# Package 'RcppDynProg'

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```
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Title 'Rcpp' Dynamic Programming
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     https://winvector.github.io/RcppDynProg/
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Description
     Dynamic Programming implemented in 'Rcpp'. Includes example partition and out of sample fit-
     ting applications. Also supplies additional custom coders for the 'vtreat' package.
License GPL-2 | GPL-3
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```

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# Description

Rcpp dynamic programming solutions for partitioning and machine learning problems. Includes out of sample fitting applications. Also supplies additional custom coders for the vtreat package. Please see <a href="https://github.com/WinVector/RcppDynProg">https://github.com/WinVector/RcppDynProg</a> for details.

# Author(s)

John Mount

#### See Also

Useful links:

- https://github.com/WinVector/RcppDynProg/
- https://winvector.github.io/RcppDynProg/
- Report bugs at https://github.com/WinVector/RcppDynProg/issues

const\_costs 3

const_costs	const_costs	

# **Description**

Built matrix of total out of sample interval square error costs for held-out means. One indexed.

# Usage

```
const_costs(y, w, min_seg, indices)
```

# **Arguments**

y Numeric Vector, values to group in order.

w NumericVector, weights.

min\_seg positive integer, minimum segment size (>=1).
indices IntegerVector, order list of indices to pair.

#### Value

xcosts NumericMatix, for j>=i xcosts(i,j) is the cost of partition element [i,...,j] (inclusive).

# **Examples**

```
const_costs(c(1, 1, 2, 2), c(1, 1, 1, 1), 1, 1:4)
```

```
const_costs_logistic const_costs_logistic
```

# Description

Built matrix of interval logistic costs for held-out means. One indexed.

# Usage

```
const_costs_logistic(y, w, min_seg, indices)
```

# **Arguments**

V	Numeric Vector, 0/1	values to group	in order (should be i	in interval [0,1]).

w Numeric Vector, weights (should be positive).
min\_seg positive integer, minimum segment size (>=1).
indices Integer Vector, order list of indices to pair.

lin\_costs

#### Value

xcosts NumericMatix, for j>=i xcosts(i,j) is the cost of partition element [i,...,j] (inclusive).

# **Examples**

```
const_costs_logistic(c(0.1, 0.1, 0.2, 0.2), c(1, 1, 1, 1), 1, 1:4)
```

lin\_costs

lin\_costs

# **Description**

Built matrix of interval costs for held-out linear models. One indexed.

#### Usage

```
lin_costs(x, y, w, min_seg, indices)
```

# **Arguments**

x Numeric Vector, x-coords of values to group.

y Numeric Vector, values to group in order.

w NumericVector, weights.

min\_seg positive integer, minimum segment size (>=1).
indices IntegerVector, ordered list of indices to pair.

#### Value

xcosts NumericMatix, for j>=i xcosts(i,j) is the cost of partition element [i,...,j] (inclusive).

```
lin_costs(c(1, 2, 3, 4), c(1, 2, 2, 1), c(1, 1, 1, 1), 1, 1:4)
```

lin\_costs\_logistic 5

<pre>lin_costs_logistic</pre>	lin_costs_logistic deviance costs.
IIII_COSCS_IOGISCIC	iii_cosis_iogisiic devidice cosis.

# **Description**

Built matrix of interval deviance costs for held-out logistic models. Fits are evaluated in-sample. One indexed.

# Usage

```
lin_costs_logistic(x, y, w, min_seg, indices)
```

# Arguments

x Numeric Vector,	x-coords of values to group.
-------------------	------------------------------

y Numeric Vector, values to group in order (should be in interval [0,1]).

w NumericVector, weights (should be positive).min\_seg positive integer, minimum segment size (>=1).

indices Integer Vector, ordered list of indices to pair.

# Value

```
xcosts NumericMatix, for j>=i xcosts(i,j) is the cost of partition element [i,...,j] (inclusive).
```

# **Examples**

```
lin_costs_logistic(c(1, 2, 3, 4, 5, 6, 7), c(0, 0, 1, 0, 1, 1, 0), c(1, 1, 1, 1, 1, 1, 1), 3, 1:7)
```

piecewise\_constant Piecewise constant fit.

# Description

```
vtreat custom coder based on RcppDynProg::solve_for_partition().
```

# Usage

```
piecewise_constant(varName, x, y, w = NULL)
```

#### **Arguments**

```
varName character, name of variable to work on.
x numeric, input values.
y numeric, values to estimate.
w numeric, weights.
```

# **Examples**

```
piecewise_constant("x", 1:8, c(-1, -1, -1, -1, 1, 1, 1, 1))
```

```
piecewise_constant_coder
```

Piecewise constant fit coder factory.

### **Description**

Build a piecewise constant fit coder with some parameters bound in.

#### Usage

```
piecewise_constant_coder(
  penalty = 1,
  min_n_to_chunk = 1000,
  min_seg = 10,
  max_k = 1000
)
```

#### **Arguments**

```
penalty per-segment cost penalty.

min_n_to_chunk minimum n to subdivied problem.

min_seg positive integer, minimum segment size.

max_k maximum segments to divide into.
```

#### Value

a vtreat coder

```
coder <- piecewise_constant_coder(min_seg = 1) coder("x", 1:8, c(-1, -1, -1, -1, 1, 1, 1, 1))
```

piecewise\_linear 7

piecewise\_linear

Piecewise linear fit.

# Description

```
vtreat custom coder based on RcppDynProg::solve_for_partition().
```

# Usage

```
piecewise_linear(varName, x, y, w = NULL)
```

#### **Arguments**

varName character, name of variable to work on.
x numeric, input values.

y numeric, values to estimate.

w numeric, weights.

# **Examples**

```
piecewise_linear("x", 1:8, c(1, 2, 3, 4, 4, 3, 2, 1))
```

```
piecewise_linear_coder
```

Piecewise linear fit coder factory.

# **Description**

Build a piecewise linear fit coder with some parameters bound in.

# Usage

```
piecewise_linear_coder(
  penalty = 1,
  min_n_to_chunk = 1000,
  min_seg = 10,
  max_k = 1000
)
```

8 score\_solution

# **Arguments**

```
penalty per-segment cost penalty.

min_n_to_chunk minimum n to subdivied problem.

min_seg positive integer, minimum segment size.

max_k maximum segments to divide into.
```

#### Value

a vtreat coder

# **Examples**

```
coder <- piecewise_linear_coder(min_seg = 1)
coder("x", 1:8, c(1, 2, 3, 4, 4, 3, 2, 1))</pre>
```

score\_solution

compute the price of a partition solution (and check is valid).

# **Description**

compute the price of a partition solution (and check is valid).

# Usage

```
score_solution(x, solution)
```

# **Arguments**

```
x NumericMatix, for j \ge i x(i,j) is the cost of partition element [i,...,j] (inclusive). solution vector of indices
```

# Value

price

```
x \leftarrow matrix(c(1,1,5,1,1,0,5,0,1), nrow=3)

s \leftarrow c(1, 2, 4)

score\_solution(x, s)
```

solve\_for\_partition 9

solve\_for\_partition Solve for a

Solve for a piecewise linear partiton.

### **Description**

Solve for a good set of right-exclusive x-cuts such that the overall graph of  $y\sim x$  is well-approximated by a piecewise linear function. Solution is a ready for use with with base::findInterval() and stats::approx() (demonstrated in the examples).

# Usage

```
solve_for_partition(
    x,
    y,
    ...,
    w = NULL,
    penalty = 0,
    min_n_to_chunk = 1000,
    min_seg = 1,
    max_k = length(x)
)
```

#### **Arguments**

```
    x numeric, input variable (no NAs).
    y numeric, result variable (no NAs, same length as x).
    ... not used, force later arguments by name.
    w numeric, weights (no NAs, positive, same length as x).
    penalty per-segment cost penalty.
    min_n_to_chunk minimum n to subdivied problem.
    min_seg positive integer, minimum segment size.
    max_k maximum segments to divide into.
```

#### Value

a data frame appropriate for stats::approx().

```
# example data
d <- data.frame(
    x = 1:8,
    y = c(1, 2, 3, 4, 4, 3, 2, 1))
# solve for break points</pre>
```

10 solve\_for\_partitionc

```
soln <- solve_for_partition(d$x, d$y)
# show solution
print(soln)

# label each point
d$group <- base::findInterval(
    d$x,
    soln$x[soln$what=='left'])
# apply piecewise approximation
d$estimate <- stats::approx(
    soln$x,
    soln$pred,
    xout = d$x,
    method = 'linear',
    rule = 2)$y
# show result
print(d)</pre>
```

solve\_for\_partitionc Solve for a piecewise constant partiton.

# Description

Solve for a good set of right-exclusive x-cuts such that the overall graph of y~x is well-approximated by a piecewise linear function. Solution is a ready for use with with base::findInterval() and stats::approx() (demonstrated in the examples).

#### Usage

```
solve_for_partitionc(
    x,
    y,
    ...,
    w = NULL,
    penalty = 0,
    min_n_to_chunk = 1000,
    min_seg = 1,
    max_k = length(x)
)
```

#### **Arguments**

```
    x numeric, input variable (no NAs).
    y numeric, result variable (no NAs, same length as x).
    ... not used, force later arguments by name.
    w numeric, weights (no NAs, positive, same length as x).
```

solve\_interval\_partition

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```
penalty per-segment cost penalty.

min_n_to_chunk minimum n to subdivied problem.

min_seg positive integer, minimum segment size.

max_k maximum segments to divide into.
```

#### Value

a data frame appropriate for stats::approx().

# **Examples**

```
# example data
d <- data.frame(</pre>
 x = 1:8,
 y = c(-1, -1, -1, -1, 1, 1, 1, 1)
# solve for break points
soln <- solve_for_partitionc(d$x, d$y)</pre>
# show solution
print(soln)
# label each point
d$group <- base::findInterval(</pre>
 d$x,
 soln$x[soln$what=='left'])
# apply piecewise approximation
d$estimate <- stats::approx(</pre>
 soln$x,
 soln$pred,
 xout = dx,
 method = 'constant',
 rule = 2)$y
# show result
print(d)
```

```
solve_interval_partition
```

solve\_interval\_partition interval partition problem.

# Description

Solve a for a minimal cost partition of the integers [1,...,nrow(x)] problem where for j>=i x(i,j). is the cost of choosing the partition element [i,...,j]. Returned solution is an ordered vector v of length k<=kmax where: v[1]==1, v[k]==nrow(x)+1, and the partition is of the form [v[i], v[i+1]) (intervals open on the right).

#### Usage

```
solve_interval_partition(x, kmax)
```

#### **Arguments**

x square NumericMatix, for  $j \ge i x(i,j)$  is the cost of partition element [i,...,j] (in-

clusive).

kmax int, maximum number of segments in solution.

#### Value

dynamic program solution.

#### **Examples**

```
costs <- matrix(c(1.5, NA ,NA ,1 ,0 , NA, 5, -1, 1), nrow = 3) solve_interval_partition(costs, nrow(costs))
```

```
solve_interval_partition_k
```

solve\_interval\_partition interval partition problem with a bound on number of steps.

# Description

Solve a for a minimal cost partition of the integers [1,...,nrow(x)] problem where for j>=i x(i,j). is the cost of choosing the partition element [i,...,j]. Returned solution is an ordered vector v of length k<=kmax where: v[1]==1, v[k]==nrow(x)+1, and the partition is of the form [v[i], v[i+1]) (intervals open on the right).

# Usage

```
solve_interval_partition_k(x, kmax)
```

# Arguments

x square NumericMatix, for  $j \ge i x(i,j)$  is the cost of partition element [i,...,j] (in-

clusive).

kmax int, maximum number of segments in solution.

#### Value

dynamic program solution.

#### **Examples**

```
costs <- matrix(c(1.5, NA ,NA ,1 ,0 , NA, 5, -1, 1), nrow = 3) solve_interval_partition(costs, nrow(costs))
```

```
solve_interval_partition_no_k
```

solve\_interval\_partition interval partition problem, no boun on the number of steps.

# **Description**

Not working yet.

#### Usage

```
solve_interval_partition_no_k(x)
```

# Arguments

Х

square NumericMatix, for  $j \ge i x(i,j)$  is the cost of partition element [i,...,j] (inclusive).

# **Details**

Solve a for a minimal cost partition of the integers [1,...,nrow(x)] problem where for j>=i x(i,j). is the cost of choosing the partition element [i,...,j]. Returned solution is an ordered vector v of length k where: v[1]==1, v[k]==nrow(x)+1, and the partition is of the form [v[i], v[i+1]) (intervals open on the right).

# Value

dynamic program solution.

```
costs <- matrix(c(1.5, NA ,NA ,1 ,0 , NA, 5, -1, 1), nrow = 3) solve_interval_partition(costs, nrow(costs))
```

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