

# Package ‘threshr’

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**Title** Threshold Selection and Uncertainty for Extreme Value Analysis

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**Description** Provides functions for the selection of thresholds for use in extreme value models, based mainly on the methodology in Northrop, Attalides and Jonathan (2017) <[doi:10.1111/rssc.12159](https://doi.org/10.1111/rssc.12159)>. It also performs predictive inferences about future extreme values, based either on a single threshold or on a weighted average of inferences from multiple thresholds, using the 'revdbayes' package <<https://cran.r-project.org/package=revdbayes>>. At the moment only the case where the data can be treated as independent identically distributed observations is considered.

**Imports** graphics, methods, revdbayes (>= 1.3.4), rust (>= 1.2.2), stats

**License** GPL (>= 2)

**LazyData** TRUE

**Encoding** UTF-8

**Depends** R (>= 3.3.0)

**RoxygenNote** 7.1.0

**Suggests** knitr, rmarkdown, testthat

**VignetteBuilder** knitr

**URL** <https://paulnorthrop.github.io/threshr/>,  
<https://github.com/paulnorthrop/threshr>

**BugReports** <https://github.com/paulnorthrop/threshr/issues>

**NeedsCompilation** no

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gom	<i>Storm peak significant wave heights from the Gulf of Mexico</i>
-----	--

---

### Description

A numeric vector containing 315 hindcasts of storm peak significant wave heights, in metres, from 1900 to 2005 at an unnamed location in the Gulf of Mexico.

### Usage

```
gom
```

### Format

A vector containing 315 observations.

### Source

Oceanweather Inc. (2005) GOMOS – Gulf of Mexico hindcast study.

### References

Northrop, P. J., N. Attalides, and P. Jonathan. (2017). Cross-Validatory Extreme Value Threshold Selection and Uncertainty with Application to Ocean Storm Severity. *Journal of the Royal Statistical Society: Series C (Applied Statistics)*, **66**(1), 93-120. doi:[10.1111/rssc.12159](https://doi.org/10.1111/rssc.12159).

ithresh

*Threshold selection in the i.i.d. case (peaks over threshold)***Description**

Produces a diagnostic plot to assist in the selection of an extreme value threshold in the case where the data can be treated as independent and identically distributed (i.i.d.) observations. For example, it could be that these observations are the cluster maxima resulting from the declustering of time series data. The predictive ability of models fitted using each of a user-supplied set of thresholds is assessed using leave-one-out cross-validation in a Bayesian setup. These models are based on a Generalized Pareto (GP) distribution for threshold excesses and a binomial model for the probability of threshold exceedance. See [Northrop et al. \(2017\)](#) for details.

**Usage**

```
ithresh(data, u_vec, ..., n_v = 1, npy = NULL, use_rcpp = TRUE)
```

**Arguments**

- |       |  |
|-------|--|
| data  | A numeric vector of observations. Any missing values will be removed. The argument <code>npy</code> (see below) may be supplied as an attribute of <code>data</code> using <code>attr(data, "npy") &lt;- value</code> , where <code>value</code> is the value of <code>npy</code> (see <a href="#">attr</a> ). If <code>npy</code> is supplied twice, as both <code>attr(data, "npy")</code> and using the <code>npy</code> argument, then the former is used.   |
| u_vec | A numeric vector. A vector of <i>training</i> thresholds at which inferences are made from a binomial-GP model. These could be set at sample quantiles of data using <a href="#">quantile</a> . Any duplicated values will be removed.   |
| ...   | Further (optional) arguments to be passed to the <a href="#">revdbayes</a> function <code>rpost_rcpp</code> (or <code>rpost</code> ), which use the generalized ratio-of-uniforms method to simulate from extreme value posterior distributions. In particular: <ul style="list-style-type: none"> <li>• <code>n</code> The size of the posterior sample used to perform predictive inference. Default: <code>n = 1000</code>.</li> <li>• <code>prior</code> A prior for GP parameters to be passed to the <a href="#">revdbayes</a> function <code>set_prior</code>. Can be either a character scalar that chooses an in-built prior, or a user-supplied R function or pointer to a compiled C++ function. See the <a href="#">set_prior</a> documentation for details of the in-built priors. See the <a href="#">revdbayes</a> vignette <a href="#">Faster simulation using revdbayes</a> for information about creating a pointer to a C++ function. See also the <b>Examples</b> section. If the user supplies an R function then <code>rpost</code> will be used for posterior simulation, rather than (the faster) <code>rpost_rcpp</code>, regardless of the input value of <code>use_rcpp</code>. Default: <code>prior = "mdi"</code> with <code>a = 0.6</code> and <code>min_xi = -1</code>. This particular prior is studied in <a href="#">Northrop et al. (2017)</a>.</li> <li>• <code>h_prior</code> A <i>list</i> of further arguments (hyperparameters) for the GP prior specified in <code>prior</code>.</li> </ul> |

- `bin_prior` A prior for the threshold exceedance probability  $p$  to be passed to the `revdbayes` function `set_bin_prior`. Can either be a character scalar that chooses an in-built prior, or a user-supplied R function. Default: `prior = "jeffreys"`, i.e.  $\text{Beta}(1/2, 1/2)$ .
- `h_bin_prior` A *list* of further arguments (hyperparameters) for the binomial prior specified in `bin_prior`. See the `set_bin_prior` documentation for details of the in-built priors.
- `trans` A character scalar: either "none" or "BC". See `rpost_rcpp` for details. The default is "none", which is usually faster than "BC". However, if there are very few threshold excesses then using `trans = "BC"` can make the optimizations involved in the generalized ratio-of-uniforms algorithm more stable. If using `trans = "none"` produces an error for a particular posterior simulation then `trans = "BC"` is used instead.

`n_v` A numeric scalar. Each of the `n_v` largest values in `u_vec` will be used (separately) as a *validation* threshold for the training thresholds in `u_vec` that lie at or below that validation threshold. If `n_v = 1` then all the training thresholds are used with validation performed using the threshold `u_vec[length(u_vec)]`. If `n_v = 2` then, in addition, the assessment is performed using `u_vec[1]`, `...`, `u_vec[length(u_vec) - 1]` with validation threshold `u_vec[length(u_vec) - 1]`, and so on.

`npv` A numeric scalar. The mean number of observations per year of data, after excluding any missing values, i.e. the number of non-missing observations divided by total number of years of non-missing data. May be supplied using as an attribute `attr(data, "npv")` of data instead. The value of `npv` does not affect any calculation in `ithresh`, it only affects subsequent extreme value inferences using `predict.ithresh`. However, setting `npv` in the call to `rpost`, or as an attribute of data avoids the need to supply `npv` when calling `predict.ithresh`.

`use_rcpp` A logical scalar. If TRUE (the default) the `revdbayes` function `rpost_rcpp` is used for posterior simulation. If FALSE, or if the user supplies an R function to set the prior for GP parameters, the (slower) function `rpost` is used.

## Details

For a given threshold in `u_vec`:

- the number of values in data that exceed the threshold, and the amounts (the *threshold excesses*) by which these value exceed the threshold are calculated;
- `rpost_rcpp` (or `rpost`) is used to sample from the posterior distributions of the parameters of a GP model for the threshold excesses and a binomial model for the probability of threshold exceedance;
- the ability of this binomial-GP model to predict data thresholded at the validation threshold(s) specified by `n_v` is assessed using leave-one-out cross-validation (the measure of this is given in equation (7) of Northrop et al. (2017)).

See Northrop et al. (2017) and the introductory threshr vignette for further details and examples.

**Value**

An object (list) of class "ithresh", containing the components

- `pred_perf`: A numeric matrix with `length(u_vec)` rows and `n_v` columns. Each column contains the values of the measure of predictive performance. Entries corresponding to cases where the training threshold is above the validation threshold will be NA.
- `u_vec`: The argument `u_vec` to `ithresh`.
- `v_vec`: A numeric vector. The validation thresholds implied by the argument `n_v` to `ithresh`.
- `u_ps`: A numeric vector. The approximate levels of the sample quantiles to which the values in `u_vec` correspond, i.e. the approximate percentage of the data that lie at or below each element in `u_vec`.
- `v_ps`: A numeric vector. The values in `u_ps` that correspond to the validation thresholds.
- `sim_vals`: A numeric matrix with 4 columns and `n x length(u_vec)` rows. The  $j$ th block of `n` rows contains in columns 1-3 the posterior samples of the threshold exceedance probability, the GP scale parameter and the GP shape parameter respectively, based on training threshold `u_vec[i]`, and in column 4 the value of  $j$ .
- `n`: A numeric scalar. The value of `n`.
- `npv`: A numeric scalar. The value of `npv`.
- `data`: The argument `data` to `ithresh` detailed above, with any missing values removed.
- `use_rcpp`: A logical scalar indicating whether `rpost_rcpp` (`use_rcpp = TRUE`) or `rpost` (`use_rcpp = FALSE`) was used for posterior simulation.
- `for_post`: A list containing arguments with which `rpost_rcpp` (or `rpost`) was called, including any user-supplied arguments to these functions.
- `call`: The call to `ithresh`.

**References**

- Northrop, P.J. and Attalides, N. (2016) Posterior propriety in Bayesian extreme value analyses using reference priors *Statistica Sinica*, **26**(2), 721–743 <https://doi.org/10.5705/ss.2014.034>.
- Northrop, P. J., Attalides, N. and Jonathan, P. (2017) Cross-validators extreme value threshold selection and uncertainty with application to ocean storm severity. *Journal of the Royal Statistical Society Series C: Applied Statistics*, **66**(1), 93-120. <https://doi.org/10.1111/rssc.12159>
- Jonathan, P. and Ewans, K. (2013) Statistical modelling of extreme ocean environments for marine design : a review. *Ocean Engineering*, **62**, 91-109. <https://doi.org/10.1016/j.oceaneng.2013.01.004>

**See Also**

- [plot.ithresh](#) for the S3 plot method for objects of class `ithresh`.
- [summary.ithresh](#) Summarizing measures of threshold predictive performance.
- [predict.ithresh](#) for predictive inference for the largest value observed in `N` years.
- [rpost](#) in the `revdbayes` package for details of the arguments that can be passed to `rpost_rcpp/rpost`.
- [set\\_prior](#) and [set\\_bin\\_prior](#) in the `revdbayes` package for details of how to set a prior distributions for GP parameters and for the exceedance probability  $p$ .
- [quantile](#).

## Examples

```

# Note:
# 1. Smoother plots result from making n larger than the default n = 1000.
# 2. In some examples below validation thresholds rather higher than is
#    advisable have been used, with far fewer excesses than the minimum of
#    50 suggested by Jonathan and Ewans (2013).

## North Sea significant wave heights, default prior -----
#' # A plot akin to the top left of Figure 7 in Northrop et al. (2017)
#' # ... but with fewer training thresholds

u_vec_ns <- quantile(ns, probs = seq(0.1, 0.9, by = 0.1))
ns_cv <- ithresh(data = ns, u_vec = u_vec_ns, n_v = 2)
plot(ns_cv, lwd = 2, add_legend = TRUE, legend_pos = "topright")
mtext("significant wave height / m", side = 3, line = 2.5)

## Gulf of Mexico significant wave heights, default prior -----

u_vec_gom <- quantile(gom, probs = seq(0.2, 0.9, by = 0.1))
# Setting a prior using its name and parameter value(s) -----
# This example gives the same prior as the default
gom_cv <- ithresh(data = gom, u_vec = u_vec_gom, n_v = 2, prior = "mdi",
                  h_prior = list(a = 0.6))

## Setting a user-defined (log-)prior R function -----
# This example also gives the same prior as the default
# (It will take longer to run than the example above because ithresh detects
# that the prior is an R function and sets use_rcpp to FALSE.)

user_prior <- function(pars, a, min_xi = -1) {
  if (pars[1] <= 0 | pars[2] < min_xi) {
    return(-Inf)
  }
  return(-log(pars[1]) - a * pars[2])
}
user_bin_prior <- function(p, ab) {
  return(stats::dbeta(p, shape1 = ab[1], shape2 = ab[2], log = TRUE))
}
gom_cv <- ithresh(data = gom, u_vec = u_vec_gom, n_v = 2, prior = user_prior,
                  h_prior = list(a = 0.6), bin_prior = user_bin_prior,
                  h_bin_prior = list(ab = c(1 / 2, 1 / 2)))

## Setting a user-defined (log-)prior (pointer to a) C++ function -----
# We make use of a C++ function and function create_prior_xptr() to create
# the required pointer from the revdbayes package

prior_ptr <- revdbayes::create_prior_xptr("gp_flat")
gom_cv <- ithresh(data = gom, u_vec = u_vec_gom, n_v = 2, prior = prior_ptr,
                  h_prior = list(min_xi = -1))

```

---

ns *Storm peak significant wave heights from the North Sea*

---

### Description

A numeric vector containing 628 hindcasts of storm peak significant wave heights, in metres, from 1964 to 1995 at an unnamed location in the North Sea.

### Usage

ns

### Format

A vector containing 628 observations.

### Source

Oceanweather Inc. (1995) NEXT – North Sea hindcast study.

### References

Northrop, P. J., N. Attalides, and P. Jonathan. (2017). Cross-Validatory Extreme Value Threshold Selection and Uncertainty with Application to Ocean Storm Severity. *Journal of the Royal Statistical Society: Series C (Applied Statistics)*, **66**(1), 93-120. doi:[10.1111/rssc.12159](https://doi.org/10.1111/rssc.12159).

---

plot.ithresh *Plot diagnostics an ithresh object*

---

### Description

plot method for class "ithresh". Produces an extreme value threshold diagnostic plot based on an analysis performed by `ithresh`. Can also be used to produce a plot of the posterior sample generated by `ithresh` for a particular training threshold.

### Usage

```
## S3 method for class 'ithresh'
plot(
  x,
  y,
  ...,
  which_v = NULL,
  prob = TRUE,
  top_scale = TRUE,
  add_legend = FALSE,
```

```

    legend_pos = "topleft",
    which_u = NULL
)

```

### Arguments

x	an object of class "ithresh", a result of a call to <a href="#">ithresh</a> .
y	Not used.
...	Additional arguments passed on to <a href="#">matplot</a> and/or <a href="#">legend</a> and/or <a href="#">axis</a> . If <a href="#">which_u</a> is supplied then these arguments are passed to <a href="#">plot.evpost</a> .
which_v	A numeric scalar or vector. If <a href="#">which_u</a> is not supplied (a threshold diagnostic plot is required) <a href="#">which_v</a> specifies the validation thresholds, that is, the components of <code>x\$v_vec</code> , to include in the plot. If <a href="#">which_u</a> is supplied (a plot of a posterior sample for a given threshold is required) then <a href="#">which_v</a> is a numeric scalar that indicates which element of <code>object\$v_vec</code> is used in selecting a single threshold (if <a href="#">which_u</a> = "best"). Note: the default, <a href="#">which_v</a> = 1 gives the <i>lowest</i> of the validation thresholds in <code>object\$v_vec</code> .
prob	A logical scalar. If TRUE then the levels of thresholds are represented by the proportion of observations that lie below a threshold. If <a href="#">prob</a> = FALSE then the values of the thresholds are used.
top_scale	A logical scalar indicating Whether or not to add a scale to the top horizontal axis. If this is added it gives the threshold on the scale not chosen by <a href="#">prob</a> .
add_legend	A logical scalar indicating whether or not to add a legend to the plot. If <a href="#">method</a> = "cv" then the legend gives the levels of the validation thresholds.
legend_pos	The position of the legend (if required) specified using the argument <code>x</code> in <a href="#">legend</a> .
which_u	Either a character scalar or a numeric scalar. If <a href="#">which_u</a> is supplied then <a href="#">plot.evpost</a> is used to produce a plot of the posterior sample generated using a particular training threshold. By default a scatter plot of the posterior sample of Generalized Pareto parameters is produced. If <a href="#">which_u</a> = "best" then the training threshold achieving the largest measure of predictive performance in <code>object\$pred_perf</code> , based on the validation threshold selected using <a href="#">which_v</a> , is used. See <a href="#">summary.ithresh</a> to print the best thresholds for each validation threshold. Otherwise, <a href="#">which_u</a> is a numeric scalar that selects training threshold <code>x\$u_vec[which_u]</code> . Therefore, <a href="#">which_u</a> must be an integer in 1, ..., <code>length(x\$u_vec)</code> .

### Details

Produces plots of the *threshold weights*, defined in equation (14) of [Northrop et al. \(2017\)](#), against training threshold. A line is produced for each of the validation thresholds chosen in [which\\_v](#). The result is a plot like those in the top row of Figure 7 in [Northrop et al. \(2017\)](#).

It is possible that a curve on the plot may be incomplete. This indicates that, for a particular threshold level, a measure of predictive performance is  $-\text{Inf}$ . This occurs when an observation in the data lies above the estimated upper end point of the predictive distribution produced when this observation is removed.



**Value**

If `which_u` is supplied then the object with which `plot.evpost` was called is returned (invisibly). Otherwise, a list is returned (again invisibly) with two components. `x` is a vector containing the coordinates plotted on the (lower) horizontal axis. `y` is an `length(u_vec)` by `n_v` matrix of *threshold weights* obtained by normalising the columns of the matrix `pred_perf` returned by `ithresh`. See equation (14) of [Northrop et al. \(2017\)](#).

**See Also**

[ithresh](#) for threshold selection in the i.i.d. case based on leave-one-out cross-validation.

[summary.ithresh](#) Summarizing measures of threshold predictive performance.

[print.ithresh](#) Prints the threshold weights.

[predict.ithresh](#) for predictive inference for the largest value observed in `N` years.

**Examples**

```
# [Smoother plots result from making n larger than the default n = 1000.]

# Threshold diagnostic plot
u_vec_gom <- quantile(gom, probs = seq(0, 0.9, by = 0.05))
gom_cv <- ithresh(data = gom, u_vec = u_vec_gom, n_v = 3)
plot(gom_cv, lwd = 2, add_legend = TRUE, legend_pos = "topleft")
mtext("significant wave height / m", side = 3, line = 2.5)

# Plot of Generalized Pareto posterior sample at the best threshold
# (based on the lowest validation threshold)
plot(gom_cv, which_u = "best")
# See which threshold was used
summary(gom_cv)

# Plot of Generalized Pareto posterior sample at the highest threshold
n_u <- length(u_vec_gom)
plot(gom_cv, which_u = n_u, points_par = list(pch = 20, col = "grey"))
```

---

plot.ithreshpred

*Plot diagnostics an ithreshpred object*


---

**Description**

plot method for class "ithreshpred". Produces plots to summarise the predictive inferences made by [predict.ithresh](#).

**Usage**

```
## S3 method for class 'ithreshpred'
plot(x, ..., ave_only = FALSE, add_best = FALSE)
```

**Arguments**

x	an object of class "ithreshpred", a result of a call to <code>ithresh</code> .
...	Additional arguments passed on to <code>plot.evpred</code> .
ave_only	A logical scalar. Only relevant if <code>predict.ithresh</code> was called with <code>which_u = "all"</code> . If TRUE then plot only a curve for the weighted average over multiple training thresholds. If FALSE then also plot a curve for each training threshold.
add_best	A logical scalar. If TRUE then the best threshold, as judged using the validation threshold selected using the argument <code>which_v</code> supplied to <code>predict.ithresh</code> , is highlighted by plotting it with a different line style.

**Details**

*Single threshold case*, where `predict.ithresh` was called with numeric scalar `which_u` or `which_u = "best"`. `plot.evpred` is called to produce the plot.

*Multiple threshold case*, where `predict.ithresh` was called with `which_u = "all"`. Again, `plot.evpred` is called but now the estimated predictive distribution function (`type = "p"` used in the call to `predict.ithresh`) or density function (`type = "d"`) is plotted for each of the training thresholds (grey lines) as is the result of the weighted average over the different training thresholds (black line). If graphical parameters, such as `lty`, `lwd` or `col` are passed via ... then the first element relates to the weighted average and the remaining `length(x$u_vec)` elements to the respective training thresholds in `u_vec`.

**Value**

A list containing the graphical parameters using in producing the plot including any arguments supplied via ... is returned (invisibly).

**See Also**

`ithresh` for threshold selection in the i.i.d. case based on leave-one-out cross-validation.

`predict.ithresh` for predictive inference for the largest value observed in N years.

`plot.ithresh` for the S3 plot method for objects of class `ithresh`.

`summary.ithresh` Summarizing measures of threshold predictive performance.

**Examples**

```
u_vec_gom <- quantile(gom, probs = seq(0, 0.9, by = 0.05))
gom_cv <- ithresh(data = gom, u_vec = u_vec_gom, n_v = 3)

# Note: gom_cv$npv contains the correct value of npv (it was set in the
#       call to ithresh, via attr(gom, "npv").
#       If object$npv doesn't exist then the argument npv must be supplied
#       in the call to predict().

### Best training threshold based on the lowest validation threshold

# Predictive distribution function
npv_gom <- length(gom)/105
```

```

best_p <- predict(gom_cv, n_years = c(100, 1000))
plot(best_p)

# Predictive density function
best_d <- predict(gom_cv, type = "d", n_years = c(100, 1000))
plot(best_d)

### All thresholds plus weighted average of inferences over all thresholds

# Predictive distribution function
all_p <- predict(gom_cv, which_u = "all")
plot(all_p)

# Predictive density function
all_d <- predict(gom_cv, which_u = "all", type = "d")
plot(all_d)

### ... and highlight the best threshold

plot(all_p, add_best = TRUE)
plot(all_d, add_best = TRUE)

```

---

plot.stability

*Plot diagnostics for a stability object*


---

## Description

plot method for objects of class "stability" returned from [stability](#)

## Usage

```

## S3 method for class 'stability'
plot(
  x,
  y,
  ...,
  prob = TRUE,
  top_scale = c("none", "excesses", "opposite"),
  vertical = TRUE
)

```

## Arguments

x	an object of class "stability", a result of a call to <a href="#">stability</a> .
y	Not used.
...	Additional arguments passed on to <a href="#">matplot</a> , <a href="#">axis</a> and/or <a href="#">segments</a> .
prob	A logical scalar. If TRUE then the levels of thresholds on the lower horizontal axis are represented by the proportion of observations that lie below a threshold. If prob = FALSE then the values of the thresholds are used.

top_scale	A character scalar. If top_scale = "none" then no axis labels appear on the upper horizontal axis. If top_scale = "excesses" then the number of threshold excesses at each threshold are indicated. If top_scale = "opposite" then the type of threshold level <i>not</i> chosen using prob is indicated.
vertical	A logical scalar. Should the confidence intervals be depicted using a vertical line for each threshold (TRUE) or by joining up confidence limits across thresholds (FALSE)?

### Details

Produces a simple threshold diagnostic plot based on the object returned from `stability`. The MLEs of the GP shape parameter  $\xi$  and approximate `conf%` confidence intervals for  $\xi$  are plotted against the threshold used to fit the GP model. This plot is used to choose a threshold above which the underlying GP shape parameter may be approximately constant. See Chapter 4 of Coles (2001). See also the vignette "Introducing threshr". as described in . See also the vignette "Introducing threshr".

### Value

In addition to producing the plot a list of the arguments used by `matplot`, `axis` is returned (invisibly).

### See Also

`stability`.

### Examples

```
u_vec_gom <- quantile(gom, probs = seq(0, 0.9, by = 0.05))
gom_stab <- stability(data = gom, u_vec = u_vec_gom)
plot(gom_stab)
```

---

predict.itthresh      *Predictive inference for the largest value observed in N years.*

---

### Description

`predict` method for class "itthresh". Predictive inferences can either be based on a *single training threshold* or using a weighted average of inferences over *multiple training thresholds*. A single threshold may chosen to be the best performing threshold, as judged by the measure of predictive performance calculated by `itthresh` or chosen by the user. The weights used in the latter case are based on the measures of predictive performance and prior probabilities assigned to the training thresholds. By default all thresholds are given the same prior probability but the user can specify their own prior.

**Usage**

```
## S3 method for class 'ithresh'
predict(
  object,
  npy = NULL,
  n_years = 100,
  which_u = c("best", "all"),
  which_v = 1L,
  u_prior = rep(1, length(object$u_vec)),
  type = c("p", "d", "q", "i", "r"),
  hpd = FALSE,
  x = NULL,
  ...
)
```

**Arguments**

object	An object of class "ithresh", a result of a call to <code>ithresh</code> .
npy	A numeric scalar. The mean number of observations per year of data, after excluding any missing values, i.e. the number of non-missing observations divided by total number of years of non-missing data.
n_years	A numeric vector. Value(s) of N. If <code>which_u = "all"</code> then <code>n_years</code> must have length one.
which_u	<p>Either a character scalar or a numeric scalar. If <code>which_u</code> is a character scalar it must be either "best" or "all".</p> <p>If <code>which_u = "best"</code> then the single threshold achieving the largest measure of predictive performance in <code>object\$pred_perf</code>, based on the validation threshold selected using <code>which_v</code>, is used to perform prediction. See <a href="#">summary.ithresh</a> to print the best thresholds for each validation threshold.</p> <p>If <code>which_u = "all"</code> then <i>all</i> the thresholds are used to perform prediction. The inferences from each threshold are weighted according to the <i>posterior threshold weights</i> given in equation (15) of <a href="#">Northrop et al. (2017)</a> based on the prior probabilities of thresholds in <code>u_prior</code> and column <code>which_v</code> of the measures of predictive performance in <code>object\$pred_perf</code>.</p> <p>Otherwise, <code>which_u</code> is a numeric scalar that indicates which element of <code>object\$u_vec</code> the user wishes to select as a single threshold on which to base prediction, that is, <code>which_u</code> must be an integer in <code>1, ..., length(object\$u_vec)</code>.</p>
which_v	A numeric scalar. Indicates which element of <code>object\$v_vec</code> is used in selecting a single threshold (if <code>which_u = "best"</code> ) or weighting the inferences from all thresholds (if <code>which_u = "all"</code> ). Note: the default, <code>which_v = 1</code> gives the <i>lowest</i> of the validation thresholds in <code>object\$v_vec</code> .
u_prior	<p>A numeric vector. Prior probabilities for the training thresholds in <code>object\$u_vec</code>. Only used if <code>which_u = "all"</code>.</p> <p>Only the first <code>length(object\$u_vec) - length(object\$v_vec) + which_v</code> elements of <code>u_prior</code> are used. This is because only training thresholds up to and including <code>object\$v_vec[which_v]</code> are relevant. <code>u_prior</code> must have length</p>

	length(object\$u_vec) or length(object\$u_vec) - length(object\$v_vec) + which_v.
	If u_prior is not supplied then all (relevant) training thresholds are given equal prior probability. u_prior is normalized to have sum equal to 1 inside predict.ithresh.
type	A character vector. Passed to <code>predict.evpost</code> . Indicates which type of inference is required: <ul style="list-style-type: none"> <li>• "p" for the predictive distribution function,</li> <li>• "d" for the predictive density function,</li> <li>• "q" for the predictive quantile function,</li> <li>• "i" for predictive intervals (see ... to set level),</li> <li>• "r" for random generation from the predictive distribution.</li> </ul> <p>If which_u = "all" then only type = "p" or type = "d" are allowed.</p>
hpd	A logical scalar. The argument hpd of <code>predict.evpost</code> . Only relevant if type = "i".
x	A numeric vector. The argument x of <code>predict.evpost</code> . In the current context this must be a vector (not a matrix, as suggested by the documentation of <code>predict.evpost</code> ). If x is not supplied then a default value is set within <code>predict.evpost</code> .
...	Additional arguments to be passed to <code>predict.evpost</code> . In particular: level, which can be used to set (multiple) levels for predictive intervals when type = "i"; lower_tail (relevant when type = "p" or "q") and log (relevant when type = "d").

## Details

The function `predict.evpost` is used to perform predictive based on the binomial-GP posterior sample generated using a given training threshold. For mathematical details of the single threshold and multiple threshold cases see Sections 2.3 and 3 of [Northrop et al. \(2017\)](#) respectively.

## Value

An list object of class "ithreshpred" with a similar structure to an object of class "evpred" returned from `predict.evpost` is returned *invisibly*. In addition, the object contains `u_vec = object$u_vec` and `v_vec = object$v_vec`, `which_v` and the index `best_u` in `u_vec = object$u_vec` of the best training threshold based on the value of `which_v`. It also contains the value of the Box-Cox transformation parameter `lambda`. This will always be equal to 1 if object was returned from `ithresh`.

If `which_u = "all"` then

- the list also contains the *posterior threshold weights* in component `post_thresh_wts`
- the component `y` is a matrix with `length{x}` rows and `1 + length(object$u_vec) - length(object$v_vec) + which_v` columns. Column 1 contains the estimated predictive distribution function (type = "p") or density function (type = "d") obtained using a weighted average of the inferences over different training thresholds. The other columns contain the estimated functions for each of the training thresholds in `u_vec`.

## References

Northrop, P. J., Attalides, N. and Jonathan, P. (2017) Cross-validators extreme value threshold selection and uncertainty with application to ocean storm severity. *Journal of the Royal Statistical Society Series C: Applied Statistics*, **66**(1), 93-120. <https://doi.org/10.1111/rssc.12159>

## See Also

[ithresh](#) for threshold selection in the i.i.d. case based on leave-one-out cross-validation.

[plot.ithreshpred](#) for the S3 plot method for objects of class `ithreshpred`.

## Examples

```
# Note:
#' In the examples below validation thresholds rather higher than is
#' advisable have been used, with far fewer excesses than the minimum of
#' 50 suggested by Jonathan and Ewans (2013).

# Gulf of Mexico significant wave heights, default priors.
u_vec_gom <- quantile(gom, probs = seq(0, 0.9, by = 0.05))
gom_cv <- ithresh(data = gom, u_vec = u_vec_gom, n_v = 3)

# Note: gom_cv$npv contains the correct value of npv (it was set in the
# call to ithresh, via attr(gom, "npv").
# If object$npv doesn't exist then the argument npv must be supplied
# in the call to predict().

### Best training threshold based on the lowest validation threshold

# Predictive distribution function
best_p <- predict(gom_cv, n_years = c(100, 1000))
plot(best_p)

# Predictive density function
best_d <- predict(gom_cv, type = "d", n_years = c(100, 1000))
plot(best_d)

# Predictive intervals
best_i <- predict(gom_cv, n_years = c(100, 1000), type = "i", hpd = TRUE,
                 level = c(95, 99))
plot(best_i, which_int = "both")

# See which threshold was used
summary(gom_cv)

### All thresholds plus weighted average of inferences over all thresholds

# Predictive distribution function
all_p <- predict(gom_cv, which_u = "all")
plot(all_p)

# Predictive density function
```

```
all_d <- predict(gom_cv, which_u = "all", type = "d")
plot(all_d)
```

---

print.ithresh                    *Print method for objects of class "ithresh"*

---

### Description

print method for class "ithresh".

### Usage

```
## S3 method for class 'ithresh'
print(x, digits = 2, ...)
```

### Arguments

x                    an object inheriting from class "ithresh", a result of a call to [ithresh](#).

digits                An integer. Used for number formatting with [format](#) and [signif](#).

...                    Additional optional arguments. At present no optional arguments are used.

### Details

Prints a matrix of the estimated threshold weights. Each row gives the weights for each training threshold for a given validation threshold. The row and column names are the approximate quantile levels of the thresholds.

### Value

The argument x, invisibly, as for all [print](#) methods.

### See Also

[ithresh](#) for threshold selection in the i.i.d. case based on leave-one-out cross-validation.

[summary.ithresh](#) Summarizing measures of threshold predictive performance.

[plot.ithresh](#) for the S3 plot method for objects of class `ithresh`.

[predict.ithresh](#) for predictive inference for the largest value observed in N years.



---

 stability

*Generalized Pareto parameter estimate stability*


---

### Description

Uses maximum likelihood estimation to fit a Generalized Pareto (GP) model to threshold excesses over a range of thresholds. The threshold excesses are treated as independent and identically distributed (i.i.d.) observations. The resulting estimates and confidence intervals can be plotted, using [plot.stability](#), to produce a crude graphical diagnostic for threshold choice.

### Usage

```
stability(
  data,
  u_vec,
  prof = FALSE,
  conf = 95,
  mult = 1:2,
  plot_prof = FALSE,
  ...
)
```

### Arguments

<code>data</code>	A numeric vector of observations.
<code>u_vec</code>	A numeric vector of thresholds to be applied to the data. Any duplicated values will be removed. These could be set at sample quantiles of data using <a href="#">quantile</a> .
<code>prof</code>	A logical scalar. Whether to calculate confidence intervals for the GP shape parameter $\xi$ based on the profile-likelihood for $\xi$ or using the MLE plus or minus a multiple of the estimated standard error (SE) of the MLE. The intervals produced by the former may be better but they take longer to calculate. Default: FALSE.
<code>conf</code>	A numeric scalar in (0, 100). Confidence level for the confidence intervals. Default: 95%.
<code>mult</code>	A numeric vector of length 2. The range of values over which the profile log-likelihood for $\xi$ is calculated is $(MLE - mult[1] \cdot c \cdot SE, MLE + mult[2] \cdot c \cdot SE)$ , where MLE and SE are the MLE and estimated standard error of $\xi$ and $c$ is the constant for which this interval gives an approximate $100conf\%$ level confidence interval for $\xi$ when $mult = c(1, 1)$ . The default, $mult = c(1, 2)$ , works well in most cases. If the routine fails because the range of $\xi$ is not sufficiently wide then the relevant components of <code>mult</code> should be increased.
<code>plot_prof</code>	A logical scalar. Only relevant if <code>prof = TRUE</code> . If <code>plot_prof = TRUE</code> then the profile log-likelihood is plotted for each threshold. If FALSE then nothing is plotted.
<code>...</code>	Further (optional) arguments to be passed to the <a href="#">optim</a> function for the optimizations on which the profile-likelihood for $x_i$ is based.

## Details

For each threshold in `u_vec` a GP model is fitted by maximum likelihood estimation to the threshold excesses, i.e. the amounts by which the data exceed that threshold. The MLEs of the GP shape parameter  $\xi$  and approximate `conf%` confidence intervals for  $\xi$  are stored for plotting (by `plot.stability`) to produce a simple graphical diagnostic to inform threshold selection. This plot is used to choose a threshold above which the underlying GP shape parameter may be approximately constant. See Chapter 4 of Coles (2001). See also the vignette "Introducing threshr".

## Value

An object (list) of class "stability" with components:

<code>ests</code>	MLEs of the GP shape parameter $\xi$ .
<code>ses</code>	Estimated SEs of the MLEs of $\xi$ .
<code>lower</code>	Lower limit of <code>100conf%</code> confidence intervals for $\xi$ .
<code>upper</code>	Upper limit of <code>100conf%</code> confidence intervals for $\xi$ .
<code>nexc</code>	The number of threshold excesses.
<code>u_vec</code>	The thresholds supplied by the user.
<code>u_ps</code>	The approximate sample quantiles to which the thresholds in <code>u_vec</code> correspond.
<code>data</code>	The input data.
<code>conf</code>	The input <code>conf</code> .

Each of these components is a numeric vector of length `length(u_vec)`.

## References

Coles, S. G. (2001) *An Introduction to Statistical Modeling of Extreme Values*, Springer-Verlag, London. [https://doi.org/10.1007/978-1-4471-3675-0\\_3](https://doi.org/10.1007/978-1-4471-3675-0_3)

## See Also

`ithresh` for threshold selection in the i.i.d. case based on leave-one-out cross-validation.  
`plot.stability` for the S3 plot method for objects of class `stability`.  
`quantile`.

## Examples

```
# Set a vector of thresholds
u_vec_gom <- quantile(gom, probs = seq(0, 0.9, by = 0.05))

# Symmetric confidence intervals
gom_stab <- stability(data = gom, u_vec = u_vec_gom)
plot(gom_stab)

# Profile-likelihood-based confidence intervals
gom_stab <- stability(data = gom, u_vec = u_vec_gom, prof = TRUE)
plot(gom_stab)
```

---

summary.ithresh	<i>Summarizing measures of threshold predictive performance</i>
-----------------	---

---

## Description

summary method for class "ithresh"

## Usage

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

## Arguments

object            an object of class "ithresh", a result of a call to `ithresh`.  
...                Additional optional arguments. At present no optional arguments are used.

## Value

Returns a numeric matrix with 5 columns and  $n_v$  rows, where  $n_v$  is an argument to `ithresh` that determines how many of the largest training thresholds are used as validation thresholds. The columns contain:

- column 1: the validation threshold  $v$
- column 2: the sample quantile to which the validation threshold corresponds
- column 3: the best training threshold  $u$  judged using the validation threshold  $v$
- column 4: the sample quantile to which the best training threshold corresponds
- column 5: the index of the vector `u_vec` of training thresholds to which the threshold in column 2 corresponds

## See Also

[ithresh](#) for threshold selection in the i.i.d. case based on leave-one-out cross-validation.

[plot.ithresh](#) for the S3 plot method for objects of class `ithresh`.

[print.ithresh](#) Prints the threshold weights.

## Examples

```
u_vec_gom <- quantile(gom, probs = seq(0, 0.9, by = 0.05))  
gom_cv <- ithresh(data = gom, u_vec = u_vec_gom, n_v = 3)  
summary(gom_cv)
```

---

threshr	<i>threshr: Threshold Selection and Uncertainty for Extreme Value Analysis</i>
---------	--

---

### Description

Provides functions for the selection of extreme value threshold. At the moment only the simplest case, where the data can be treated as independent identically distributed observations, is considered. Future releases will tackle more general situations. See the 'threshr' website for more information, documentation and examples.

### Details

The main function in the threshr package is `ithresh`, which uses leave-one-out cross-validation in a Bayesian setup to compare the predictive ability resulting from the use of each of a user-supplied set of thresholds.

See `vignette("threshr-vignette", package = "threshr")` for an overview of the package.

### References

Northrop, P. J. (2017). `revdbayes`: Ratio-of-Uniforms Sampling for Bayesian Extreme Value Analysis. R package version 1.2.1. <https://cran.r-project.org/package=revdbayes>.

Northrop, P. J., Attalides, N. and Jonathan, P. (2017) Cross-validators extreme value threshold selection and uncertainty with application to ocean storm severity. *Journal of the Royal Statistical Society Series C: Applied Statistics*, **66**(1), 93-120. <https://doi.org/10.1111/rssc.12159>

### See Also

The packages `revdbayes` and `rust`.

`ithresh` for threshold selection in the i.i.d. case based on leave-one-out cross-validation.

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