

# Package ‘Qindex’

October 12, 2022

**Type** Package

**Title** Quantile-Based Predictors for Survival Outcome

**Version** 0.1.0

**Date** 2022-09-14

**Author** Misung Yi [aut, cph],  
Tingting Zhan [aut, cre, cph] (<<https://orcid.org/0000-0001-9971-4844>>),  
Inna Chervoneva [aut, cph]

**Maintainer** Tingting Zhan <tingtingzhan@gmail.com>

**Description** Select optimal quantile-based predictors for survival outcome, to include the means to handle dichotomizing the quantiles, mean-signal-intensity or other continuous markers, and to obtain estimates for dichotomized predictors by Bootstrap-based bias correction.

**LazyData** true

**LazyDataCompression** xz

**RoxygenNote** 7.2.1

**Encoding** UTF-8

**License** GPL-2

**Depends** R (>= 4.2),

**Language** en-US

**Imports** stats, boot, rpart, survival, matrixStats

**Suggests** knitr

**Collate** '0info.R' 'pkg\_data\_doc.R' 'optQp.R' 'sampleQp.R'  
'BBC\_dichotom.R'

**NeedsCompilation** no

**Repository** CRAN

**Date/Publication** 2022-09-19 08:36:11 UTC

**R topics documented:**

Qindex-package . . . . .	2
BBC_dichotom . . . . .	2
celldata . . . . .	4
coxph_optQ . . . . .	5
dichotom_optQ . . . . .	5
evalQp . . . . .	6
optQp . . . . .	7
print.evalQp . . . . .	7
sampleQp . . . . .	8
summary.evalQp . . . . .	9

<b>Index</b>	<b>10</b>
--------------	-----------

---

Qindex-package	<i>Qindex</i>
----------------	---------------

---

**Description**

Optimize quantile-based predictors for survival outcome (and includes the means to handle dichotomizing the MSI or other continuous markers).

**Details**

Will provide later

**References**

paper in review

---

BBC_dichotom	<i>Bootstrap Bias Correction for Dichotomization</i>
--------------	--

---

**Description**

Bootstrap-based optimism correction for dichotomizing quantiles

**Usage**

BBC\_dichotom(formula, data, optQ, R, seed, ...)

**Arguments**

formula	<a href="#">formula</a>
data	<a href="#">data.frame</a>
optQ	<a href="#">character</a> scalar
R	see <a href="#">boot</a>
seed	<a href="#">integer</a> scalar, random seeds for generating repeated samples, see <a href="#">set.seed</a>
...	..

**Details**

The bootstrap optimism correction procedure (1) is performed as described for a general model selection in (2). First, R bootstrap samples are drawn with replacement from the main sample. In each bootstrap sample, the recursive partitioning tree model is used to establish an objective data-driven optimal cut-point for a specific optimal quantile. The cut-point from the current bootstrap sample is used to compute the log odds-ratio (OR) and/or hazards ratio (HR) for dichotomized quantile predictor in the current bootstrap sample (**bootstrap performance**) and in the main sample (**test performance**), and the optimism in log OR/HR estimation is computed as the difference between log OR/HR for “Bootstrap performance” and for “Test performance”. The median optimism estimate is computed as the median of optimism estimates over all bootstrap samples. The cutpoint for dichotomizing each selected optimal quantile is also established in the main sample and its “apparent performance” is computed as the log OR/HR for dichotomized quantile in the univariate logistic regression or Cox models. Finally, the optimism-corrected performance estimate is computed by subtracting the mean optimism estimate from the apparent performance estimate.

**Value**

..

**Examples**

```
Ki67_Qps = sampleQp(data = Ki67, endpoint = 'RECURRENCE', subjID = 'PATIENT_ID',
  Qpredictor = 'Marker', probs = seq(from = .05, to = .95, by = .05))
sapply(Ki67_Qps, FUN = class)
head(Ki67_Qps$Marker)
(Ki67_eval = evalQp(RECURRENCE ~ Marker, data = Ki67_Qps, seeds = 1:100))
lapply(Ki67_eval, FUN = dim)
summary(Ki67_eval)
Ki67_opt = optQp(Ki67_eval, n = 2L)
colnames(Ki67_opt$Marker) = paste0('Ki67_', colnames(Ki67_opt$Marker))

PR_Qps = sampleQp(data = PR, endpoint = 'RECURRENCE', subjID = 'PATIENT_ID',
  Qpredictor = 'Marker', probs = seq(from = .05, to = .95, by = .05))
head(PR_Qps$Marker)
PR_eval = evalQp(RECURRENCE ~ Marker, data = PR_Qps, seeds = 1:100)
PR_opt = optQp(PR_eval, n = 3L)
colnames(PR_opt$Marker) = paste0('PR_', colnames(PR_opt$Marker))
```

```

cellOpt0 = merge(Ki67_opt, PR_opt, by = setdiff(names(Ki67_opt), 'Marker'),
  suffixes = c('.Ki67', '.PR'))
cellOpt = within(cellOpt0, expr = {
  Marker = cbind(Marker.Ki67, Marker.PR)
  Marker.Ki67 = Marker.PR = NULL
})
dim(cellOpt)
names(cellOpt)
head(cellOpt$Marker)

mod = BBC_dichotom(RECURRENCE ~ NodeSt + Tstage, data = cellOpt,
  optQ = 'Marker', R = 100, seed = 1)
names(mod)
mod$ucox_dictom
mod$mcox_dictom
mod$median.optimism

```

---

celldata

*Subset of Ki67 and PR Data*


---

## Description

A subset of Ki67 cell data containing 623 patients and PR cell data containing 631 patients. A total of 572 patients are in common.

## Usage

Ki67

PR

## Format

**PATIENT\_ID** *factor*, ..

**tissueID** *character*, ..

**RECURRENCE** *Surv*, ...

**Marker** *numeric*, ...

**all other columns**

An object of class `data.frame` with 467603 rows and 20 columns.

---

coxph_optQ	..
------------	----

---

### Description

..

### Usage

```
coxph_optQ(
  formula,
  data,
  optQ,
  dQ = dichotom_optQ(edp = eval(formula[[2L]], envir = data), optQ = data[[optQ]])
)
```

### Arguments

formula	..
data	..
optQ	<a href="#">character</a> scalar
dQ	returned object from <a href="#">dichotom_optQ</a>

### Value

[coxph\\_optQ](#) returns a [numeric](#) vector of the concatenated coefficients of **univariable model** (which is returned in attribute 'ucox') and coefficients **multivariable model** (which is returned in attribute 'mcox').

---

dichotom_optQ	<i>Helper function <a href="#">dichotom_optQ</a></i>
---------------	--

---

### Description

..

### Usage

```
dichotom_optQ(edp, optQ, control = rpart.control(minbucket = 40))
```

### Arguments

edp	<a href="#">Surv</a> object, the endpoint
optQ	<a href="#">matrix</a> of optimal quantiles
control	see <a href="#">rpart</a>

**Value**

`dichotom_optQ` returns a [logical matrix](#) of the same dimension and dimension names as argument `optQ`, as dichotomized using the first node of `rpart` as threshold. The thresholds (per column) are returned as an [numeric](#) vector in attribute 'thres'.

---

 evalQp

*Optimal Quantile Predictor*


---

**Description**

From the sequence of sample quantiles, selects the optimal quantile that has highest predictive value for a given survival outcome

**Usage**

```
evalQp(formula, data, seeds, pct_train = 0.8, minbucket.p = 0.5)
```

**Arguments**

formula	<a href="#">formula</a> , currently only supports one <a href="#">Surv</a> endpoint and one predictor of the markers. See details of parameter data.
data	<a href="#">data.frame</a> , with at least <ul style="list-style-type: none"> <li>• one <a href="#">Surv</a> column as the survival outcome</li> <li>• one <a href="#">matrix</a> column as the predictor of quantile sequence</li> </ul>
seeds	<a href="#">integer</a> vector of random seeds for generating repeated samples, see <a href="#">set.seed</a>
pct_train	<a href="#">numeric</a> scalar, proportion of the training set, default .8
minbucket.p	(for <code>rpart</code> function) the minimum number of observations in any terminal <b>leaf</b> node. If only one of <code>minbucket</code> or <code>minsplit</code> is specified, the code either sets <code>minsplit</code> to <code>minbucket*3</code> or <code>minbucket</code> to <code>minsplit/3</code> , as appropriate.

**Value**

`evalQp` returns a `evalQp` object, which is a [list](#) with two [matrix](#) elements `thresholds` and `coefs`.

- `thresholds` is ...
- `coefs` is ...

**Examples**

```
# see ?optQp
```

---

optQp	<i>optQp</i>
-------	--------------

---

**Description***optQp***Usage**

optQp(x, n = 3L, ...)

**Arguments**

x	<i>evalQp</i> object
n	<i>integer</i> scalar
...	additional parameters, not currently in use

**Value***optQp* returns ..**Examples**

# see ?BBC\_dichotom

---

print.evalQp	<i>Print evalQp Object</i>
--------------	----------------------------

---

**Description**

..

**Usage**

```
## S3 method for class 'evalQp'
print(x, n = 3L, ...)
```

**Arguments**

x	..
n	..
...	..

**Details**

`print.evalQp` simply calls `summary.evalQp ..`

**Value**

`print.evalQp` does not have a returned value.

---

sampleQp	<i>Cluster-specific sample quantiles for clustered data</i>
----------	---

---

**Description**

For a user-supplied sequence of percentiles, calculates sample quantiles in each independent cluster of observations.

**Usage**

```
sampleQp(
  data,
  endpoint = "RECURRENCE",
  subjID = "PATIENT_ID",
  sampleID,
  Qpredictor = "Marker",
  probs = seq(from = 0.05, to = 0.95, by = 0.05),
  aggregate_sampleID_quantile = c("median", "mean", "min", "Q1", "Q3", "max"),
  ...
)
```

**Arguments**

<code>data</code>	<a href="#">data.frame</a>
<code>endpoint</code>	<a href="#">character</a> scalar, column name for the <a href="#">Surv</a> endpoint
<code>subjID</code>	<a href="#">character</a> scalar, column name for subject/patient ID
<code>sampleID</code>	(optional) <a href="#">character</a> scalar, column name for <code>sampleID</code> , which is nested within <code>subjID</code> . When <code>sampleID</code> is missing, the analysis is performed with only one-level cluster of <code>subjID</code>
<code>Qpredictor</code>	<a href="#">character</a> scalar, column name of the predictor variable
<code>probs</code>	<a href="#">numeric</a> vector, probabilities of the <a href="#">quantiles</a>
<code>aggregate_sampleID_quantile</code>	<a href="#">character</a> vector, will be implemented in the next release
<code>...</code>	additional parameters, currently not in use ..



**Value**

`sampleQp` returns data set with one row per subjID,  $N_p$  columns with  $Q(p)$ ,  $p=1, \dots, N_p$ , and additional subject level variables designated to be kept in the data set. The code should allow for multiple clusters (`clusterID`) per subject (`subjID`) and output the mean or other summary statistic (according to user input for type of summary statistic)  $Q(p)$  for every  $p$ . (\*) Let us first make use of `summary.default`, which cover all options I can think about for now

**Examples**

```
# see ?BBC_dichotom
```

---

summary.evalQp	<i>Summary Information of <code>evalQp</code> Object</i>
----------------	--

---

**Description**

Summary method for `optQp` Object

**Usage**

```
## S3 method for class 'evalQp'
summary(object, FUN = median, ...)
```

**Arguments**

object	<code>evalQp</code> object
FUN	<code>median</code> (default) or <code>mean</code>
...	..

**Value**

`summary.evalQp` summarize ...

# Index

- \* **datasets**
  - cellldata, 4
- \* **package**
  - Qindex-package, 2
- BBC\_dichotom, 2
- boot, 3
- cellldata, 4
- character, 3–5, 8
- coxph\_optQ, 5, 5
- data.frame, 3, 6, 8
- dichotom\_optQ, 5, 5, 6
- evalQp, 6, 6, 7, 9
- factor, 4
- formula, 3, 6
- integer, 3, 6, 7
- Ki67 (cellldata), 4
- list, 6
- logical, 6
- matrix, 5, 6
- mean, 9
- median, 9
- numeric, 4–6, 8
- optQp, 7, 7, 9
- PR (cellldata), 4
- print.evalQp, 7, 8
- Qindex-package, 2
- quantile, 8
- rpart, 5, 6
- sampleQp, 8, 9
- set.seed, 3, 6
- summary.default, 9
- summary.evalQp, 8, 9, 9
- Surv, 4–6, 8