRING geometries.
`crs_m`	Metric CRS used for length calculation (and intersection splitting) when ‘roads‘ is lon/lat. Default 3857. For best accuracy, pass a local UTM EPSG.
`keep_attrs`	Optional character vector of non-geometry columns to keep. If ‘NULL‘, keeps all attributes.
`drop_zero`	Logical; drop segments with non-positive length. Default TRUE.
`split_at_intersections`	Logical; if TRUE, split lines at all intersections. Implemented via GEOS noding (‘sf::st_union‘ + ‘sf::st_cast‘) Default FALSE.
`verbose`	Logical; emit simple messages about dropped rows. Default FALSE.

Details

v1 behavior: * Drops Z/M dimensions * Casts MULTILINESTRING -> LINESTRING (one row per linestring) * Optionally splits at intersections (noding) when ‘split_at_intersections=TRUE‘ * Computes ‘length_m‘ in meters (projects if lon/lat) * Drops empty and (optionally) zero-length segments

Value

An ‘sf‘ with columns: * ‘seg_id‘ integer 1..n * ‘length_m‘ numeric meters * geometry LINESTRING plus kept attributes.

`sample_proper_car` *Gibbs sampler for a proper CAR latent Gaussian model*

Description

Model:

$$y \mid x, \tau \sim N(x, \tau^{-1} I)$$
$$x \mid \kappa \sim N(0, (\kappa Q)^{-1}), \quad Q = D - \rho A \text{ (proper CAR)}$$
$$\tau \sim \mathrm{Gamma}(a_\tau, b_\tau) \quad \text{(shape-rate)}$$
$$\kappa \sim \mathrm{Gamma}(a_\kappa, b_\kappa) \quad \text{(shape-rate)}$$

Usage

```
sample_proper_car(
  y,
  A,
  rho = 0.99,
  n_iter,
  burn = 0L,
  thin = 1L,
  a_tau = 1,
  b_tau = 1,
  a_kappa = 1,
  b_kappa = 1,
  init = NULL,
  symmetrize = FALSE,
  check = TRUE
)
```

Arguments

`y`	Numeric vector of observations (length n).
`A`	Adjacency matrix (dense or sparse). Diagonal ignored.
`rho`	Proper CAR dependence parameter (must satisfy car_precision checks).
`n_iter`	Integer number of iterations.
`burn`	Integer burn-in iterations to drop (default 0).
`thin`	Integer thinning interval (default 1).
`a_tau, b_tau`	Gamma(shape, rate) prior for tau.
`a_kappa, b_kappa`	Gamma(shape, rate) prior for kappa.
`init`	Optional list with elements x, tau, kappa.
`symmetrize`	Passed to car_precision().
`check`	Passed to car_precision().

Value

List with x (matrix), tau, kappa, and settings.

`simplify_network` *Simplify a built network object by removing parallel edges (and loops)*

Description

Simplify a built network object by removing parallel edges (and loops)

Usage

```
simplify_network(net, keep_edge = c("first", "shortest"))
```

Arguments

`net`	A network list returned by [build_network()].
`keep_edge`	Which edge to keep when multiple edges connect the same unordered node pair. One of "first" or "shortest".

Value

A network list with updated ‘edges‘, ‘graph‘, and ‘A‘ (and the same ‘nodes‘).

`weights_from_adjacency` *Construct spatial weights matrix*

Description

Construct spatial weights matrix

Usage

```
weights_from_adjacency(
  A,
  style = c("binary", "row-standardized"),
  symmetrize = FALSE
)
```

Arguments

`A`	Adjacency matrix.
`style`	One of "binary" or "row-standardized".
`symmetrize`	Passed to adjacency coercion.

Value

Sparse weight matrix.

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