

# Package ‘ppgam’

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**Type** Package

**Title** Generalised Additive Point Process Models

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**Description** Methods for fitting point processes with parameters of generalised additive model (GAM) form are provided. For an introduction to point processes see Cox, D.R & Isham, V. (Point Processes, 1980, CRC Press), GAMs see Wood, S.N. (2017) <doi:10.1201/9781315370279>, and the fitting approach see Wood, S.N., Pya, N. & Safken, B. (2016) <doi:10.1080/01621459.2016.1180986>.

**License** GPL-3

**Depends** R (>= 3.5.0)

**Imports** MASS, mgcv, evgam

**RoxygenNote** 7.1.0

**NeedsCompilation** no

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ppgam

*Fit a generalised additive point process model***Description**

Fit a generalised additive point process model

**Usage**

```
ppgam(
  formula,
  data,
  nodes = NULL,
  weights = 1,
  nquad = 15,
  approx = c("midpoint", "exact"),
  knots = NULL,
  use.data = TRUE,
  trace = 0
)
```

**Arguments**

formula	a formula for a Poisson process log intensity function (compatible with <a href="#">gam</a> )
data	a data frame
nodes	a list or data frame; see ‘Details’
weights	a scalar, list or vector; see ‘Details’
nquad	a scalar giving the number of quadrature nodes for each variable
approx	a length 2 character string; see ‘Details’
knots	spline knots to pass to <a href="#">gam</a>
use.data	should splines should be constructed from data (otherwise uses nodes)?
trace	integers controlling what’s reported. Defaults to 0

**Details**

ppgam fits a Poisson process with intensity function  $\lambda(\mathbf{x})$  for covariate  $\mathbf{x} = (x_1, \dots, x_d)$ . The likelihood for this model with events occurring at  $\mathbf{x}_i$ , for  $i = 1, \dots, n$ , is approximated by quadrature with

$$\exp \left[ - \sum_{j=1}^m w_j \lambda(\mathbf{x}_j^*) \right] \prod_{i=1}^n \lambda(\mathbf{x}_i)$$

where  $\mathbf{x}_j^*$  and  $w_j$  are quadrature nodes and weights, for  $j = 1, \dots, m$ , defined with nodes and weights.

`formula` gives the formula for the log intensity function of a Poisson process. It is passed to `gam`. If `formula` has no response, i.e.  $\sim s(\dots)$ , then `data` is assumed to give the times at which events occur. Then `nodes` is used to control integration of the intensity function. If `formula` has a response, e.g.  $y \sim s(\dots)$ , then `y` is assumed binary, comprising only zeros and ones. Then `data` is assumed to give the state space of the Poisson process, (e.g. daily time steps if occurrences of events are measured in days) and ones in `y` identify when events occur. Note that if `formula` has no response, `data` will have  $n$  rows, and  $m$  rows otherwise.

`nodes` is used to supply nodes for integrating the Poisson process intensity function by quadrature. It is supplied as a list or data frame.

If `nodes` is a list, its names must correspond to variables on the r.h.s. of `formula`. Elements of the list, `x`, say, can be a vector or 2-column matrix, where  $\text{length}(x) > 1$  or  $\text{nrow}(x) > 1$ . If a matrix, its first and second columns are taken as integration nodes and weights, respectively. If a vector of length 2, it is assumed to give the range of the `nquad` midpoints used as integration nodes. If a longer vector, it is assumed to be the integration nodes, and `nquad` is ignored.

If `nodes` is a data frame, it is assumed to give the integration nodes.

`nquad` specifies the number of integration nodes per variable, unless `nodes` are specified in `nodes`. If a single integer and `is.null(names(nquad))` it is used for all variables. Otherwise, names are matched to variables. An error is returned if any variables do not have values specified.

`weights` controls the quadrature weights. If `nodes` is a list, a scalar multiplies any weights calculated alongside `nodes`, i.e. node separations. If `nodes` is a data frame, `weights` can be a scalar that is repeated  $\text{nrow}(\text{nodes})$ , or a vector of length  $\text{nrow}(\text{nodes})$  that gives the weights for each row of `nodes`.

`approx` controls quadrature details. Its first term controls the integration method, which uses either midpoint ("midpoint", default), Simpson's ("Simpson") or Gauss-Legendre ("Gauss") rules. The second term of `approx` controls the integration range, which is either the range of the variable ("exact"), or by calling `pretty()` ("pretty").

`trace` controls what is reported. Details of convergence are printed with `trace = 1`, of `nodes` with `trace = 2`, and `trace = 3` prints both.

## Value

An object of class `gam`, as returned by `mgcv::gam`, with parameters, covariance matrices and a few other things swapped

## References

- Wood, S. N., Pya, N., & Säfken, B. (2016). Smoothing parameter and model selection for general smooth models. *Journal of the American Statistical Association*, 111(516), 1548-1563.
- Youngman, B. D., & Economou, T. (2017). Generalised additive point process models for natural hazard occurrence. *Environmetrics*, 28(4), e2444.

## Examples

```
# Times of landfalling US hurricanes
data(USlandfall)
```

```

# convert dates to years, as a continuous variable
year <- as.integer(format(USlandfall$date, "%Y"))
day <- as.integer(format(USlandfall$date, "%j"))
USlandfall$year <- year + pmin(day / 365, 1)
hits <- subset(USlandfall, landfall == 1)

# this creates nodes in the default way
m1 <- ppgam( ~ s(year), hits)

# some examples of providing nodes
nodes.year <- list(year=pretty(USlandfall$year, 20))
# as 2 is in trace, nodes and weights are printed
m2 <- ppgam( ~ s(year), hits, nodes = nodes.year, trace = 2)

# alternatively, we might just want to specify how many nodes to use
m3 <- ppgam( ~ s(year), hits, nquad = 30)

data(windstorm)
m4 <- ppgam(~ s(lon, lat, k=20), windstorm)

## Storm peak locations, given the North Atlantic Oscillation (NAO) index
# NAO values from https://crudata.uea.ac.uk/cru/data/nao/nao.dat
# NAO midpoints and weights based on `hist`

NAO.mids <- c(-2.75, -2.25, -1.75, -1.25, -0.75, -0.25, 0.25, 0.75, 1.25, 1.75, 2.25)
NAO.wts <- c(0.002, 0.014, 0.057, 0.145, 0.302, 0.427, 0.463, 0.364, 0.171, 0.047, 0.007)

m5 <- ppgam(~ te(lat, lon, NAO, d = 2:1, k = c(40, 8), bs = c("ts", "cr")), windstorm,
  nodes = list(NAO = cbind(NAO.mids, NAO.wts)))

```

---

USlandfall

*Times of landfalling US hurricanes*


---

### Description

A data frame:

### Usage

```
data(USlandfall)
```

### Format

A data frame with 61129 rows and 2 variables

The variables are as follows:

**date** date of landfall, as class "Date"

**landfall** an integer: did a hurricane make landfall on this day?

## References

<https://www.nhc.noaa.gov/data/>

## Examples

```
data(USlandfall)
plot(USlandfall, type="h")
```

---

windstorm

*Locations of windstorm peaks and tracks over the North Atlantic*

---

## Description

A dataset in windstorm peaks between 1st January 1979 and 31st December 2014 occurring in [-50, 33] longitude and [36, 77] latitude.

## Usage

```
data(windstorm)
```

## Format

A data frame with 3133 rows and 4 variables

The variables are as follows:

**date** date of peak, as class "Date"

**lon** longitude, in degrees

**lat** latitude, in degrees

**NAO** North Atlantic Oscillation index

## References

Youngman, B. D., & Economou, T. (2017). Generalised additive point process models for natural hazard occurrence. *Environmetrics*, 28(4), e2444.

## Examples

```
data(windstorm)
plot(windstorm[,c("lon", "lat")])
```

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\* **datasets**

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