Package 'FastImputation'

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Title Learn from training data then quickly fill in missing data.

Maintainer Stephen R. Haptonstahl < srh@haptonstahl.org>

Description TrainFastImputation uses training data to describe a

Type Package

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Suggests testthat

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Author Stephen R. Haptonstahl

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```
BoundNormalizedVariable
```

Take a normalized variable and transform it back to a bounded variable.

Description

This takes variables on the real line and constrains them to be on a half-line (constrained above or below) or a segment (constrained both above and below). This is approximately the inverse of NormalizeBoundedVariable; this does not completely reverse the effect of NormalizeBoundedVariable because NormalizeBoundedVariable first forces values away from the bounds, and this information is lost.

Usage

```
BoundNormalizedVariable(x, constraints)
```

Arguments

A vector, matrix, array, or dataframe with value to be coerced into a range or set.
 Constraints
 A list of constraints. See the examples below for formatting details.

Value

An object of the same class as x with the values transformed into the desired half-line or segment.

Author(s)

```
Stephen R. Haptonstahl < srh@haptonstahl.org>
```

Examples

```
constraints=list(lower=5)  # lower bound when constraining to an interval
  constraints=list(upper=10)  # upper bound when constraining to an interval
  constraints=list(lower=5, upper=10)  # both lower and upper bounds
```

FastImputation *Use the pattern learned from the training data to impute (fill in good guesses for) missing values.*

Description

Like Amelia, FastImputation assumes that the columns of the data are multivariate normal or can be transformed into approximately multivariate normal.

Usage

```
FastImputation(x, patterns, verbose = TRUE)
```

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Arguments

Vector, matrix, dataframe, or object that can be coerced into a dataframe, possi-

bly with some missing (NA) values.

patterns An object of class 'FastImputationPatterns' generated by TrainFastImputation.

verbose If TRUE then the progress in imputing the data will be shown.

Value

An object of class 'FastImputationPatterns' that contains information needed later to impute on a single row.

Author(s)

```
Stephen R. Haptonstahl < srh@haptonstahl.org>
```

References

```
http://gking.harvard.edu/amelia/
```

See Also

TrainFastImputation

Examples

FItest

Fraud Imputation Test Data

Description

Observations of Web financial transactions with some cells missing. Used with FastImputation.

Usage

FItest

4 FItrain

Format

A data frame with 10 variables and 10000 observations.

- 1. cust.id: Internal customer identification number
- 2. order.id: Unique identification number for this transaction (row)
- 3. is.fraud: 1 if the transaction is fraudulent, 0 otherwise
- 4. customer.age.yrs: Customer age in years; may be a decimal
- 5. spent.days.0to2: Amount spent in dollars by customer between 0 and 2 days before the current transaction
- 6. spent.days.3to10: Amount spent in dollars by customer between 3 and 10 days before the current transaction
- 7. spent.days.11to30: Amount spent in dollars by customer between 11 and 30 days before the current transaction
- 8. geo.ip.fraud.rate: Fraction between 0 and 1 of transactions from that geographic location (identified by IP address) that have been fraudulent
- 9. account.age.days: Integer number of days the customer has had the account
- 10. days.to.first.purchase Integer number of days between account creation and the first purchase by the customer

Author(s)

Stephen R. Haptonstahl < srh@haptonstahl.org>

Source

This is simulated data generated to be similar to real data.

FItrain

Fraud Training Data

Description

Complete observations of Web financial transactions. Used with TrainFastImputation to prepare for imputing individual transactions as they come in.

Usage

FItrain

Format

A data frame with 10 variables and 10000 observations.

- 1. cust.id: Internal customer identification number
- 2. order.id: Unique identification number for this transaction (row)
- 3. is.fraud: 1 if the transaction is fraudulent, 0 otherwise
- 4. customer.age.yrs: Customer age in years; may be a decimal

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5. spent.days.0to2: Amount spent in dollars by customer between 0 and 2 days before the current transaction

- 6. spent.days.3to10: Amount spent in dollars by customer between 3 and 10 days before the current transaction
- 7. spent.days.11to30: Amount spent in dollars by customer between 11 and 30 days before the current transaction
- 8. geo.ip.fraud.rate: Fraction between 0 and 1 of transactions from that geographic location (identified by IP address) that have been fraudulent
- 9. account.age.days: Integer number of days the customer has had the account
- 10. days.to.first.purchase Integer number of days between account creation and the first purchase by the customer

Author(s)

Stephen R. Haptonstahl < srh@haptonstahl.org>

Source

This is simulated data generated to be similar to real data.

FItrue

Fraud "True" Data

Description

Complete observations of Web financial transactions. Used to gauge the accuracy of imputation of FItest.

Usage

FItrue

Format

A data frame with 10 variables and 10000 observations.

- 1. cust.id: Internal customer identification number
- 2. order.id: Unique identification number for this transaction (row)
- 3. is. fraud: 1 if the transaction is fraudulent, 0 otherwise
- 4. customer.age.yrs: Customer age in years; may be a decimal
- 5. spent.days.0to2: Amount spent in dollars by customer between 0 and 2 days before the current transaction
- 6. spent.days.3to10: Amount spent in dollars by customer between 3 and 10 days before the current transaction
- 7. spent.days.11to30: Amount spent in dollars by customer between 11 and 30 days before the current transaction
- 8. geo.ip.fraud.rate: Fraction between 0 and 1 of transactions from that geographic location (identified by IP address) that have been fraudulent
- 9. account.age.days: Integer number of days the customer has had the account
- 10. days.to.first.purchase Integer number of days between account creation and the first purchase by the customer

6 LimitToSet

Author(s)

Stephen R. Haptonstahl < srh@haptonstahl.org>

Source

This is simulated data generated to be similar to real data.

LimitToSet

Coerce numeric values into a given set.

Description

Given some values x and a set of values set, each value in x is changed to the value in set that is closest.

Usage

```
LimitToSet(x, set)
```

Arguments

A vector, matrix, array, or dataframe with value to be coerced into a range or set.

set A list of values that x will be forced to take on.

Value

An object of the same class as x with values replaced as needed to satisfy the constraints.

Author(s)

Stephen R. Haptonstahl < srh@haptonstahl.org>

Examples

```
x <- runif(100, min=0, max=10)
y <- LimitToSet(x, set=c(1:10))
plot(x, y)</pre>
```

NormalizeBoundedVariable

NormalizeBoundedVariable

Take a variable bounded above/below/both and return an unbounded (normalized) variable.

Description

This transforms bounded variables so that they are not bounded. First variables are coerced away from the boundaries. by a distance of tol. The natural log is used for variables bounded either above or below but not both. The inverse of the standard normal cumulative distribution function (the quantile function) is used for variables bounded above and below.

Usage

```
NormalizeBoundedVariable(x, constraints, tol = pnorm(-5),
    trim = TRUE)
```

Arguments

x A vector, matrix, array, or dataframe with value to be coerced into a range or

set.

constraints A list of constraints. See the examples below for formatting details.

tol Variables will be forced to be at least this far away from the boundaries.

trim If TRUE values in x < lower and values in <math>x > upper will be set to lower and

upper, respectively, before normalizing.

Value

An object of the same class as x with the values transformed so that they spread out over any part of the real line.

A variable x that is bounded below by lower is transformed to log(x - lower).

A variable x that is bounded above by upper is transformed to log(upper - x).

A variable x that is bounded below by lower and above by upper is transformed to qnorm((x-lower)/(upper - lower)

Author(s)

```
Stephen R. Haptonstahl < srh@haptonstahl.org>
```

Examples

```
constraints=list(lower=5)  # lower bound when constraining to an interval
constraints=list(upper=10)  # upper bound when constraining to an interval
constraints=list(lower=5, upper=10) # both lower and upper bounds
```

8 TrainFastImputation

TrainFastImputation

Learn from the training data so that later you can fill in missing data

Description

Like Amelia, FastImputation assumes that the columns of the data are multivariate normal or can be transformed into approximately multivariate normal.

Usage

```
TrainFastImputation(x, constraints = list())
```

Arguments

Dataframe containing training data. Can have incomplete rows.A list of constraints. See the examples below for formatting details.

Value

An object of class 'FastImputationPatterns' that contains information needed later to impute on a single row.

Author(s)

```
Stephen R. Haptonstahl < srh@haptonstahl.org>
```

References

```
http://gking.harvard.edu/amelia/
```

See Also

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FastImputation
```

Examples

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UnfactorColumns

Convert columns of a dataframe from factors to character or numeric.

Description

Convert columns of a dataframe from factors to character or numeric.

Usage

UnfactorColumns(x)

Arguments

Х

A dataframe

Value

A dataframe containing the same data but any factor columns have been replaced with numeric or character columns.

Author(s)

Stephen R. Haptonstahl < srh@haptonstahl.org>

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