# root

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Name	ML_Core
Version	3.2.2
Description	Common definitions for Machine Learning
License	SeeLICENSE.TXT
Copyright	Copyright (C) 2019 HPCC Systems
Authors	HPCCSystems
Platform	6.2.0

## **OVERVIEW**

#### ML Core

Core ECL Machine Learning library

Definitions for common types and data manipulation utilities.

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# **Analysis**

Go Up

### **IMPORTS**

```
_versions.ML_Core.V3_2_2.ML_Core.Types | 
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```

#### **DESCRIPTIONS**

## **ANALYSIS** Analysis

Analysis

Analyze and assess the effectiveness of a Machine Learning model.

Sub-modules provide support for both Classification and Regression.

Each of the functions in this module support multi-work-item (i.e. Myriad interface) data, as well as multi-variate data (supported by some ML bundles). The number field, which is usually = 1 for uni-variate data is used to distinguish multiple regressors in the case of multi- variate models.

#### Children

- 1. Classification (From Analysis): This sub-module provides functions for analyzing and assessing the effectiveness of an ML Classification model
- 2. Regression (From Analysis): This sub-module provides functions for analyzing and assessing the effectiveness of an ML Regression model
- 3. FeatureSelection (From Analysis): This sub module provides functions for assessing the features of a dataset, to perform feature selection

4. Clustering (From Analysis): This sub module provides various tests that help evaluate the effectiveness of clustering algorithms

#### **CLASSIFICATION** Classification

#### Analysis \

#### Classification

This sub-module provides functions for analyzing and assessing the effectiveness of an ML Classification model. It can be used with any ML Bundle that supports classification.

#### Children

- 1. ClassStats (From Classification): Given a set of expected dependent values, assess the number and percentage of records that were of each class
- 2. ConfusionMatrix (From Classification): Returns the Confusion Matrix, counting the number of cases for each combination of predicted Class and actual Class
- 3. Accuracy (From Classification): Assess the overall accuracy of the classification predictions
- 4. AccuracyByClass (From Classification): Provides per class accuracy / relevance statistics (e.g.
- 5. AUC (From Classification): AUC Area under the Receiver Operating Characteristics (ROC) curve, is a measure of how well a classifier is able to distinguish between classes

### **CLASSSTATS** ClassStats

#### Analysis \ Classification \

DATASET(Class\_Stats) ClassStats

(DATASET(DiscreteField) actual)

Given a set of expected dependent values, assess the number and percentage of records that were of each class.

**PARAMETER** <u>actual</u> ||| TABLE ( DiscreteField ) — The set of training-data or test-data dependent values in DATASET(DiscreteField) format.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , INTEGER4 class , INTEGER4 classCount , REAL8 classPct } ) — DATASET(Class\_Stats), one record per work-item, per classifier (i.e. number field) per class.

SEE ML\_Core.Types.Class\_Stats

## **CONFUSIONMATRIX** ConfusionMatrix

Analysis \ Classification \

DATASET(Confusion_Detail)	ConfusionMatrix	
(DATASET(DiscreteField) predicted, DATASET(DiscreteField) actual)		

Returns the Confusion Matrix, counting the number of cases for each combination of predicted Class and actual Class.

**PARAMETER** <u>predicted</u> ||| TABLE ( DiscreteField ) — The predicted values for each id in DATASET(DiscreteField) format.

**PARAMETER** <u>actual</u> ||| TABLE ( DiscreteField ) — The actual (i.e. expected) values for each id in DATASET(DiscreteField) format.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , INTEGER4 actual\_class , INTEGER4 predict\_class , UNSIGNED4 occurs , BOOLEAN correct , REAL8 pctActual , REAL8 pctPred } ) — DATASET(Confusion\_Detail). One record for each combination of work-item, number (i.e. classifier), predicted class, and actual class.

SEE ML\_Core.Types.Confusion\_Detail

## **ACCURACY** Accuracy

Analysis \ Classification \

#### DATASET(Classification\_Accuracy)

Accuracy

(DATASET(DiscreteField) predicted, DATASET(DiscreteField) actual)

Assess the overall accuracy of the classification predictions.

ML\_Core.Types.Classification\_Accuracy provides a detailed description of the return values.

**PARAMETER predicted** ||| TABLE ( DiscreteField ) — The predicted values for each id in DATASET(DiscreteField) format.

**PARAMETER** <u>actual</u> ||| TABLE ( DiscreteField ) — The actual (i.e. expected) values for each id in DATASET(DiscreteField) format.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , UNSIGNED8 recCnt , UNSIGNED8 errCnt , REAL8 Raw\_Accuracy , REAL8 PoD , REAL8 PoDE , REAL8 Hamming\_Loss } ) — DATASET(Classification\_Accuracy). One record for each combination of work-item, and number (i.e. classifier).

SEE ML\_Core.Types.Classification\_Accuracy

### **ACCURACYBYCLASS** AccuracyByClass

Analysis \ Classification \

## DATASET(Class\_Accuracy) | AccuracyByClass

(DATASET(DiscreteField) predicted, DATASET(DiscreteField) actual)

Provides per class accuracy / relevance statistics (e.g. Precision / Recall, False-positive Rate).

ML\_Core.Types.Class\_Accuracy provides a detailed description of the return values.

**PARAMETER predicted** ||| TABLE ( DiscreteField ) — The predicted values for each id in DATASET(DiscreteField) format.

**PARAMETER** <u>actual</u> ||| TABLE ( DiscreteField ) — The actual (i.e. expected) values for each id in DATASET(DiscreteField) format.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , INTEGER4 class , REAL8 precision , REAL8 recall , REAL8 FPR , REAL8 f\_score } ) — DATASET(Class\_Accuracy). One record for each combination of work-item, number (i.e. classifier), and class.

SEE ML\_Core.Types.Class\_Accuracy

## **AUC AUC**

Analysis \ Classification \

DATASET(AUC\_Result) AUC

(DATASET(Classification\_Scores) scores, DATASET(DiscreteField) actual)

AUC Area under the Receiver Operating Characteristics (ROC) curve, is a measure of how well a classifier is able to distinguish between classes. The ROC curve is a plot of the true positive rate vs. the false positive rate with varying threshold values. The value of this metric ranges from 0 to 1. Higher values are an indication of better classifiers.

**PARAMETER** scores ||| TABLE (Classification\_Scores) — The probability or confidence per class that a sample belongs to that class in DATASET(Classification\_Scores) format

**PARAMETER** <u>actual</u> ||| TABLE ( DiscreteField ) — The actual class to which a sample belongs in DATASET(DiscreteField) format

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , INTEGER4 class , REAL8 auc } ) — DATASET(AUC\_Result) The AUC score, per class, per classifier, per work item

SEE ML\_Core.Types.AUC\_Result, ML\_Core.Types.Classification\_Score

## **REGRESSION** Regression

Analysis \

#### Regression

This sub-module provides functions for analyzing and assessing the effectiveness of an ML Regression model. It can be used with any ML Bundle that supports regression.

#### Children

1. Accuracy (From Regression): Assess the overall accuracy of the regression predictions

# **ACCURACY** Accuracy

Analysis \ Regression \

DATASET(Regression_Accuracy)	Accuracy	
(DATASET(NumericField) predicted, DATASET(NumericField) actual)		

Assess the overall accuracy of the regression predictions.

 $\operatorname{ML\_Core}$ . Types. Regression\\_Accuracy provides a detailed description of the return values.

**PARAMETER** <u>predicted</u> ||| TABLE ( NumericField ) — The predicted values for each id in DATASET(NumericField) format.

**PARAMETER** <u>actual</u> ||| TABLE ( NumericField ) — The actual (i.e. expected) values for each id in DATASET(NumericField) format.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 regressor , REAL8 R2 , REAL8 MSE , REAL8 RMSE } ) — DATASET(Regression\_Accuracy). One record for each combination of work-item, and number (i.e. regressor).

SEE ML\_Core.Types.Regression\_Accuracy

## FEATURESELECTION FeatureSelection

Analysis \

#### **FeatureSelection**

This sub module provides functions for assessing the features of a dataset, to perform feature selection.

#### Children

- 1. Contingency (From FeatureSelection): Contingency Provides the contingency table for each combination of feature and sample (classifier)
- 2. Chi2 (From FeatureSelection): Chi2 Provides Chi2 coefficient and number of degrees of freedom for each combination of feature and classifier

## **CONTINGENCY** Contingency

Analysis \ FeatureSelection \

```
DATASET(Contingency_Table) Contingency

(DATASET(DiscreteField) samples, DATASET(DiscreteField) features)
```

Contingency Provides the contingency table for each combination of feature and sample (classifier). The contingency table represents the number of samples present in the data for each combination of sample category and feature category. Can only be used when both classifier and feature are discrete. The sets provided need not be sample / feature sets. They can be any two discrete fields whose contingency table is needed.

```
PARAMETER samples ||| TABLE ( DiscreteField ) — The samples or dependent values in DATASET(DiscreteField) format
```

**PARAMETER** <u>features</u> ||| TABLE ( DiscreteField ) — The features or independent values in DATASET(DiscreteField) format

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 fnumber , UNSIGNED4 snumber , INTEGER4 fclass , INTEGER4 sclass , INTEGER8 cnt } ) — DATASET(Contingency\_Table) The contingency table for each combination of sample (classifier) and feature, per work item

SEE ML\_Core.Types.Contingency\_Table

## CHI2 Chi2

Analysis \ FeatureSelection \

DATASET(Chi2_Result)	Chi2		
(DATASET(DiscreteField) features, DATASET(DiscreteField) samples)			

Chi2 Provides Chi2 coefficient and number of degrees of freedom for each combination of feature and classifier. Chi squared test is a statistical measure that helps establish the dependence of two categorical variables. In machine learning, it can be used to determine whether a classifier is dependent on a certain feature, and thus helps in feature selection. This test can only be used when both variables are categorical.

```
PARAMETER samples ||| TABLE ( DiscreteField ) — The samples or dependent values in DATASET(DiscreteField) format
```

**PARAMETER** <u>features</u> ||| TABLE ( DiscreteField ) — The features or independent values in DATASET(DiscreteField) format

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 fnumber , UNSIGNED4 snumber , INTEGER8 dof , REAL8 x2 , REAL8 p } ) — DATASET(Chi2\_Result) Chi square values and degrees of freedom for each combination of feature and classifier, per work item.

SEE ML\_Core.Types.Chi2\_Result

## **CLUSTERING** Clustering

Analysis \

Clustering

This sub module provides various tests that help evaluate the effectiveness of clustering algorithms.

#### Children

1. ARI (From Clustering) : ARI The Rand index is a measure of the similarity between two data clusterings

- 2. SampleSilhouetteScore (From Clustering): SampleSilhouetteScore Silhouette analysis measures the closeness of a point, both with its assigned cluster and with other clusters
- 3. SilhouetteScore (From Clustering): SilhouetteScore Silhouette analysis measures the closeness of a point, both with its assigned cluster and with other clusters

## **ARI** ARI

Analysis \ Clustering \

DATASET(ARI\_Result) ARI

(DATASET(ClusterLabels) predicted, DATASET(ClusterLabels)
actual)

ARI The Rand index is a measure of the similarity between two data clusterings. Adjusted Rand Index (ARI) is a version of rand index which is corrected for chance. This measure assumes values between -1 and 1. It produces values close to zero for random clusterings, values close to 1 for good clusterings and values close to -1 for clusterings that are worse than random guesses.

PARAMETER predicted || TABLE ( ClusterLabels ) — The labels predicted by the model in DATASET(ClusteringLabels) Format

PARAMETER <u>actual</u> ||| TABLE ( ClusterLabels ) — The actual labels, or the 'Ground Truth' in DATASET(ClusteringLabels) Format

**RETURN TABLE ( { UNSIGNED2 wi , REAL8 value } )** — DATASET(ARI\_Result) The adjusted rand index per work item

SEE ML\_Core.Types.ClusterLabels, ML\_Core.Types.ARI\_Result

# SAMPLESILHOUETTESCORE SampleSilhouetteScore

Analysis  $\setminus$  Clustering  $\setminus$ 

DATASET(SampleSilhouette\_Result) SampleSilhouetteScore

(DATASET(NumericField) samples,
DATASET(ClusterLabels) labels)

SampleSilhouetteScore Silhouette analysis measures the closeness of a point, both with its assigned cluster and with other clusters. It provides an easy way of finding the optimum value for k during k-means clustering. Silhouette values lie in the range of (-1, 1). A value of +1 indicates that the sample point is far away from its neighboring cluster and very close to the cluster to which it is assigned. The euclidian distance metric is used to measure the distances between points.

**PARAMETER** samples ||| TABLE ( NumericField ) — The datapoints / independent data in DATASET(NumericField) format

**PARAMETER** <u>labels</u> ||| TABLE ( ClusterLabels ) — The labels assigned to these datapoints in DATASET(ClusterLabels) format

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , REAL8 value } ) -

SEE ML\_Core.Types.SampleSilhouette\_Result

RESULT DATASET(SampleSilhouette\_Result) The silhouette coefficient per sample, per work item

# SILHOUETTESCORE SilhouetteScore

Analysis \ Clustering \

```
DATASET(Silhouette_Result) SilhouetteScore

(DATASET(NumericField) samples, DATASET(ClusterLabels) labels)
```

SilhouetteScore Silhouette analysis measures the closeness of a point, both with its assigned cluster and with other clusters. It provides an easy way of finding the optimum value for k during k-means clustering. Silhouette values lie in the range of (-1, 1). A value of +1 indicates that the sample point is far away from its neighboring cluster and very close to the cluster to which it is assigned. The euclidian distance metric is used to measure the distances between points. This function produces an average over SampleSilhouetteScore

**PARAMETER** samples ||| TABLE ( NumericField ) — The datapoints/independent data in DATASET(NumericField) format

RETURN TABLE ( { UNSIGNED2 wi , REAL8 score } ) —

- SEE ML\_Core.Types.SampleSilhouette\_Result, ML\_Core.Analysis.SampleSilhouetteScore
- **RESULT** DATASET(Silhouette\_Result) The silhouette coefficient per work item

# AppendID

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#### **DESCRIPTIONS**

## **APPENDID** AppendID

/ EXPORT	AppendID
(dIn,idfield,dOut)	

Macro takes any structured dataset, and appends a unique 1-based record ID column to it. Values will not be sequential and values will not be dense because of data skew. Gaps will appear when data ends on each node. If dense and sequential values are required, use AppendSeqID.

Note that, as a macro, nothing is returned, but attribute named in dOut will be defined to contain the resulting dataset.

Example:

ML\\_Core.AppendID(dOrig, recID, dOrigWithId);

PARAMETER  $\underline{dIn}$  ||| INTEGER8 — The name of the input dataset.

**PARAMETER** <u>idfield</u> ||| INTEGER8 — The name of the field to be appended containing the id for each row.

PARAMETER <u>dOut</u> || INTEGER8 — The name of the resulting dataset.

RETURN —

# AppendSeqID

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# **DESCRIPTIONS**

## APPENDSEQID AppendSeqID

- EXPORT	AppendSeqID
(dIn,idfield,dOut)	

Macro takes any structured dataset, and appends a unique 1-based record ID column to it. Values will be in data sequence. Note: implemented as a count project, each node processes the data in series instead of parallel. For better cluster performance, use AppendID as long as dense, sequential ids are not needed.

Note that, as a macro, nothing is returned, but attribute named in dOut will be defined to contain the resulting dataset.

Example:

ML\\_Core.AppendSeqID(dOrig, recID, dOrigWithId);

PARAMETER  $\underline{dIn}$  ||| INTEGER8 — The name of the input dataset.

**PARAMETER** <u>idfield</u> ||| INTEGER8 — The name of the field to be appended containing the id for each row.

PARAMETER <u>dOut</u> || INTEGER8 — The name of the resulting dataset.

RETURN —

# Config

Go Up

# **DESCRIPTIONS**

# **CONFIG** Config

Config

Global configuration constants that can be modified if needed.

#### Children

- 1. MaxLookup (From Config): The maximum amount of data to use in a LOOKUP JOIN
- 2. Discrete (From Config): The default number of groups to use when discretizing data
- 3. RoundingError (From Config): The tolerance for rounding error

# MAXLOOKUP MaxLookup

Config  $\setminus$ 

MaxLookup

The maximum amount of data to use in a LOOKUP JOIN.

RETURN INTEGER8 —

# **DISCRETE** Discrete

Config '	\
----------	---

Discrete

The default number of groups to use when discretizing data.

RETURN INTEGER8 —

# ROUNDINGERROR RoundingError

## Config $\setminus$

RoundingError

The tolerance for rounding error.

RETURN REAL8 —

# **Constants**

Go Up

# **DESCRIPTIONS**

# **CONSTANTS** Constants

Constants

Useful constants used in ML.

#### Children

- 1. Pi (From Constants) : Constant PI
- 2. Root\_2 (From Constants) : Constant square root of 2

PI Pi

Constants  $\setminus$ 

 $\mathbf{Pi}$ 

Constant PI

RETURN REAL8 —

# ROOT\_2 Root\_2

Constants \

Root\_2

Constant square root of 2

RETURN REAL8 —

# **CrossValidation**

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## **IMPORTS**

```
__versions.ML_Core.V3_2_2.ML_Core |
__versions.ML_Core.V3_2_2.ML_Core.Types |
```

## **DESCRIPTIONS**

# **CROSSVALIDATION** CrossValidation

#### CrossValidation

This module is a container for any cross-validation methods

#### Children

1. NFoldCV (From CrossValidation):

N-Fold Cross Validation is a way to validate the effectiveness of a regression or classification without having to segregate test data from training data

## **NFOLDCV** NFoldCV

CrossValidation \

#### **NFoldCV**

(LearnerName, IndepDS, DepDS, NumFolds)

N-Fold Cross Validation is a way to validate the effectiveness of a regression or classification without having to segregate test data from training data.

The results of the N-Fold Cross Validation approximate the expected result of training on all of the data samples and testing those results on other data from the same distribution.

This allows a model that is built on all available labeled data to be effectively assessed. Note that this process does not produce the target model, but only estimates the 'out-of-sample' error statistics that such a model would produce.

The method is as follows:

- Randomly split independent and dependent data into N (e.g. 10) 'folds'.
- Train N separate models, using N-1 of the folds as training data (e.g. 9).
- Test each model using the 1 fold that was not in the training set.
- Aggregate the test results across the N tests.

Any of the HPCC Machine Learning methods may be used with N-Fold Cross Validation The ML module to be used is passed as a parameter.

N-Fold Cross Validation can be used for regression or classification. If the dependent data is in NumericField format, it is treated as a regression and regression analytics are returned. If it is in DiscreteField format, then it is treated as a Classification, and Classification analytics are return.

Using the wrong dependent data type for the given learner will result in un-handled errors.

The returned MODULE exports the following attributes:

For Classification:

- ClassStats Assesses Classes Contained in the Training Data (see Types.Class\_Stats).
- Accuracy Overall Accuracy of the classification (see Types. Classification Accuracy).
- AccuracyByClass Precision and Recall for each class (see Types.Class\_Accuracy).
- ConfusionMatrix Frequency of predicted / actual class pairings (see Types.Consusion\_Detail).

#### For Regression:

• Accuracy (see Types.Regression Accuracy).

- PARAMETER <u>LearnerName</u> || INTEGER8 The attribute that holds the instantiated ML module.
- PARAMETER IndepDS || INTEGER8 The independent data to be used for training and testing.
- PARAMETER DepDS || INTEGER8 The dependent data to be used for training and testing.
- PARAMETER NumFolds || INTEGER8 The number of folds to use. Ten is typically considered adequate.

**RETURN BOOLEAN** — Result MODULE with attributes for assessing the strength of the model.

# **Discretize**

Go Up

## **IMPORTS**

\_versions.ML\_Core.V3\_2\_2.ML\_Core.Types |

## **DESCRIPTIONS**

## **DISCRETIZE** Discretize

Discretize

This module is used to turn a dataset of NumericFields into a dataset of DiscreteFields. This is not quite as trivial as it seems as there are a number of different ways to make the underlying data discrete; and even within one method there may be different parameters. Further - it is quite probable that different methods are going to be desired for each field.

There are two methods of interfacing:

- Call a discretization method directly to apply to all fields.
- Build a set of instructions on how to discretize each field and then call 'Do'.

The record format 'r Method is used to build the set of instructions in the latter case.

For each discretization method (e.g. ByRounding), there is a corresponding attribute preceded by 'i\_' that is used to build the r\_Method instruction for using that method (e.g. i\_ByRounding).

Three methods are currently provided:

• ByRounding – Numerically round the number to the nearest integer.

- ByBucketing Split the range of each variable into a number of evenly spaced buckets.
- ByTiling Splits the datapoints into an ordered set of equal-sized groups.

#### Children

- 1. c\_Method (From Discretize): Enumerate the available discretization methods
- 2. r\_Method (From Discretize): This format is used to construct an 'instruction stream' to allow a dataset to be discretized according to a set of instructions which are in (meta)data
- 3. i\_ByRounding (From Discretize): Construct an instruction (rMethod) that will cause certain fields to be discretized by rounding
- 4. ByRounding (From Discretize): Round the values passed in to create a discrete element Scale is applied (by multiplication) first and can be used to bring the data into a desired range (rParam1), Delta is applied (by addition) second and can be used to re-base a range OR to cause truncation or roundup as required (rParam2)
- 5. i\_ByBucketing (From Discretize): Construct an instruction (rMethod) that will cause certain fields to be discretized by bucketing
- 6. ByBucketing (From Discretize): Allocates a continuous variable into one of N buckets based upon an equal division of the RANGE of the variable
- 7. i\_ByTiling (From Discretize): Construct an instruction (rMethod) that will cause certain fields to be discretized by tiling
- 8. ByTiling (From Discretize): Allocate a continuous variable into one of N groups such that each group (tile) contains roughly the same number of entries and that all of the elements of group 2 have a higher value than group 1, etc
- 9. Do (From Discretize): Execute a set of discretization instructions in order to discretize all of the fields of the dataset using the appropriate methods

## C\_METHOD c\_Method

Discretize \

c Method

Enumerate the available discretization methods.

RETURN UNSIGNED4 —

VALUE Rounding = 1

VALUE Bucketing = 2

**VALUE** Tiling = 3

## R\_METHOD r\_Method

#### Discretize \

#### r Method

This format is used to construct an 'instruction stream' to allow a dataset to be discretized according to a set of instructions which are in (meta)data. It can be created directly, though the preferred method is to call i\_ByRounding(...), i\_ByBucketing(...), or i\_ByTiling(...) to create each record.

**FIELD** <u>method</u> ||| UNSIGNED4 — Indicator of the method to use (see c\_method).

**FIELD** <u>iParam1</u> || INTEGER8 — The first integer parameter to the discretization method.

FIELD <u>rParam1</u> ||| REAL8 — The first real parameter.

FIELD <u>rParam2</u> ||| REAL8 — The second real parameter.

FIELD fields ||| SET ( UNSIGNED4 ) — No Doc

## I\_BYROUNDING i\_ByRounding

#### Discretize \

### i\_ByRounding

(SET OF Types.t FieldNumber f, REAL Scale=1.0, REAL Delta=0.0)

Construct an instruction (rMethod) that will cause certain fields to be discretized by rounding. See ByRounding below.

**PARAMETER**  $\underline{\mathbf{f}} \parallel \parallel \text{SET} (\text{UNSIGNED4}) - \text{A set of field numbers to which to apply this method.}$ 

**PARAMETER** Scale ||| REAL8 — (Optional) A number by which to multiply each field before rounding.

**PARAMETER** <u>Delta</u> ||| REAL8 — (Optional) An offset that is applied after scaling but before rounding.

**RETURN TABLE** ( **r\_Method** ) — DATASET(r\_Method) containing one record.

## BYROUNDING ByRounding

Discretize \

#### **ByRounding**

(DATASET(Types.NumericField) d, REAL Scale=1.0, REAL Delta=0.0)

Round the values passed in to create a discrete element Scale is applied (by multiplication) first and can be used to bring the data into a desired range (rParam1), Delta is applied (by addition) second and can be used to re-base a range OR to cause truncation or roundup as required (rParam2).

**PARAMETER** Scale ||| REAL8 — (Optional) A number by which to multiply each field before rounding.

PARAMETER Delta ||| REAL8 — (Optional) An offset that is applied after scaling but before rounding.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , INTEGER4 value } ) — DATASET(DiscreteField) containing the discretized dataset.

## I\_BYBUCKETING i\_ByBucketing

Discretize \

#### $i\_ByBucketing$

(SET OF Types.t\_FieldNumber f, Types.t\_Discrete N=ML\_Core.Config.Discrete)

Construct an instruction (rMethod) that will cause certain fields to be discretized by bucketing. See ByBucketing below.

**PARAMETER**  $\underline{\mathbf{f}}$  ||| SET (UNSIGNED4) — A set of field numbers to which to apply this method.

**PARAMETER** N || INTEGER4 — (Optional) The number of buckets into which to split the range. The default is to use the ML\_Core. Config.Discrete configuration parameter.

**RETURN TABLE** ( **r\_Method** ) — DATASET(r\_Method) containing one record.

## BYBUCKETING ByBucketing

Discretize \

#### **ByBucketing**

(DATASET(Types.NumericField) d, Types.t Discrete N=ML Core.Config.Discrete)

Allocates a continuous variable into one of N buckets based upon an equal division of the RANGE of the variable.

The buckets will NOT have an even number of elements unless the underlying distribution of the variable is uniform.

**PARAMETER** N || INTEGER4 — (Optional) The number of buckets into which to split the range. The default is to use the ML\_Core. Config.Discrete configuration parameter.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , INTEGER4 value } ) — DATASET(DiscreteField) containing the discretized dataset.

## I\_BYTILING i\_ByTiling

Discretize \

#### i\_ByTiling

(SET OF Types.t FieldNumber f, Types.t Discrete N=ML Core.Config.Discrete)

Construct an instruction (rMethod) that will cause certain fields to be discretized by tiling. See ByTiling below.

**PARAMETER** f || SET (UNSIGNED4) — A set of field numbers to which to apply this method.

**PARAMETER** N || INTEGER4 — (Optional) The number of tiles into which to split the data. The default is to use the ML\_Core. Config.Discrete configuration parameter.

**RETURN TABLE** ( **r\_Method** ) — DATASET(r\_Method) containing one record.

## **BYTILING** ByTiling

Discretize \

#### **ByTiling**

(DATASET(Types.NumericField) d, Types.t\_Discrete N=ML\_Core.Config.Discrete)

Allocate a continuous variable into one of N groups such that each group (tile) contains roughly the same number of entries and that all of the elements of group 2 have a higher value than group 1, etc.

**PARAMETER** N | | INTEGER4 — (Optional) The number of tiles to create. The default is to use the ML\_Core. Config.Discrete configuration parameter.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , INTEGER4 value } ) — DATASET(DiscreteField) containing the discretized dataset.

## DO Do

Discretize \

 $\mathbf{Do}$ 

(DATASET(Types.NumericField) d, DATASET(r Method) to do)

Execute a set of discretization instructions in order to discretize all of the fields of the dataset using the appropriate methods.

Note that the file d is read once for each instruction - so it is much better to combine the instructions for multiple fields into one (provided the parameters and method are the same).

PARAMETER <u>d</u> || TABLE ( NumericField ) — The NumericField dataset to be dicretized.

**PARAMETER** to\_do ||| TABLE ( r\_Method ) — The DATASET(r\_Method) that contains the discretization instructions.

**RETURN** TABLE ( DiscreteField ) — DATASET(DiscreteField) containing the discretized dataset.

# **FieldAggregates**

Go Up

### **IMPORTS**

```
__versions.ML_Core.V3_2_2.ML_Core.Types |
__versions.ML_Core.V3_2_2.ML_Core.Utils | std.system.ThorLib |
```

#### **DESCRIPTIONS**

# FIELDAGGREGATES FieldAggregates

```
FieldAggregates
```

(DATASET(Types.NumericField) d)

Calculate various statistical aggregations of the fields in a NumericField dataset.

PARAMETER d | | TABLE ( NumericField ) — The dataset to be aggregated.

#### Children

- 1. Simple (From FieldAggregates): Calculate basic statistics about each field
- 2. SimpleRanked (From FieldAggregates): Calculate the rank (order) of each cell for each field
- 3. Medians (From FieldAggregates): Calculate the median value of each field
- 4. MinMedNext (From FieldAggregates): No Documentation Found
- 5. Buckets (From FieldAggregates): Bucketize the datapoints into N buckets for each field
- 6. BucketRanges (From FieldAggregates): Return the ranges associated with each of N buckets as computed by 'Buckets' above

- 7. Modes (From FieldAggregates): Calculate the mode (i.e.
- 8. Cardinality (From FieldAggregates): Returns the cardinality of each field
- 9. RankedInput (From FieldAggregates): No Documentation Found
- 10. NTiles (From FieldAggregates): Calculate the N-tile of each datapoint within its field
- 11. NTileRanges (From FieldAggregates): Return the ranges associated with each of N-tiles as computed by 'Ntiles' above
- 12. HistBins (From FieldAggregates): No Documentation Found
- 13. HistBinRanges (From FieldAggregates): No Documentation Found
- 14. PearsonCorr (From FieldAggregates): No Documentation Found
- 15. SpearmanCorr (From FieldAggregates): No Documentation Found
- 16. KendallCorr (From FieldAggregates): No Documentation Found
- 17. GenSpearman2Corr (From FieldAggregates): No Documentation Found

## **SIMPLE** Simple

FieldAggregates \

#### Simple

Calculate basic statistics about each field.

Calculates: min, max, sum, count, mean, variance, and standard deviation for each field.

There are no parameters.

Example:

```
myAggs := FieldAggregates(myDS).simple;
```

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 number , REAL8 minval , REAL8 maxval , REAL8 sumval , REAL8 countval , REAL8 mean , REAL8 var , REAL8 sd } ) —

## SIMPLERANKED SimpleRanked

FieldAggregates \

#### SimpleRanked

Calculate the rank (order) of each cell for each field.

The returned data adds a 'Pos' field to each cell, indicating its rank within it's field number.

There are no parameters.

Example:

myRankedDS := FieldAggregates(myDS).SimpleRanked;

# **MEDIANS** Medians

FieldAggregates \

Medians

Calculate the median value of each field.

There are no parameters.

**RETURN** TABLE ( { UNSIGNED2 wi , UNSIGNED4 number , REAL8 median } ) — DATASET({wi, number, median}), one record per work-item and field number. Example: </pr>myFieldMedians := FieldAggregates(myDS).Medians;

# MINMEDNEXT MinMedNext

 ${\bf Field Aggregates} \ \backslash \\$ 

MinMedNext

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 number , REAL8 median , REAL8 nextval , REAL8 minval , REAL8 maxval , REAL8 sumval , REAL8 countval , REAL8 mean , REAL8 var , REAL8 sd } ) —

### **BUCKETS** Buckets

FieldAggregates \

#### **Buckets**

(Types.t\_Discrete n)

Bucketize the datapoints into N buckets for each field.

Bucketization splits the range of the data into N equal size range buckets. The data will not normally be evenly split among buckets unless it is uniformly distributed. Contrast this with N-tile, where the data is split nearly evenly.

**PARAMETER**  $\underline{\mathbf{n}}$  ||| INTEGER4 — The number of buckets to use.

**RETURN** TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value , UNSIGNED8 Pos , INTEGER4 bucket } ) — DATASET OF {wi, id, number, value, pos, bucket}, where pos is the rank within each field, and bucket is the bucket number.

### **BUCKETRANGES BucketRanges**

FieldAggregates \

#### **BucketRanges**

(Types.t\_Discrete n)

Return the ranges associated with each of N buckets as computed by 'Buckets' above.

**PARAMETER**  $\underline{\mathbf{n}}$  ||| INTEGER4 — The number of buckets.

RETURN TABLE ( { UNSIGNED2	wi , UNSIGNED4 number , INTEGER4 bucket ,
REAL8 Min , REAL8 Max , U	${f NSIGNED8}$ ${f cnt}$ $\}$ ) — DATASET OF {wi, number, bucket,
Min, and Max}, one for each bucket	t for each field.

# **MODES** Modes

FieldAggregates \

Modes

Calculate the mode (i.e. the most common value) for each field

There are no parameters.

**RETURN** TABLE ( { UNSIGNED2 wi , UNSIGNED4 number , REAL8 mode , UNSIGNED8 cnt } ) — DATASET OF {wi, number, mode, cnt}, one per field. 'cnt' is the number of times the mode value occurred.

## **CARDINALITY** Cardinality

FieldAggregates \

Cardinality

Returns the cardinality of each field. That is the number of different values occurring in each field.

There are no parameters.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 number , UNSIGNED8 cardinality } ) — DATASET OF {wi, number, cardinality}, one per field.

## RANKEDINPUT RankedInput

FieldAggregates \

RankedInput

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value , REAL8 Pos } ) —

### **NTILES NTiles**

FieldAggregates \

NTiles

(Types.t\_Discrete n)

Calculate the N-tile of each datapoint within its field. For example, if N is 100, we calculate percentiles.

PARAMETER  $\underline{\mathbf{n}}$  ||| INTEGER4 — The number of groups into which to balance the data

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value , REAL8 Pos , INTEGER4 ntile } ) — DATASET OF {wi, id, number, value, pos, ntile}, where pos is the rank within each field.

## **NTILERANGES NTileRanges**

FieldAggregates \

NTileRanges

(Types.t Discrete n)

Return the ranges associated with each of N-tiles as computed by 'Ntiles' above.

PARAMETER <u>n</u> || INTEGER4 — The number of N-tile groups.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 number , INTEGER4 ntile , REAL8 Min , REAL8 Max , UNSIGNED8 cnt } ) — DATASET OF {wi, number, bucket, Min, and Max}, one for each N-tile group for each field.

## **HISTBINS** HistBins

FieldAggregates \

**HistBins** 

(Types.t\_Discrete n)

No Documentation Found

PARAMETER <u>n</u> ||| INTEGER4 — No Doc

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value , INTEGER4 hbin } ) —

# **HISTBINRANGES** HistBinRanges

FieldAggregates \

HistBinRanges

(Types.t\_Discrete n)

No Documentation Found

PARAMETER <u>n</u> ||| INTEGER4 — No Doc

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 number , INTEGER4 hbin , REAL8 Min , REAL8 Max , UNSIGNED8 cnt } ) —

## PEARSONCORR PearsonCorr

FieldAggregates \

PearsonCorr

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , INTEGER4 number1 , INTEGER4 number2 , REAL8 Correl } ) —

## **SPEARMANCORR** SpearmanCorr

FieldAggregates \

SpearmanCorr

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , INTEGER4 number1 , INTEGER4 number2 , REAL8 Correl } ) —

## KENDALLCORR KendallCorr

FieldAggregates \

KendallCorr

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , INTEGER4 number1 , INTEGER4 number2 , REAL8 Correl } ) —

## **GENSPEARMAN2CORR** GenSpearman2Corr

FieldAggregates \

	GenSpearman2Corr	
(	(dep = 1)	

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , INTEGER4 number1 , INTEGER4 number2 , REAL8 Correl } ) —

## **FromField**

Go Up

#### **DESCRIPTIONS**

#### **FROMFIELD** FromField

```
/ EXPORT FromField

(dIn,10ut,dOut,dMap=")
```

Macro to convert a NumericField formatted, cell-based dataset to a Record formatted dataset. Typically used to return converted NumericField data back to its original layout.

Note that as a Macro, nothing is returned, but new attributes are created in-line for use in subsequent definitions.

In the simplest case, the assumption is that the field order of the resulting table is in line with the field number in the input dataset, with the ID field as the first field.

For example:

```
myRec := RECORD
   UNSIGNED recordId;
   REAL height;
   REAL weight;
END;
Value of NumericField records with field number = 1 would go to height.
Value of NumericField records with field number = 2 would go to weight.
The id field of the NumericField record would be mapped to the recordId field of the result.
```

If the field orders have been changed (e.g. by customizing the ToField process, a field-mapping should be specified (See dMap below). Usage Examples:

PARAMETER <u>dIn</u> || INTEGER8 — The name of the input dataset in NumericField format.

PARAMETER <u>IOut</u> ||| INTEGER8 — The name of the layout record defining the records of the result dataset.

PARAMETER <u>dOut</u> || INTEGER8 — The name of the result dataset.

PARAMETER <u>dMap</u> || INTEGER8 — [OPTIONAL] A Field\_Mapping dataset as produced by ToField that describes the mapping between field name and field number. The format of this map is defined by Types.Field\_Mapping.

**RETURN** — Nothing. The MACRO creates new attributes in-line as described above.

**SEE** Types.NumericField

SEE Types.Field\_Mapping

SEE ToField

## Generate

Go Up

## **IMPORTS**

 $\_$ versions.ML $\_$ Core.V3 $\_$ 2 $\_$ 2.ML $\_$ Core.Types |

#### **DESCRIPTIONS**

#### **GENERATE** Generate

Generate

Increase dimensionality by adding polynomial transforms of the data to create new feature columns. This can be useful, for example, when building a linear model against data that may not have linear relationships.

#### Children

- 1. tp\_Method (From Generate): Enumeration of polynomial methods
- 2. MethodName (From Generate) : Convert a column number into a descriptive label
- 3. ToPoly (From Generate) : Generate up to seven, successively higher order, features from a single given feature

## TP\_METHOD tp\_Method

Generate \

#### $tp\_Method$

Enumeration of polynomial methods.

#### RETURN UNSIGNED1 —

**VALUE** Log X = 1

VALUE X = 2

**VALUE** XLog X = 3

**VALUE** XX = 4 - X squared

**VALUE** XXLogX = 5

**VALUE** XXX = 6 - X cubed

**VALUE** XXXLogX = 7

## **METHODNAME** MethodName

#### Generate \

#### MethodName

(tp\_Method x)

Convert a column number into a descriptive label.

**PARAMETER**  $\underline{\mathbf{x}}$  || UNSIGNED1 — The column number to describe.

**RETURN STRING7** — The descriptive label.

## **TOPOLY** ToPoly

#### Generate \

#### **ToPoly**

(DATASET(Types.NumericField) seedCol, UNSIGNED maxN=7)

Generate up to seven, successively higher order, features from a single given feature.

The generated features are:

- 1. LogX (logs are base 10)
- 2. X
- 3. XLogX
- 4. X<sup>2</sup>
- 5. X<sup>2</sup>LogX
- 6. X<sup>3</sup>
- 7. X<sup>3</sup>LogX

Note that the returned fields will be numbered 1-7, as above.

**PARAMETER** seedCol ||| TABLE ( NumericField ) — A single column of NumericField data. The number field is ignored.

**PARAMETER** maxN ||| UNSIGNED8 — (Optional) The number of new columns to generate. For example: If 1, then one feature, LogX is generated. If 3, then LogX, X, and X^2 features are generated. The default is 7, in which case, all features are generated.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — DATASET(NumericField) with numOriginalRecs \* maxN records.

SEE Types.NumericField

# ModelOps2

Go Up

#### **IMPORTS**

\_versions.ML\_Core.V3\_2\_2.ML\_Core.Types |

#### **DESCRIPTIONS**

## MODELOPS2 ModelOps2

ModelOps2

This module provides a set of operations to provide manipulation of machine learning models (version 2) in the Types.Layout\_Model2 format.

Layout\_Model2 defines a flexible structure that allows storage of model information for any Machine Learning algorithm.

The model is based on a "Naming Tree" paradigm.

The naming tree is a data structure that allows a hierarchical name (e.g. object-id) to be attached to each data-cell. Examples of naming-trees are OID trees such as those used in various network identifiers such as MIBs.

This structure is used within ML to store model information. It is a useful format for several reasons:

- It has the flexibility to store complex sets of data in a generic way.
- It easily stores scalar as well as matrix oriented data.
- It allows a model to contain data elements within scopes that are defined at different level. For example, part of the model may be defined globally, another may be common for a bundle, while another section is specific to a given module.

• It readily allows composite models to be created by encapsulating entire complex models (or sets of models) within branches of another model. The individual models can then be extracted from the composite model, and passed to the modules that created them.

#### Theory of Operation

The naming tree (NT) is conceptually simple. Each cell is identified by a hierarchical numbering scheme of arbitrary depth. Take, for example, the following NT:

```
1
1.1
1.1.1
1.1.2
1.2
1.2.1
1.2.2
```

This tree defines the following leaf (scalar) elements: 1.1.1, 1.1.2, 1.2.1, 1.2.2, 2.

Note that the deepest node on any branch is considered a leaf, and branches can be of variable depth. Note also that there is no explicit creation of branch nodes. The branches are implicitly defined by the ids of the leafs.

In this example, node 1.1 can be thought as representing an array, thought it could also be thought of as a structure of two distinct scalars, depending on whether the user expects a variable length list under 1.1 (i.e. 1.1.1 - 1.1.N) or a fixed set of cells.

Likewise node 1 can be thought of as a matrix (1.r.c, where r is the row index and c is the column index), in cases where r and c are of variable size.

This naming tree also supports the myriad interface, allowing multiple independent work-items to be represented, each of which may duplicate the same structure.

The id is represented by an ECL SET of Unsigned identifiers (e.g. [1,2,1] represents the OID 1.2.1).

Each cell is defined by three fields: wi (work-item-id), value (the cell contents) and indexes (the id).

A naming tree can be constructed as an inline dataset. For example, the following creates the tree in the example above:

```
DATASET([\{1, 3.2, [1,1,1]\}, \{1, .0297, [1,1,2]\}, \{1, 2.0, [1,2,1]\}, \{1, 1550, [1,2,2]\},
```

```
\{1, 8.1, [2]\}\, Layout\_Model2);
```

There are attributes in this module to assist with manipulation of naming trees:

- Creating a NT from a NumericField matrix.
- Extracting a NumericField matrix from an NT branch.
- Inserting an NT onto a branch of another NT.
- Extracting an NT from a branch of an NT.

#### SEE Types.Layout\_Model2

#### Children

- 1. Extract (From ModelOps2): Extract an inner sub-tree from an existing model
- 2. ExtendIndices (From ModelOps2): Extend the indices of a model to fit within a deeper model
- 3. Insert (From ModelOps2): Insert a model into a sub-tree of an existing model
- 4. ToNumericField (From ModelOps2): Convert a two-level model or model sub-tree into a NumericField dataset
- 5. FromNumericField (From ModelOps2): Convert a NumericField dataset to a 2 level model (or model subtree)
- 6. GetItem (From ModelOps2): Get a single record (cell) from a model by index
- 7. SetItem (From ModelOps2): Add a single record (cell) to an model at a given set of coordinates

#### **EXTRACT** Extract

#### ModelOps2 \

```
DATASET(Layout_Model2) Extract

(DATASET(Layout_Model2) mod, t_indexes fromIndx, t_work_item fromWi=0)
```

Extract an inner sub-tree from an existing model.

Work-item = 0 (default) will extract all work-items

This is the opposite of Insert. For example:

```
If I have a tree:
    1
    2
    3
    3.1
    3.2
```

and I extract from index 3, it will return the Naming Tree:

1

containing the two sub-cells of the original index 3

```
PARAMETER <u>mod</u> ||| TABLE ( Layout_Model2 ) — The model from which to extract the sub-tree.
```

PARAMETER <u>fromIndx</u> ||| SET ( UNSIGNED4 ) — The index from which to extract the subtree.

**PARAMETER** from Wi || UNSIGNED2 — The work-item to extract or 0 to extract the same sub-tree from all work-items.

RETURN TABLE ( { UNSIGNED2 wi , REAL8 value , SET ( UNSIGNED4 ) indexes } )
— A model containing all of the sub-cells below fromIndx with the indexes adjusted to the top of the tree.

## **EXTENDINDICES** ExtendIndices

 $ModelOps2 \ \backslash$ 

```
DATASET(Layout_Model2) ExtendIndices

(DATASET(Layout_Model2) mod, t_indexes atIndex)
```

Extend the indices of a model to fit within a deeper model.

For example, a cell with index [1,2] could be moved to index [1,2,3,1,2] by using at Index := [1,2,3].

PARAMETER <u>mod</u> ||| TABLE ( Layout\_Model2 ) — The model whose indexes are to be extended.

**PARAMETER** <u>atIndex</u> ||| SET ( UNSIGNED4 ) — The prefix indexes to be prepended to the indexes of each cell in mod.

RETURN TABLE ( { UNSIGNED2 wi , REAL8 value , SET ( UNSIGNED4 ) indexes } ) — A model with extended indexes.

#### **INSERT** Insert

#### $ModelOps2 \setminus$

Insert a model into a sub-tree of an existing model.

Extends the indexes of the provided model to fit onto a branch of another model, and concatenates the two models. This is the opposite of extract. For example:

```
If I have a model:

1
2
and a second model:

1
2
3
That I would like to insert into the first tree at index 3, I would end up with the tree:

1
2
3
3.1
3.1
3.2
3.3
```

Example code:

```
mod3 := Insert(mod1, mod2, [3]);
```

```
PARAMETER mod1 || TABLE (Layout_Model2) — The first (base) model.
```

**PARAMETER** mod2 ||| TABLE ( Layout\_Model2 ) — The sub-model that is to be inserted into mod1.

**PARAMETER** at Indx ||| SET (UNSIGNED4) — The index prefix (in mod1) that will contain the cells from mod2.

**RETURN** TABLE (Layout\_Model2) — a new model containing the cells from both models.

#### **TONUMERICFIELD ToNumericField**

ModelOps2 \

```
DATASET(NumericField) ToNumericField

(DATASET(Layout_Model2) mod, t_indexes fromIndx = [])
```

Convert a two-level model or model sub-tree into a NumericField dataset.

The last two indexes of the model subtree are used as the indexes for the NumericField matrix. The second to last index corresponds to the NF's id field and the last index corresponds to the NF's number field.

PARAMETER <u>mod</u> ||| TABLE ( Layout\_Model2 ) — The model from which to extract the NumericField matrix.

**PARAMETER** from Indx ||| SET (UNSIGNED4) — The index from which to extract the matrix. Example: [3,1,5]. The default is from the top of the tree i.e. [].

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — NumericField matrix in DATASET(NumericField) format.

#### FROMNUMERICFIELD FromNumericField

#### ModelOps2 \

DATASET(Layout_Model2)	FromNumericField
(DATASET(NumericField) nf, t_indexes atIndex=[])	

Convert a NumericField dataset to a 2 level model (or model subtree).

A two level model is created and appended to atIndex.

The first new index will contain the value of the NumericField's id field, and the second will contain the value of the NumericField's number field.

Example: If I have a NumericField with id=1 and number=3, and I use atIndex = [3,1,5], it will create a Naming Tree cell with indexes: [3,1,5,1,3].

**PARAMETER** <u>nf</u> || TABLE ( NumericField ) — A NumericField dataset to be converted.

**PARAMETER** at Index | | SET (UNSIGNED4) — The index at which to place the new subtree e.g., [3,1,5].

RETURN TABLE ( { UNSIGNED2 wi , REAL8 value , SET ( UNSIGNED4 ) indexes } ) — DATASET(ntNumeric) Naming Tree.

#### **GETITEM** GetItem

#### $ModelOps2 \setminus$

```
Layout_Model2 GetItem

(DATASET(Layout_Model2) mod, t_indexes indxs, wi_num=1)
```

Get a single record (cell) from a model by index.

**PARAMETER** mod ||| TABLE ( Layout\_Model2 ) — The model (DATASET(layout\_model2)) from which to extract the cell.

**PARAMETER** indxs ||| SET (UNSIGNED4) — The id of the cell to extract (e.g. [3,1,5]).

PARAMETER wi\_num || INTEGER8 — The work-item number to extract the cell from, default = 1.

**RETURN** ROW ( Layout\_Model2 ) — The model cell (Layout\_Model2) or an empty cell (wi=0) if not found.

## **SETITEM SetItem**

#### ModelOps2 \

```
DATASET(Layout_Model2) SetItem

(DATASET(Layout_Model2) mod, t_work_item wi, t_indexes indexes, t_fieldReal value)
```

Add a single record (cell) to an model at a given set of coordinates.

PARAMETER mod | TABLE (Layout\_Model2) — The model to which to add a cell.

PARAMETER wi || UNSIGNED2 — The work-item associated with the cell.

PARAMETER indexes || SET ( UNSIGNED4 ) — The indices for the cell.

PARAMETER value ||| REAL8 — The value of the cell.

**RETURN** TABLE ( Layout\_Model2 ) — Model with the added cell.

## **ToField**

Go Up

#### **DESCRIPTIONS**

#### TOFIELD ToField

```
/ EXPORT ToField

(dIn, dOut, idfield=", wifield=", wivalue=", datafields=")
```

Convert a record-oriented dataset to a cell-oriented NumericField dataset for use with Machine Learning mechanisms.

To Field Macro takes a record-oriented dataset, with each row containing an ID and one or more numeric fields, and expands it into the Numeric Field format used by ML.

Note that as a Macro, nothing is returned, but new attributes are created in-line for use in subsequent definitions.

Along with creating the NumericField table, this macro produces two simple functions to assist the user in mapping the field names to their corresponding numbers. These are "STRING dOut\_ToName(UNSIGNED)" and "UNSIGNED dOut\_ToNumber(STRING)", where the "dOut" portion of the function name is the name passed into that parameter of the macro.

The macro also produces a mapping table named "dOut\_Map", again where "dOut" refers to the parameter, that contains a table of the field mappings. See Types.Field\_Mapping for the layout of this mapping dataset. Examples:

- PARAMETER dIn || INTEGER8 The name of the input dataset.
- PARAMETER <u>dOut</u> || INTEGER8 The name of the resulting dataset.
- **PARAMETER** <u>idfield</u> ||| INTEGER8 [OPTIONAL] The name of the field that contains the UID for each row. If omitted, it is assumed to be the first field.
- **PARAMETER** wifield ||| INTEGER8 [OPTIONAL] The name of the field that contains the work item value. A constant is used if the field name is not supplied (as provided by wivalue below).
- **PARAMETER** wivalue ||| INTEGER8 [OPTIONAL] The constant value to use for work item. The value 1 is used if not supplied.
- PARAMETER <u>datafields</u> ||| INTEGER8 [OPTIONAL] A STRING containing a comma-delimited list of the fields to be treated as axes. If omitted, all numeric fields that are not the idfield or wifield will be treated as axes. NOTE: idfield defaults to the first field in the table, so if that field is specified as an axis field, then the user should be sure to specify a value in the idfield param.
- **RETURN** Nothing. The MACRO creates new attributes in-line as described above.
- **SEE** Types.NumericField
- SEE Types.Field\_Mapping

## **Types**

Go Up

### **DESCRIPTIONS**

## **TYPES** Types

**Types** 

This module provides the major data type definitions for use with the various ML Bundles

#### Children

- 1. t\_RecordID (From Types) : No Documentation Found
- 2. t\_FieldNumber (From Types) : No Documentation Found
- 3. t\_FieldReal (From Types): No Documentation Found
- 4. t\_FieldSign (From Types) : No Documentation Found
- 5. t\_Discrete (From Types) : No Documentation Found
- 6. t\_Item (From Types) : No Documentation Found
- 7. t\_Count (From Types) : No Documentation Found
- 8. t\_Work\_Item (From Types) : No Documentation Found
- 9. t\_index (From Types) : No Documentation Found
- 10. t\_indexes (From Types) : No Documentation Found
- 11. AnyField (From Types): No Documentation Found
- 12. NumericField (From Types): The NumericField layout defines a matrix of Real valued data-points
- 13. DiscreteField (From Types) : The Discrete Field layout defines a matrix of Integer valued data-points
- 14. Layout\_Model2 (From Types): Layout for Model dataset (version 2) Generic Layout describing the model 'learned' by a Machine Learning algorithm

- 15. Layout\_Model (From Types): No Documentation Found
- 16. Classify Result (From Types): No Documentation Found
- 17. l\_result (From Types): No Documentation Found
- 18. Class\_Stats (From Types) : Class\_Stats
- 19. Confusion\_Detail (From Types): Confusion\_Detail
- 20. Classification Accuracy (From Types): Classification Accuracy
- 21. Class\_Accuracy (From Types): Class\_Accuracy Results layout for Analysis.Classification.AccuracyByClass See https://en.wikipedia.org/wiki/Precision\_and\_recall for a more detailed explanation
- 22. AUC\_Result (From Types): AUC\_Result Result layout for Analysis.Classification.AUC
- 23. Regression\_Accuracy (From Types): Regression\_Accuracy
- 24. Contingency\_Table (From Types): Contingency\_Table Contains the contingency table for every combination of feature and classifier
- 25. Chi2\_Result (From Types): Chi2\_Result Result layout for Analysis.FeatureSelection.Chi2 Contains chi2 value for every combination of feature and classifier per work item, and its corresponding p value
- 26. ARI\_Result (From Types): ARI\_Result Result layout for Analysis.Clustering.ARI Contains the Adjusted Rand Index for each work item
- 27. SampleSilhouette\_Result (From Types): SampleSilhouette\_Result Result layout for Analysis.Clustering.SampleSilhouetteScore Contains the silhouette score for each sample datapoint
- 28. Silhouette\_Result (From Types): Silhouette\_Result Result layout for Analysis.Clustering.SilhouetteScore Contains the silhouette score for each work item
- 29. ClusterLabels (From Types): ClusterLabels format defines the distance space where each cluster defined by a center and its closest samples
- 30. Data\_Diagnostic (From Types): No Documentation Found
- 31. Field\_Mapping (From Types): Field\_Mapping is the format produced by ToField for field-name mapping
- 32. LUCI\_Rec (From Types): LUCI Record A dataset of lines each containing a string This is the DATASET format in which ML algorithm export LUCI files
- 33. Classification\_Scores (From Types): Classification\_Scores The probability or confidence, per class, that a sample belongs to that class

# T\_RECORDID t\_RecordID Types \ t\_RecordID No Documentation Found RETURN UNSIGNED8 — T\_FIELDNUMBER t\_FieldNumber Types \ t FieldNumber No Documentation Found RETURN UNSIGNED4 — T\_FIELDREAL t\_FieldReal Types \ $t_FieldReal$ No Documentation Found

RETURN REAL8 —

# T\_FIELDSIGN t\_FieldSign Types \ $t_FieldSign$ No Documentation Found RETURN INTEGER1 — T\_DISCRETE t\_Discrete Types \ t Discrete No Documentation Found RETURN INTEGER4 — T\_ITEM t\_Item Types \ $t_{\perp}$ Item No Documentation Found

RETURN UNSIGNED4 —

# T\_COUNT t\_Count Types \ t\_Count No Documentation Found RETURN UNSIGNED8 — T\_WORK\_ITEM t\_Work\_Item Types \ $t_Work_Item$ No Documentation Found RETURN UNSIGNED2 — T\_INDEX t\_index Types \ $t_{index}$ No Documentation Found RETURN UNSIGNED4 —

## T\_INDEXES t\_indexes

Types \

t indexes

No Documentation Found

RETURN SET ( UNSIGNED4 ) -

## **ANYFIELD** AnyField

Types \

AnyField

No Documentation Found

FIELD wi || UNSIGNED2 — No Doc

**FIELD** <u>id</u> ||| UNSIGNED8 — No Doc

**FIELD** <u>number</u> ||| UNSIGNED4 — No Doc

### **NUMERICFIELD NumericField**

Types \

NumericField

The NumericField layout defines a matrix of Real valued data-points. It acts as the primary Dataset layout for interacting with most ML Functions. Each record represents a single cell in a matrix. It is most often used to represent a set of data-samples or observations, with the 'id' field representing the data-sample or observation, and the 'number' field representing the various fields within the observation.

- FIELD <u>wi</u> ||| UNSIGNED2 The work-item id, supporting the Myriad style interface. This allows multiple independent matrixes to be contained within a single dataset, supporting independent ML activities to be processed in parallel.
- **FIELD** <u>id</u> ||| UNSIGNED8 This field represents the row-number of this cell of the matrix. It is also considered the record-id for observations / data-samples.
- **FIELD** <u>number</u> ||| UNSIGNED4 This field represents the matrix column number for this cell. It is also considered the field number of the observation
- FIELD <u>value</u> ||| REAL8 The value of this cell in the matrix.

#### **DISCRETEFIELD** DiscreteField

Types \

#### DiscreteField

The Discrete Field layout defines a matrix of Integer valued data-points. It is similar to the NumericField layout above, except for only containing discrete (integer) values. It is typically used to convey the class-labels for classification algorithms.

- **FIELD** <u>wi</u> || UNSIGNED2 The work-item id, supporting the Myriad style interface. This allows multiple independent matrixes to be contained within a single dataset, supporting independent ML activities to be processed in parallel.
- **FIELD** <u>id</u> ||| UNSIGNED8 This field represents the row-number of this cell of the matrix. It is also considered the record-id for observations / data-samples.
- **FIELD** <u>number</u> ||| UNSIGNED4 This field represents the matrix column number for this cell. It is also considered the field number of the observation
- FIELD value || INTEGER4 The value of this cell in the matrix.

#### LAYOUT\_MODEL2 Layout\_Model2

Types \

Layout\_Model2

Layout for Model dataset (version 2) Generic Layout describing the model 'learned' by a Machine Learning algorithm. Models for all new ML bundles are stored in this format. Some older bundles may still use the Layout\_Model (version 1) layout. Models are thought of as opaque data structures. They are not designed to be understandable except to the bundle that produced them. Most bundles contain mechanisms to extract useful information from the model. This version of the model is based on a Naming-Tree paradigm. This provides a flexible generic mechanism for storage and manipulation of models. For bundle developers (or the curious), the file modelOps2 provides a detailed description of the theory and usage of this model layout as well as a set of functions to manipulate models for use by bundle developers.

FIELD wi || UNSIGNED2 — The work-item-id

FIELD <u>value</u> ||| REAL8 — The value of the cell

**FIELD** indexes ||| SET (UNSIGNED4) — The identifier for the cell – a set of unsigned integers e.g., [1,2,1,3]

## LAYOUT\_MODEL Layout\_Model

Types \

 $Layout\_Model$ 

No Documentation Found

FIELD wi || UNSIGNED2 — No Doc

FIELD id || UNSIGNED8 — No Doc

**FIELD** <u>number</u> ||| UNSIGNED4 — No Doc

FIELD <u>value</u> ||| REAL8 — No Doc

## CLASSIFY\_RESULT Classify\_Result

Types \

Classify\_Result

#### No Documentation Found

FIELD wi || UNSIGNED2 — No Doc

FIELD id || UNSIGNED8 — No Doc

FIELD <u>number</u> ||| UNSIGNED4 — No Doc

FIELD value || INTEGER4 — No Doc

FIELD conf ||| REAL8 — No Doc

## L\_RESULT | \_result

Types \

l result

No Documentation Found

## **CLASS\_STATS** Class\_Stats

Types \

Class Stats

Class\_Stats Layout for data returned from Analysis.Regression.ClassStats

FIELD <u>wi</u> || UNSIGNED2 — Work-item identifier

**FIELD** <u>classifier</u> || UNSIGNED4 — The field number associated with this dependent variable, for multi-variate classification. Otherwise 1.

FIELD <u>class</u> ||| INTEGER4 — The class label associated with this record

FIELD <u>classCount</u> ||| INTEGER4 — The number of times the class was seen in the data

**FIELD** <u>classPct</u> ||| REAL8 — The percent of records with this class.

## CONFUSION\_DETAIL Confusion\_Detail

#### Types \

#### Confusion Detail

Confusion\_Detail Layout for storage of the confusion matrix for ML Classifiers Each row represents a pairing of a predicted class and an actual class

- FIELD wi || UNSIGNED2 Work item identifier
- **FIELD** <u>classifier</u> ||| UNSIGNED4 The field number associated with this dependent variable, for multi-variate. Otherwise 1.
- **FIELD** actual\_class ||| INTEGER4 The target class number the expected result.
- FIELD predict\_class ||| INTEGER4 The class number predicted by the ML algorithm
- **FIELD** <u>occurs</u> || UNSIGNED4 The number of times this pairing of (actual / predicted) classes occurred
- **FIELD** <u>correct</u> ||| BOOLEAN Boolean indicating if this represents a correct prediction (i.e. predicted = actual)
- **FIELD** <u>pctActual</u> ||| REAL8 The percent of items that were actually of <actual\_class> that were predicted as cpredict\_class>.</predict\_class></actual\_class></predict\_class>

## CLASSIFICATION\_ACCURACY Classification\_Accuracy

#### Types \

#### Classification\_Accuracy

Classification\_Accuracy Results layout for Analysis.Classification/Accuracy

- FIELD wi || UNSIGNED2 Work item identifier
- **FIELD** <u>classifier</u> || UNSIGNED4 The field number associated with this dependent variable, for multi-variate. Otherwise 1.

- FIELD <u>errCnt</u> || UNSIGNED8 The number of errors (i.e. predicted &lt;&gt; actual)
- FIELD recCnt || UNSIGNED8 The total number or records in the test set
- FIELD Raw\_Accuracy ||| REAL8 The percentage of samples properly classified (0.0 1.0)
- FIELD PoD ||| REAL8 Power of Discrimination. Indicates how this classification performed relative to a random guess of class. Zero or negative indicates that the classification was no better than a random guess. 1.0 indicates a perfect classification. For example if there are two equi-probable classes, then a random guess would be right about 50% of the time. If this classification had a Raw Accuracy of 75%, then its PoD would be .5 (half way between a random guess and perfection).
- PoDE ||| REAL8 Power of Discrimination Extended. Indicates how this classification performed relative to guessing the most frequent class (i.e. the trivial solution). Zero or negative indicates that this classification is no better than the trivial solution. 1.0 indicates perfect classification. For example, if 95% of the samples were of class 1, then the trivial solution would be right 95% of the time. If this classification had a raw accuracy of 97.5%, its PoDE would be .5 (i.e. half way between trivial solution and perfection).
- FIELD <u>Hamming\_Loss</u> ||| REAL8 Hamming loss. The percentage of records misclassified. Useful for multilabel classification. It is equal to 1 Raw\_Accuracy.

## CLASS\_ACCURACY Class\_Accuracy

Types \

#### Class\_Accuracy

Class\_Accuracy Results layout for Analysis.Classification.AccuracyByClass See https://en.wikipedia.org/wiki/Precision and recall for a more detailed explanation.

- FIELD wi || UNSIGNED2 Work item identifier
- **FIELD** <u>classifier</u> ||| UNSIGNED4 The field number associated with this dependent variable, for multi-variate. Otherwise 1.
- FIELD <u>class</u> ||| INTEGER4 The class to which the analytics apply
- **FIELD precision** ||| REAL8 The precision of the classification for this class (i.e. True Positives / (True Positives + FalsePositives)). What percentage of the items that we predicted as being in this class are actually of this class?
- **FIELD** <u>recall</u> ||| REAL8 The completeness of recall for this class (i.e. True Positives / (True Positives + False Negatives)) What percentage of the items that are actually in this class did we correctly predict as this class?

- **FIELD FPR** ||| REAL8 The false positive rate for this class (i.e. False Positives / (False Positives + True Negatives)) What percentage of the items not in this class did we falsely predict as this class?
- **FIELD** <u>f\_score</u> ||| REAL8 The balanced F-score for this class (i.e. 2 \* (precision \* recall) / (precision + recall)) The harmonic mean of precision and recall. Higher values are better.

## AUC\_RESULT AUC\_Result

Types \

AUC Result

AUC\_Result Result layout for Analysis.Classification.AUC. Provides the area under the Receiver Operating Characteristic curve for the given given data. This area is a measure of the classifier's ability to distinguish between classes.

- FIELD wi || UNSIGNED2 Work item identifier
- **FIELD** <u>classifier</u> ||| UNSIGNED4 The field number associated with this dependent variable, for multi-variate. Otherwise 1.
- FIELD <u>class</u> ||| INTEGER4 The class to which the analytics apply.
- **FIELD** <u>AUC</u> ||| REAL8 The value of the Area Under the Receiver Operating Characteristic curve for this class. This value ranges between 0 and 1. A higher value is an indication of a better classifier.

#### REGRESSION\_ACCURACY Regression\_Accuracy

Types \

Regression\_Accuracy

Regression Accuracy Results layout for Analysis.Regression.Accuracy

FIELD <u>wi</u> || UNSIGNED2 — Work item identifier

- FIELD <u>regressor</u> ||| UNSIGNED4 The field number associated with this dependent variable, for multi-variate. Otherwise 1.
- FIELD <u>R2</u> ||| REAL8 The R-Squared value (Coefficient of Determination) for the regression. R-squared of zero or negative indicates that the regression has no predictive value. R2 of 1 would indicate a perfect regression.
- **FIELD** <u>MSE</u> ||| REAL8 Mean Squared Error = SUM((predicted actual)^2) / N (number of datapoints)
- **FIELD RMSE** ||| REAL8 Root Mean Squared Error = MSE^.5 (Square root of MSE)

#### CONTINGENCY\_TABLE Contingency\_Table

Types \

Contingency\_Table

Contingency\_Table Contains the contingency table for every combination of feature and classifier. Result layout for Analysis.FeatureSelection.Contingency

- FIELD wi || UNSIGNED2 Work item identifier
- **FIELD** <u>fnumber</u> ||| UNSIGNED4 The feature number
- FIELD <u>snumber</u> || UNSIGNED4 The sample number or the classifier number
- FIELD <u>fclass</u> ||| INTEGER4 The feature label / class
- FIELD sclass || INTEGER4 The sample (classifier) label / class
- **FIELD** <u>cnt</u> || INTEGER8 The number of samples with feature label fclass and classifier label sclass Does not contain entries for combinations with no members.

## CHI2\_RESULT Chi2\_Result

Types \

Chi2\_Result

Chi2\_Result Result layout for Analysis.FeatureSelection.Chi2 Contains chi2 value for every combination of feature and classifier per work item, and its corresponding p value.

- FIELD wi || UNSIGNED2 Work item identifier
- FIELD <u>fnumber</u> || UNSIGNED4 Feature number
- FIELD snumber || UNSIGNED4 Sample number / number of classifier
- FIELD <u>dof</u> || INTEGER8 The number of degrees of freedom
- FIELD <u>x2</u> ||| REAL8 The chi2 value for this combination. Higher values indicate more closely related variables
- FIELD  $\underline{\mathbf{p}}$  ||| REAL8 The p-value, which is the area under the chi-square probability density function curve to the right of the specified x2 value. The probability that the variables are not closely related

## ARI\_RESULT ARI\_Result

Types \

#### ARI\_Result

ARI\_Result Result layout for Analysis.Clustering.ARI Contains the Adjusted Rand Index for each work item.

- FIELD wi || UNSIGNED2 Work item identifier
- FIELD value ||| REAL8 The ARI for the model

## SAMPLESILHOUETTE\_RESULT SampleSilhouette\_Result

Types \

 $Sample Silhouette \_Result$ 

SampleSilhouette\_Result Result layout for Analysis.Clustering.SampleSilhouetteScore Contains the silhouette score for each sample datapoint.

- FIELD wi || UNSIGNED2 Work item identifier
- FIELD id || UNSIGNED8 Sample datapoint identifier
- FIELD <u>value</u> ||| REAL8 Silhouette score

#### SILHOUETTE\_RESULT Silhouette\_Result

Types \

Silhouette Result

Silhouette\_Result Result layout for Analysis.Clustering.SilhouetteScore Contains the silhouette score for each work item.

- FIELD wi || UNSIGNED2 Work item identifier
- FIELD score ||| REAL8 Silhouette score

#### **CLUSTERLABELS** ClusterLabels

Types \

ClusterLabels

ClusterLabels format defines the distance space where each cluster defined by a center and its closest samples. It is the same as KMeans.Types.KMeans\_Model.Labels.

- FIELD <u>wi</u> || UNSIGNED2 The model identifier.
- **FIELD** <u>id</u> || UNSIGNED8 The sample identifier.
- **FIELD** <u>label</u> || UNSIGNED8 The identifier of the closest center to the sample.

## DATA\_DIAGNOSTIC Data\_Diagnostic

Types \

Data\_Diagnostic

No Documentation Found

- FIELD wi || UNSIGNED2 No Doc
- FIELD <u>valid</u> ||| BOOLEAN No Doc
- FIELD message\_text ||| SET ( VARSTRING ) No Doc

## FIELD\_MAPPING Field\_Mapping

Types \

Field\_Mapping

Field\_Mapping is the format produced by ToField for field-name mapping.

- FIELD orig\_name || STRING The name of the field in the original layout
- **FIELD** <u>assigned\_name</u> ||| STRING The integer field number used in the ML algorithm stored as a STRING

## LUCI\_REC LUCI\_Rec

Types  $\setminus$ 

 ${\bf LUCI\_Rec}$ 

LUCI Record – A dataset of lines each containing a string This is the DATASET format in which ML algorithm export LUCI files.

## CLASSIFICATION\_SCORES Classification\_Scores

#### Types \

#### Classification\_Scores

Classification\_Scores The probability or confidence, per class, that a sample belongs to that class.

- FIELD <u>wi</u> || UNSIGNED2 The work-item identifier.
- **FIELD** <u>id</u> ||| UNSIGNED8 The record-id of the sample.
- **FIELD** <u>classifier</u> ||| UNSIGNED4 The field number associated with this dependent variable, for multi-variate. Otherwise 1.
- FIELD <u>class</u> ||| INTEGER4 The class label.
- **FIELD** <u>prob</u> ||| REAL8 The percentage of trees that assigned this class label, which is a rough stand-in for the probability that the label is correct.

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IRegression2.ecl

Interface Definition for Regression Modules (Version 2)

# Interfaces/ IClassify

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## **ICLASSIFY** IClassify

**IClassify** 

\*\*\*DEPRECATED\*\*\* Interface Definition for Classification Modules (version 1). This interface is being deprecated and should not be used for new bundles or bundles undergoing substantial revision. Please use IClassify2 going forward. Interface definition for Classification. Actual implementation modules will probably take parameters.

#### Children

- 1. GetModel (From IClassify): Calculate the model to fit the observation data to the observed classes
- 2. Classify (From IClassify): Classify the observations using a model
- 3. Report (From IClassify): Report the confusion matrix for the classifier and training data

### **GETMODEL** GetModel

### IClassify \

DATASET(Types.Layout_Model)	GetModel
(DATASET(Types.NumericField) DATASET(Types.DiscreteField)	

Calculate the model to fit the observation data to the observed classes.

```
PARAMETER observations || TABLE (NumericField) — the observed explanatory values.
```

**PARAMETER** <u>classifications</u> ||| TABLE ( DiscreteField ) — the observed classification used to build. the model

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — the encoded model

### **CLASSIFY** Classify

### IClassify \

```
DATASET(Types.Classify_Result) Classify

(DATASET(Types.Layout_Model) model,
DATASET(Types.NumericField) new_observations)
```

Classify the observations using a model.

**PARAMETER** <u>model</u> ||| TABLE ( Layout\_Model ) — The model, which must be produced by a corresponding getModel function.

**PARAMETER** new\_observations ||| TABLE ( NumericField ) — observations to be classified.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , INTEGER4 value , REAL8 conf } ) — Classification with a confidence value.

## **REPORT** Report

### IClassify \

```
DATASET(Types.Confusion_Detail) Report

(DATASET(Types.Layout_Model) model,
DATASET(Types.NumericField) observations,
DATASET(Types.DiscreteField) classifications)
```

Report the confusion matrix for the classifier and training data.

```
PARAMETER <u>model</u> ||| TABLE ( Layout_Model ) — the encoded model.
```

PARAMETER observations || TABLE ( NumericField ) — the explanatory values.

**PARAMETER** <u>classifications</u> ||| TABLE ( DiscreteField ) — the classifications associated with the observations.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , INTEGER4 actual\_class , INTEGER4 predict\_class , UNSIGNED4 occurs , BOOLEAN correct , REAL8 pctActual , REAL8 pctPred } ) — the confusion matrix showing correct and incorrect results.

## ${\bf Interfaces/}\\ {\bf IClassify2}$

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## **IMPORTS**

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### **DESCRIPTIONS**

### **ICLASSIFY2** IClassify2

IClassify2

Interface definition for Classification (Version 2). Classification learns a function that maps a set of input data to one or more output class-label (i.e. Discrete) variables. The resulting learned function is known as the model. That model can then be used repetitively to predict the class(es) for each sample when presented with new input data. Actual implementation modules will probably take configuration parameters to control the classification process. The Classification modules also expose attributes for assessing the effectiveness of the classification.

#### Children

- 1. GetModel (From IClassify2) : Calculate the model to fit the independent data to the observed classes (i.e
- 2. Classify (From IClassify2): Classify the observations using a model
- 3. Accuracy (From IClassify2): Return accuracy metrics for the given set of test data

  This is equivalent to calling Predict followed by Analysis. Classification. Accuracy(...)
- 4. AccuracyByClass (From IClassify2): Return class-level accuracy by class metrics for the given set of test data

5. ConfusionMatrix (From IClassify2): Return the confusion matrix for a set of test data

### **GETMODEL** GetModel

### IClassify2 \

DATASET(Layout_Model2)	$\operatorname{GetModel}$	
(DATASET(NumericField) dependents)	independents, DATASET(DiscreteField)	

Calculate the model to fit the independent data to the observed classes (i.e. dependent data).

PARAMETER dependents || TABLE (DiscreteField) — The observed dependent (class label) values.

PARAMETER independents ||| TABLE ( NumericField ) — No Doc

RETURN TABLE ( { UNSIGNED2 wi , REAL8 value , SET ( UNSIGNED4 ) indexes } )
— The encoded model.

SEE Types.Layout\_Model2

SEE Types.NumericField

SEE Types.DiscreteField

## **CLASSIFY** Classify

### IClassify2 \

DATASET(DiscreteField)	Classify
(DATASET(Layout_Model2) observations)	model, DATASET(NumericField)

Classify the observations using a model.

**PARAMETER** <u>model</u> ||| TABLE ( Layout\_Model2 ) — The model, which must be produced by a corresponding getModel function.

**PARAMETER** <u>observations</u> ||| TABLE ( NumericField ) — New observations (independent data) to be classified.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , INTEGER4 value } ) — Predicted class values.

### **ACCURACY Accuracy**

### IClassify2 \

```
DATASET(Classification_Accuracy) Accuracy

(DATASET(Layout_Model2) model,
DATASET(DiscreteField) actuals,
DATASET(NumericField) observations)
```

Return accuracy metrics for the given set of test data

This is equivalent to calling Predict followed by Analysis. Classification. Accuracy(...).

Provides accuracy statistics as follows:

- errCount The number of misclassified samples.
- errPct The percentage of samples that were misclasified (0.0 1.0).
- RawAccuracy The percentage of samples properly classified (0.0 1.0).
- PoD Power of Discrimination. Indicates how this classification performed relative to a random guess of class. Zero or negative indicates that the classification was no better than a random guess. 1.0 indicates a perfect classification. For example if there are two equiprobable classes, then a random guess would be right about 50% of the time. If this classification had a Raw Accuracy of 75%, then its PoD would be .5 (half way between a random guess and perfection).
- PoDE Power of Discrimination Extended. Indicates how this classification performed relative to guessing the most frequent class (i.e. the trivial solution). Zero or negative indicates that this classification is no better than the trivial solution. 1.0 indicates perfect classification. For example, if 95% of the samples were of class 1, then the trivial solution would be right 95% of the time. If this classification had a raw accuracy of 97.5%, its PoDE would be .5 (i.e. half way between trivial solution and perfection).

Normally, this should be called using data samples that were not included in the training set. In that case, these statistics are considered Out-of-Sample error statistics. If it is called with the X

and Y from the training set, it provides In-Sample error statistics, which should never be used to rate the classification model.

- **PARAMETER** <u>model</u> ||| TABLE ( Layout\_Model2 ) The encoded model as returned from GetModel.
- **PARAMETER** <u>actuals</u> ||| TABLE ( DiscreteField ) The actual class values associated with the observations.
- **PARAMETER** observations ||| TABLE (NumericField) The independent (explanatory) values on which to base the test.
- RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , UNSIGNED8 recCnt , UNSIGNED8 errCnt , REAL8 Raw\_Accuracy , REAL8 PoD , REAL8 PoDE , REAL8 Hamming\_Loss } ) DATSET(Classification\_Accuracy), one record per work-item.
- SEE Types.Classification\_Accuracy

### **ACCURACYBYCLASS** AccuracyByClass

IClassify2 \

```
DATASET(Class_Accuracy) AccuracyByClass

(DATASET(Layout_Model2) model, DATASET(DiscreteField)
actuals, DATASET(NumericField) observations)
```

Return class-level accuracy by class metrics for the given set of test data.

This is equivalent to calling Predict followed by Analysis. Classification. Accuracy By Class (...).

- **PARAMETER** model ||| TABLE (Layout\_Model2) The encoded model as returned from GetModel.
- **PARAMETER** <u>actuals</u> ||| TABLE ( DiscreteField ) The actual class values associated with the observations.
- **PARAMETER** observations ||| TABLE ( NumericField ) The independent (explanatory) values on which to base the test
- RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , INTEGER4 class , REAL8 precision , REAL8 recall , REAL8 FPR , REAL8 f\_score } ) DATASET(Class\_Accuracy), one record per work-item per class.

### **CONFUSIONMATRIX** ConfusionMatrix

### IClassify2 \

```
DATASET(Confusion_Detail) ConfusionMatrix

(DATASET(Layout_Model2) model, DATASET(DiscreteField)
actuals, DATASET(NumericField) observations)
```

Return the confusion matrix for a set of test data. This is equivalent to calling Predict follwed by Analysis.Classification.ConfusionMatrix(...).

The confusion matrix indicates the number of datapoints that were classified correctly or incorrectly for each class label.

The matrix is provided as a matrix of size numClasses x numClasses with fields as follows:

- 'wi' The work item id
- 'pred' the predicted class label (from Classify).
- 'actual' the actual (target) class label.
- 'samples' the count of samples that were predicted as 'pred', but should have been 'actual'.
- 'totSamples' the total number of samples that were predicted as 'pred'.
- 'pctSamples' the percentage of all samples that were predicted as 'pred', that should have been 'actual' (i.e. samples / totSamples)

This is a useful tool for understanding how the algorithm achieved the overall accuracy. For example: were the common classes mostly correct, while less common classes often misclassified? Which classes were most often confused? This should be called with test data that is independent of the training data in order to understand the out-of-sample (i.e. generalization) performance.

```
PARAMETER <u>model</u> ||| TABLE ( Layout_Model2 ) — The encoded model as returned from GetModel.
```

PARAMETER <u>actuals</u> ||| TABLE ( DiscreteField ) — The actual class values.

PARAMETER observations || TABLE (NumericField) — The independent (explanatory) values.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 classifier , INTEGER4 actual\_class , INTEGER4 predict\_class , UNSIGNED4 occurs , BOOLEAN correct , REAL8 pctActual , REAL8 pctPred } ) — DATASET(Confusion\_Detail), one record per cell of the confusion matrix.

**SEE** Types.Confusion\_Detail.

### Interfaces/

## **IRegression**

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### **DESCRIPTIONS**

## **IREGRESSION** IRegression

/ EXPORT | IRegression

(DATASET(NumericField) X=empty\_data, DATASET(NumericField) Y=empty\_data)

\*\*\*DEPRECATED\*\*\* Interface Definition for Regression Modules (version 1). This interface is being deprecated and should not be used for new bundles or bundles undergoing substantial revision. Please use IRegression2 going forward. Regression learns a function that maps a set of input data to one or more output variables. The resulting learned function is known as the model. That model can then be used repetitively to predict (i.e. estimate) the output value(s) based on new input data.

**PARAMETER** X ||| TABLE (NumericField) — The independent data in DATASET(NumericField) format. Each statistical unit (e.g. record) is identified by 'id', and each feature is identified by field number (i.e. 'number').

**PARAMETER** Y ||| TABLE (NumericField) — The dependent variable(s) in DATASET(NumericField) format. Each statistical unit (e.g. record) is identified by 'id', and each feature is identified by field number (i.e. 'number').

### Children

- 1. GetModel (From IRegression): Calculate and return the 'learned' model
- 2. Predict (From IRegression): Predict the output variable(s) based on a previously learned model

### **GETMODEL** GetModel

IRegression \

DATASET(Layout\_Model) GetModel

Calculate and return the 'learned' model. The model may be persisted and later used to make predictions using 'Predict' below.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — DATASET(LayoutModel) describing the learned model parameters.

### PREDICT Predict

IRegression \

 ${\tt DATASET(NumericField)} \mid \mathbf{Predict}$ 

(DATASET(NumericField) newX, DATASET(Layout Model) model)

Predict the output variable(s) based on a previously learned model.

**PARAMETER** <u>newX</u> ||| TABLE ( NumericField ) — DATASET(NumericField) containing the X values to b predicted.

PARAMETER model || TABLE ( Layout\_Model ) — No Doc

**RETURN** TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — DATASET(NumericField) containing one entry per observation (i.e. id) in newX. This represents the predicted values for Y.

## ${\bf IRegression 2}$

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### **DESCRIPTIONS**

## IREGRESSION2 | IRegression2

IRegression2

Interface Definition for Regression Modules (Version 2). Regression learns a function that maps a set of input data to one or more continuous output variables. The resulting learned function is known as the model. That model can then be used repetitively to predict (i.e. estimate) the output value(s) based on new input data. Actual implementation modules will probably take configuration parameters to control the regression process. The regression modules also expose attributes for assessing the effectiveness of the regression.

### Children

- 1. GetModel (From IRegression2): Calculate and return the 'learned' model
- 2. Predict (From IRegression2): Predict the output variable(s) based on a previously learned model
- 3. Accuracy (From IRegression2): Assess the accuracy of a set of predictions

### **GETMODEL** GetModel

### IRegression2 \

DATASET(Layout_Model2)	$\operatorname{GetModel}$
(DATASET(NumericField) dependents)	independents, DATASET(NumericField)

Calculate and return the 'learned' model.

The model may be persisted and later used to make predictions using 'Predict' below.

**PARAMETER** independents ||| TABLE (NumericField) — The independent data in DATASET(NumericField) format. Each statistical unit (e.g. record) is identified by 'id', and each feature is identified by field number (i.e. 'number').

PARAMETER <u>dependents</u> ||| TABLE ( NumericField ) — The dependent variable(s) in DATASET(NumericField) format. Each statistical unit (e.g. record) is identified by 'id', and each feature is identified by field number (i.e. 'number').

RETURN TABLE ( { UNSIGNED2 wi , REAL8 value , SET ( UNSIGNED4 ) indexes } )
— The encoded model.

SEE Types.NumericField

SEE Types.Layout\_Model2

### **PREDICT** Predict

### IRegression2 \

# DATASET(NumericField) Predict (DATASET(Layout\_Model2) model, DATASET(NumericField) observations)

Predict the output variable(s) based on a previously learned model

**PARAMETER** independents ||| — the observations upon which to predict.

PARAMETER model || TABLE ( Layout\_Model2 ) — No Doc

PARAMETER observations || TABLE ( NumericField ) — No Doc

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — one entry per observation (i.e. id) in observations. This represents the predicted values for the dependent variable(s).

### **ACCURACY** Accuracy

### IRegression2 \

DATASET(Regression_Accuracy)	Accuracy
(DATASET(Layout_Model2) model actuals, DATASET(NumericField	

Assess the accuracy of a set of predictions. This is equivalent to calling predict and then Analysis.Regression.Accuracy.

PARAMETER model || TABLE ( Layout\_Model2 ) — The model as returned from GetModel

**PARAMETER** <u>actuals</u> ||| TABLE ( NumericField ) — The actual values of the dependent variable to compare with the predictions.

**PARAMETER** observations ||| TABLE (NumericField) — The independent data upon which the accuracy assessment is to be based.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 regressor , REAL8 R2 , REAL8 MSE , REAL8 RMSE } ) — Accuracy statistics (see Types.Regression Accuracy for details)

## Math

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Compute the beta value of two positive real numbers, x and y

### Distributions.ecl

Compute PDF, CDF, and PPF values for various Probability Distributions

#### DoubleFac.ecl

Compute the double factorial

### Fac.ecl

Factorial function, (i)(i-1)(i-2)...(2)

### gamma.ecl

Compute the value of gamma function of real number **x** 

### log\_gamma.ecl

Compute the value of the log gamma function of the absolute value of X

### lowerGamma.ecl

Compute the lower incomplete gamma value of two real numbers, x and y

### NCK.ecl

N Choose K – finds the number of combinations of K elements out of a possible N

### Poly.ecl

Evaluate a polynomial from a set of coefficients

### StirlingFormula.ecl

Stirling's formula

### upperGamma.ecl

Compute the upper incomplete gamma value of two real numbers, x and y

## Math/ Beta

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## **IMPORTS**

Math |

## **DESCRIPTIONS**

## **BETA** Beta

```
/ EXPORT Beta

(REAL8 x, REAL8 y)
```

Compute the beta value of two positive real numbers, x and y.

PARAMETER  $\underline{\mathbf{x}}$  ||| REAL8 — the value of the first number PARAMETER  $\underline{\mathbf{y}}$  ||| REAL8 — the value of the second number

**RETURN REAL8** — the beta value

## Math/ **Distributions**

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### **IMPORTS**

Constants | Math |

### **DESCRIPTIONS**

### **DISTRIBUTIONS** Distributions

**Distributions** 

Compute PDF, CDF, and PPF values for various Probability Distributions.

The Probability Density Function (PDF(x)) of a distribution is the relative likelihood of a sample drawn from that distribution being of value x.

The Cumulative Distribution Function (CDF(x)) of a distribution is the probability of a sample drawn from that distribution to be less than or equal to x.

The Percentage Point Function (PPF(x)) of a distribution is the inverse of the CDF. Given a probability, it returns the value at which the probability of occurrence is less than or equal to the given probability.

### Children

- 1. Normal\_CDF (From Distributions): Cumulative Distribution Function (CDF) of the standard normal distribution
- 2. Normal\_PPF (From Distributions): Percentage Point Function (PPF) for the Normal Distribution

- 3. T\_CDF (From Distributions): Cumulative Distribution Function (CDF) for Students t distribution
- 4. T\_PPF (From Distributions): Percentage point function (PPF) for the T distribution
- 5. Chi2\_CDF (From Distributions): The Cumulative Distribution Function (CDF) for the Chi Square distribution for the specified degrees of freedom
- 6. Chi2\_PPF (From Distributions): Probability Point Function (PPF) for the Chi Squared distribution

### NORMAL\_CDF Normal\_CDF

### Distributions \

REAL8 Normal\_CDF

(REAL8 x)

Cumulative Distribution Function (CDF) of the standard normal distribution. The probability that a normal random variable will be smaller than or equal to x standard deviations above or below the mean.

Taken from C/C++ Mathematical Algorithms for Scientists and Engineers, n. Shammas, McGraw-Hill, 1995.

PARAMETER x ||| REAL8 — the number of standard deviations.

**RETURN REAL8** — probability of exceeding x.

### NORMAL\_PPF Normal\_PPF

### Distributions \

REAL8 Normal\_PPF

(REAL8 x)

Percentage Point Function (PPF) for the Normal Distribution.

Translated from C/C++ Mathematical Algorithms for Scientists and Engineers, N. Shammas, McGraw-Hill, 1995.

PARAMETER <u>x</u> ||| REAL8 — probability.

**RETURN REAL8** — number of standard deviations from the mean.

### T\_CDF T\_CDF

### Distributions \

REAL8 T\_CDF

(REAL8 x, REAL8 df)

Cumulative Distribution Function (CDF) for Students t distribution.

The integral evaluated between negative infinity and x.

Translated from NIST SEL DATAPAC Fortran TCDF.f source.

**PARAMETER**  $\underline{\mathbf{x}} \parallel \parallel \text{REAL8} - \text{value of the evaluation.}$ 

**PARAMETER**  $\underline{\mathbf{df}} \parallel \parallel \text{REAL8} - \text{degrees of freedom.}$ 

**RETURN REAL8** — the probability that a value will be less than or equal to the specified value.

### T\_PPF T\_PPF

### Distributions \

REAL8 T\_PPF
(REAL8 x, REAL8 df)

Percentage point function (PPF) for the T distribution.

Translated from NIST SEL DATAPAC Fortran TPPF.f source.

PARAMETER <u>x</u> ||| REAL8 — the probability.

PARAMETER df | REAL8 — degrees of freedom of the distribution.

**RETURN REAL8** — the value with that probability.

### CHI2\_CDF Chi2\_CDF

### Distributions \

REAL8 Chi2\_CDF

(REAL8 x, REAL8 df)

The Cumulative Distribution Function (CDF) for the Chi Square distribution for the specified degrees of freedom.

Translated from the NIST SEL DATAPAC Fortran subroutine CHSCDF.

**PARAMETER**  $\underline{\mathbf{x}} \parallel \parallel \text{REAL8}$  — the value at which to compute.

**PARAMETER**  $\underline{\mathbf{df}} \parallel \mathbf{REAL8}$  — the degrees of freedom of the distribution.

**RETURN REAL8** — the cumulative probability.

### CHI2\_PPF Chi2\_PPF

### Distributions \

REAL8 Chi2\_PPF

(REAL8 x, REAL8 df)

Probability Point Function (PPF) for the Chi Squared distribution.

Translated from the NIST SEL DATAPAC Fortran subroutine CHSPPF.

**PARAMETER**  $\underline{\mathbf{x}} \parallel \parallel \text{REAL8}$  — the probability value.

**PARAMETER**  $\underline{\mathbf{df}}$  ||| REAL8 — the degrees of freedom of the distribution.

**RETURN REAL8** — the value with that probability.

## $\frac{\mathrm{Math/}}{\mathrm{DoubleFac}}$

Go Up

## **DESCRIPTIONS**

## **DOUBLEFAC** DoubleFac

/ EXPORT REAL8	DoubleFac
(INTEGER2 i)	

Compute the double factorial. The double factorial is defined for odd n as the product of all the odd numbers up to and including that number.

For even numbers it is the product of the even numbers up to and including that number.

Thus DoubleFac(8) = 8\*6\*4\*2.

IF i < 2, the value 1 is returned.

**PARAMETER**  $\underline{\mathbf{i}}$  ||| INTEGER2 — the input value.

**RETURN REAL8** — the numeric result.

## $\frac{\mathrm{Math}/}{Fac}$

 ${\rm Go}\ {\rm Up}$ 

## **DESCRIPTIONS**

## FAC Fac

/ EXPORT REAL8	Fac
(UNSIGNED2 i)	

Factorial function, (i)(i-1)(i-2)...(2)

**RETURN REAL8** — the factorial i!.

## Math/

## gamma

Go Up

## **DESCRIPTIONS**

## **GAMMA** gamma

/ EXPORT REAL8	gamma
(REAL8 x)	

Compute the value of gamma function of real number  $\mathbf{x}$ .

This is a wrapper for the standard C tgamma function.

**PARAMETER**  $\underline{\mathbf{x}} \parallel \parallel \text{REAL8}$  — the input value.

**RETURN REAL8** — the value of GAMMA evaluated at x.

### Math/

## log\_gamma

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## **DESCRIPTIONS**

## LOG\_GAMMA log\_gamma

/ EXPORT REAL8	log_gamma
(REAL8 x)	

Compute the value of the log gamma function of the absolute value of X.

This is wrapper for the standard C lgamma function. Avoids the race condition found on some platforms by taking the absolute value of the input argument.

**PARAMETER**  $\underline{\mathbf{x}} \parallel \parallel \text{REAL8}$  — the input x.

**RETURN REAL8** — the value of the log of the GAMMA evaluated at ABS(x).

## Math/ lowerGamma

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## **DESCRIPTIONS**

## **LOWERGAMMA** lowerGamma

/ EXPORT REAL8	lowerGamma
(REAL8 x, REAL8	y)

Compute the lower incomplete gamma value of two real numbers,  ${\bf x}$  and  ${\bf y}$ .

**PARAMETER**  $\underline{\mathbf{x}}$  ||| REAL8 — the value of the first number.

**PARAMETER**  $\underline{\mathbf{y}}$  ||| REAL8 — the value of the second number.

**RETURN REAL8** — the lower incomplete gamma value.

## Math/ NCK

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## **IMPORTS**

Math |

## **DESCRIPTIONS**

## NCK NCK

/ EXPORT REAL8 NCK

(INTEGER2 N, INTEGER2 K)

N Choose K – finds the number of combinations of K elements out of a possible N.

**PARAMETER** N || INTEGER2 — the number of items in the population.

**PARAMETER**  $\underline{\mathbf{K}}$  ||| INTEGER2 — the number of items to choose.

**RETURN REAL8** — the number of combinations.

## Math/Poly

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## **DESCRIPTIONS**

## **POLY** Poly

/ EXPORT REAL8 Poly

(REAL8 x, SET OF REAL8 Coeffs)

Evaluate a polynomial from a set of coefficients.

Coeffs 1 is assumed to be the HIGH order of the equation.

Thus for  $ax^2+bx+c$  - the set would need to be Coef := [a,b,c];

**PARAMETER**  $\underline{\mathbf{x}} \parallel \parallel \text{REAL8}$  — the value of x in the polynomial.

**PARAMETER** Coeffs ||| SET ( REAL8 ) — a set of coefficients for the polynomial. The ALL set is considered to be all zero values.

**RETURN REAL8** — value of the polynomial at x.

## ${\bf Stirling Formula}$

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## **IMPORTS**

Math | Constants |

## **DESCRIPTIONS**

## **STIRLINGFORMULA** StirlingFormula

/ EXPORT StirlingFormula

(REAL x)

Stirling's formula.

**PARAMETER**  $\underline{\mathbf{x}} \parallel \parallel \text{REAL8}$  — the point of evaluation.

**RETURN REAL8** — evaluation result.

## $\begin{array}{c} {\rm Math/} \\ {\bf upper Gamma} \end{array}$

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## **DESCRIPTIONS**

## **UPPERGAMMA** upperGamma

/ EXPORT REAL8	upperGamma
(REAL8 x, REAL8 y)	

Compute the upper incomplete gamma value of two real numbers, x and y.

**PARAMETER**  $\underline{\mathbf{x}} \parallel \parallel \text{REAL8}$  — the value of the first number.

**PARAMETER**  $\mathbf{y} \parallel \parallel \text{REAL8}$  — the value of the second number.

**RETURN REAL8** — the upper incomplete gamma value.

## Preprocessing

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Allows to convert categorical values into numeric format

### MinMaxScaler.ecl

Scale the input data to a defined range [Min, Max]

#### Normalizer.ecl

Normalizer Normalizes each sample to its unit norm (row-wise normalization) with below options L1 norm

#### OneHotEncoder.ecl

OneHotEncoder OneHotEncode is used to convert each of the designated categorical features to a binary (absent/present) value (i.e.oneHot) for use by algorithms that don't directly support categorical values

### Split.ecl

Split input data into training and test sets based on the split ratio

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Standardize the data by mapping to zero mean and standard deviation of 1.0

### StratifiedSplit.ecl

Split input data into training and test sets based on the split ratio

### Types.ecl

Record structures for Preprocessing modules

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Utils

### Preprocessing/

## LabelEncoder

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### **DESCRIPTIONS**

## LABELENCODER LabelEncoder

LabelEncoder

Allows to convert categorical values into numeric format. For example: use LabelEncoder to convert below raw data: raw := DATASET([{'apple'}, {'grape'}], {STRING fruit}); The result is as following: convertedDs := DATASET([{0}, {1}], {INTEGER fruit}); Currently does not support Myriad interface

#### Children

- 1. GetKey (From LabelEncoder): Builds a mapping between feature names and categories
- 2. GetMapping (From LabelEncoder): Builds a lookup table that maps each category of a feature to a unique number
- 3. Encode (From LabelEncoder): Replaces each categorical value in the data with its index in the key
- 4. Decode (From LabelEncoder): Converts back the categorical values into their original labels

### **GETKEY** GetKey

LabelEncoder \

GetKey

(dataForUndefinedCategories, partialKey)

Builds a mapping between feature names and categories.

- PARAMETER dataForUndefinedCategories: ||| INTEGER8 any record-oriented dataset.

  The data from which the categories are extracted if not predefined in the list of categorical features.
- PARAMETER partialKey: || INTEGER8 same record structure as the key (see below). Mapping between feature names and categories. Some names are mapped to empty categories such that their categories could be extracted from dataForUndefinedCategories.
- **RETURN BOOLEAN** key: DATASET(KeyLayout) The full mapping between categorical feature names and their categories. Its record structure has the following format: 

  <name of="" categorical="" feature="">; SET OF STRING <name of=""
  categorical="" feature="">; ... SET OF STRING <name of="" categorical="" feature=""
  n="">; END; </name></name></name>
  SET OF STRING <name > SET OF STRING

## **GETMAPPING GetMapping**

LabelEncoder \

GetMapping (key)

Builds a lookup table that maps each category of a feature to a unique number. Each category is assigned its index in the category set.

PARAMETER key: || INTEGER8 — DATASET(KeyLayout). Mapping between feature names and categories.

**RETURN BOOLEAN** — categoriesMapping: DATASET(MappingLayout). A table with each feature name mapped to its categories and each category mapped to its value. 
//record mapping a category to its value.
br/> Category := RECORD
br/> STRING categoryName;
br/> INTEGER value;
br/> END;
br/> //record mapping feature names to their categories.
br/> MappingLayout := RECORD
br/> STRING featureName;
br/> DATASET(Category) categories;
br/> END;
br/> END;
c/pre>

### **ENCODE** Encode

### LabelEncoder \

### Encode

(dataToEncode, key)

Replaces each categorical value in the data with its index in the key. Every unknown category (not in the key) is replaced by -1.

PARAMETER dataToEncode: || INTEGER8 — any dataset. The data to encode.

PARAMETER <u>key:</u> ||| INTEGER8 — DATASET(KeyLayout). Mapping between feature names and their categories.

**RETURN BOOLEAN** — encodedData: same record structure as dataToEncode with the datatype of all categorical features changed to INTEGER. Data with categorical values replaced by numbers.

### **DECODE** Decode

### LabelEncoder \

### Decode

(dataToDecode, encoderKey)

Converts back the categorical values into their original labels. Every -1 is replaced by an empty string.

PARAMETER dataToDecode: || INTEGER8 — any dataset. The data to decode.

PARAMETER key: || — DATASET(KeyLayout). Mapping between feature names and their categories.

PARAMETER encoderkey || INTEGER8 — No Doc

**RETURN BOOLEAN** — decodedData: same record structure as dataToDecode with the datatype of all categorical features changed to STRING. Data with categorical values replaced by their original labels.

### Preprocessing/

## MinMaxScaler

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### **IMPORTS**

std |

### **DESCRIPTIONS**

### MINMAXSCALER MinMaxScaler

# / EXPORT MinMaxScaler (DATASET(NumericField) baseData = DATASET([], NumericField), t\_FieldReal lowBound = 0.0, t\_FieldReal highBound = 1.0, DATASET(KeyLayout) key = DATASET([], KeyLayout))

Scale the input data to a defined range [Min, Max]. Curently does not support Myriad interface

- **PARAMETER** baseData: ||| TABLE ( NumericField ) DATASET(NumericField), Default = DATASET([], NumericField). The data from which the minimums and maximums are determined.
- PARAMETER <u>low:</u> ||| t\_FieldReal, Default = 0.0 The minimum value of the normalized data.
- PARAMETER <u>high:</u> ||| t\_FieldReal, Default = 1.0 The maximum value of the normalized data.
- PARAMETER <u>key:</u> ||| TABLE ( KeyLayout ) DATASET(KeyLayout), default = DATASET([], KeyRec). The key to be reused for scaling/unscaling.
- PARAMETER <u>lowbound</u> ||| REAL8 No Doc

### PARAMETER highbound ||| REAL8 — No Doc

### SEE StandardScaler

### Children

- 1. GetKey (From MinMaxScaler): Computes the key or reuses it if already given
- 2. Scale (From MinMaxScaler): scales the data using the following formula:
- 3. unscale (From MinMaxScaler): unscales the data using the following formula

## **GETKEY** GetKey

MinMaxScaler \

GetKey ()

Computes the key or reuses it if already given.

**RETURN** TABLE ( { REAL8 lowBound , REAL8 highBound , TABLE ( FeatureMinMax ) minsMaxs } ) — the key: DATASET(KeyLayout).

### **SCALE** Scale

MinMaxScaler \

Scale

(DATASET(NumericField) dataToScale)

scales the data using the following formula:  $x' = \min + ([(x - x_min)(max - min)]/(x_max - x_min))$ 

PARAMETER <u>dataToScale:</u> ||| TABLE ( NumericField ) — DATASET(NumericField). The data to scale.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — the scaled data: DATASET(NumericField)

### **UNSCALE** unscale

MinMaxScaler \

unscale

(DATASET(NumericField) dataToUnscale)

unscales the data using the following formula  $x = x_{min} + ((x' - min)(x_{max} - x_{min}))/(max-min)$ 

**PARAMETER** <u>dataToUnscale:</u> ||| TABLE ( NumericField ) — DATASET(NumericField) The data to unscale.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — the unscaled data: DATASET(NumericField).

#### Preprocessing/

# Normalizer

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#### **IMPORTS**

\_versions.ML\_Core.V3\_2\_2.ML\_Core.Types

#### **DESCRIPTIONS**

#### **NORMALIZER Normalizer**

#### Normalizer

(DATASET(MTypes.NumericField) dataToNormalize, STRING3 norm = '12')

Normalizer Normalizes each sample to its unit norm (row-wise normalization) with below options L1 norm.

Given a set of values, the L1 norm is the sum of absolute values. L2 norm.

Given a set of values, the L2 norm is the square root of the sum of squares. L-Infinty norm.

Given a set of values the l-infinty norm is the value with highest absolute value.

PARAMETER <u>dataToNormalize:</u> ||| TABLE ( NumericField ) — DATASET(Types.NumericField) The data to normalize.

**PARAMETER** norm: ||| STRING3 — STRING3, Default = 'l2'. The norm based on which the data will be normalized. valid values: 'l1', 'l2', 'inf'.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8

<b>value</b> $\}$ ) — the normalized Data: interface.	${\bf DATASET (Numeric Field)}.$	Curently does not support My	vriad

#### Preprocessing/

# OneHotEncoder

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#### **IMPORTS**

Preprocessing. Types |

#### **DESCRIPTIONS**

#### **ONEHOTENCODER** OneHotEncoder

```
/ EXPORT OneHotEncoder

(DATASET(NumericField) ds = DATASET([], NumericField), DATASET(l_cFeatures)
categoricalFeatures = DATASET([], l_cFeatures))
```

OneHotEncoder OneHotEncode is used to convert each of the designated categorical features to a binary (absent/present) value (i.e.oneHot) for use by algorithms that don't directly support categorical values. Also can convert back from oneHot encoding to numerical category. Each categorical field will produce additional features according to its cardinality. For example, if there are four possible categories, then the original feature will be replaced by four binary features. Supports Myriad Interface.

PARAMETER <u>ds</u> ||| TABLE ( NumericField ) — dataset to be encoded.

**PARAMETER** categoricalFeatures ||| TABLE ( l\_cFeatures ) — categorical feature IDs for each work item. e.g. to encoded field number 3 for work item 1, below categoricalFeatures can be used: DATASET([{1, 3}], l\_cFeatures)

#### Children

1. isValidInput (From OneHotEncoder): Validates input

- 2. getMappings (From OneHotEncoder): No Documentation Found
- 3. encode (From OneHotEncoder): No Documentation Found
- 4. decode (From OneHotEncoder): Revert the encoded data to its original form

#### **ISVALIDINPUT** isValidInput

OneHotEncoder \

is Valid Input

()

Validates input.

**RETURN BOOLEAN** — True when input is valid, False otherwise.

## **GETMAPPINGS** getMappings

OneHotEncoder \

getMappings

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED4 number , REAL8 value , UNSIGNED4 newNum } ) —

# **ENCODE** encode

OneHotEncoder \

encode

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) —

#### **DECODE** decode

OneHotEncoder \

decode

(DATASET(NumericField) encodedDS)

Revert the encoded data to its original form

PARAMETER encodedDS || TABLE ( NumericField ) — encoded data

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — decoded decoded data

#### Preprocessing/

# **Split**

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#### **IMPORTS**

Preprocessing | Preprocessing. Types |

#### **DESCRIPTIONS**

#### **SPLIT** Split

#### Split

(DATASET(NumericField) dataToSplit, REAL4 splitRatio = 0.0, BOOLEAN shuffle = FALSE)

Split input data into training and test sets based on the split ratio. It requires the data has sequential id starting with 1. Curently does not support Myriad interface

PARAMETER dataToSplit: ||| TABLE ( NumericField ) — DATASET(Types.NumericField). The data to split.

**PARAMETER** splitRatio: ||| REAL4 — REAL4, DEFAULT = 0.5. The percentage of input data split as training data.

**PARAMETER** shuffle: ||| BOOLEAN — Boolean, DEFAULT = false. if true, the data is shuffled before splitting.

**RETURN** — training and test data Note: currently not support Myraid interface.

#### Children

- 1. trainData (From Split): No Documentation Found
- 2. testData (From Split): No Documentation Found

## TRAINDATA trainData

Split \

trainData

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) —

## TESTDATA testData

Split \

testData

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) —

#### Preprocessing/

# StandardScaler

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#### **IMPORTS**

#### **DESCRIPTIONS**

#### STANDARDSCALER StandardScaler

```
/ EXPORT StandardScaler

(DATASET(NumericField) baseData = DATASET([], NumericField),
DATASET(KeyLayout) key = DATASET([], KeyLayout))
```

Standardize the data by mapping to zero mean and standard deviation of 1.0. Curently does not support Myriad interface

**PARAMETER** baseData: ||| TABLE ( NumericField ) — DATASET(NumericField), default = DATASET([], Types.NumericField) The data from which the means and standard deviations are determined for each feature.

**PARAMETER key:** ||| TABLE ( KeyLayout ) — DATASET(KeyLayout), default = DATASET([], KeyRec) The key to be reused for scaling/unscaling.

#### Children

- 1. GetKey (From StandardScaler): Compute the mean and standard deviation per feature or reuses the key if provided
- 2. Scale (From StandardScaler): scale the data using the following formula
- 3. unscale (From StandardScaler): unscale the data using the following formula:

#### **GETKEY** GetKey

StandardScaler \

	GetKey	
()		

Compute the mean and standard deviation per feature or reuses the key if provided.

RETURN TABLE ( { UNSIGNED4 featureId , REAL8 avg , REAL8 stdev } ) — key: DATASET(KeyLayout).

#### **SCALE** Scale

StandardScaler \

Scale

(DATASET(NumericField) dataToScale)

scale the data using the following formula x' = (x - mean)/stdev

**PARAMETER** <u>dataToScale:</u> ||| TABLE ( NumericField ) — DATASET(NumericField). The data to scale

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — the scaled data: DATASET(NumericField)

#### **UNSCALE** unscale

StandardScaler \

unscale

(DATASET(NumericField) dataToUnscale)

unscale the data using the following formula: x = (x' \* stdev) + mean

**PARAMETER** <u>dataToUnscale:</u> ||| TABLE ( NumericField ) — DATASET(NumericField). The data to unscale.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — the unscaled data: DATASET(NumericField).

#### Preprocessing/

# StratifiedSplit

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#### **IMPORTS**

\_\_versions.ML\_Core.V3\_2\_2.ML\_Core.Types

#### **DESCRIPTIONS**

#### STRATIFIEDSPLIT StratifiedSplit

```
/ EXPORT | StratifiedSplit

(DATASET(NumericField) ds, REAL4 trainSize = 0, REAL4 testSize = 0,
UNSIGNED labelId = 0, BOOLEAN shuffle = FALSE)
```

Split input data into training and test sets based on the split ratio. The result preservees the percentage of the samples for the specific feature or class. It requires the data has sequential id starting with 1. Curently does not support Myriad interface.

PARAMETER ds: || TABLE ( NumericField ) — DATASET(NumericField). The data to split.

**PARAMETER** <u>trainSize:</u> ||| REAL4 — REAL4, Default = 0.0 The training size.

**PARAMETER** <u>testSize:</u> ||| REAL4 — REAL4, Default = 0.0 The test size.

**PARAMETER** <u>labelId:</u> ||| UNSIGNED8 — UNSIGNED, Default = 0. The number of the field whose proportions has to be maintained.

PARAMETER shuffle || BOOLEAN — No Doc

**RETURN** — the training data, test data as DATASET(NumericField).

#### Children

- 1. trainData (From StratifiedSplit): No Documentation Found
- 2. testData (From StratifiedSplit): No Documentation Found

#### TRAINDATA trainData

StratifiedSplit \

trainData

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) —

#### TESTDATA testData

StratifiedSplit \

testData

No Documentation Found

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) —

#### Preprocessing/

# **Types**

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# **TYPES** Types

Types

Record structures for Preprocessing modules.

#### Children

- 1. valueLayout (From Types): No Documentation Found
- 2. numberLayout (From Types): No Documentation Found
- 3. idLayout (From Types): No Documentation Found
- 4. OneHotEncoder (From Types): record structures for OneHotEncoder
- 5. StandardScaler (From Types): record structures for StandardScaler
- 6. MinMaxScaler (From Types): record structures for MinMaxScaler
- 7. Normaliz (From Types): record structures for normalize function

#### **VALUELAYOUT** valueLayout

Types \

valueLayout

No Documentation Found

FIELD value || REAL8 — No Doc

# **NUMBERLAYOUT** numberLayout

Types \

numberLayout

No Documentation Found

FIELD <u>number</u> ||| UNSIGNED4 — No Doc

## **IDLAYOUT** idLayout

Types \

idLayout

No Documentation Found

FIELD id || UNSIGNED8 — No Doc

## **ONEHOTENCODER** OneHotEncoder

Types \

#### OneHotEncoder

record structures for OneHotEncoder.

#### Children

1. cFeatures (From OneHotEncoder): No Documentation Found

## **CFEATURES** cFeatures

Types \ OneHotEncoder \

**cFeatures** 

No Documentation Found

FIELD wi || UNSIGNED8 — No Doc

FIELD <u>number</u> || UNSIGNED8 — No Doc

## STANDARDSCALER StandardScaler

Types \

StandardScaler

record structures for StandardScaler.

#### Children

1. KeyLayout (From StandardScaler): No Documentation Found

# **KEYLAYOUT** KeyLayout

Types \ StandardScaler \

**KeyLayout** 

No Documentation Found

FIELD <u>featureid</u> || UNSIGNED4 — No Doc

FIELD avg ||| REAL8 — No Doc

FIELD stdev ||| REAL8 — No Doc

#### MINMAXSCALER MinMaxScaler

Types \

MinMaxScaler

record structures for MinMaxScaler.

#### Children

- 1. FeatureMinMax (From MinMaxScaler): No Documentation Found
- 2. KeyLayout (From MinMaxScaler): No Documentation Found

#### FEATUREMINMAX FeatureMinMax

Types  $\setminus$  MinMaxScaler  $\setminus$ 

FeatureMinMax

No Documentation Found

- FIELD <u>featureid</u> || UNSIGNED4 No Doc
- FIELD minvalue ||| REAL8 No Doc
- FIELD maxvalue ||| REAL8 No Doc

# **KEYLAYOUT** KeyLayout

Types \ MinMaxScaler \

#### KeyLayout

No Documentation Found

- FIELD <u>lowbound</u> ||| REAL8 No Doc
- **FIELD** highbound ||| REAL8 No Doc
- **FIELD** minsmaxs ||| TABLE (FeatureMinMax) No Doc

#### **NORMALIZ** Normaliz

Types \

Normaliz

record structures for normalize function.

#### Children

1. normsLayout (From Normaliz): No Documentation Found

# NORMSLAYOUT normsLayout

Types \ Normaliz \

normsLayout

No Documentation Found

FIELD id || UNSIGNED8 — No Doc

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TestNormalize
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Data and Record structures used by TestLabelEncoder Modules

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- 2. key (From TestDataAndTypes): No Documentation Found
- 3. sampleDataLayout (From TestDataAndTypes): No Documentation Found
- 4. sampleData (From TestDataAndTypes): No Documentation Found
- 5. sampleData2 (From TestDataAndTypes): No Documentation Found
- 6. EncodedLayout (From TestDataAndTypes): No Documentation Found
- 7. encodedData1 (From TestDataAndTypes): No Documentation Found
- 8. encodedData2 (From TestDataAndTypes): No Documentation Found
- 9. DecodedLayout (From TestDataAndTypes): No Documentation Found
- 10. decodedData1 (From TestDataAndTypes): No Documentation Found
- 11. decodedData2 (From TestDataAndTypes): No Documentation Found

## **KEYLAYOUT** KeyLayout

TestDataAndTypes \

**KeyLayout** 

No Documentation Found

FIELD <u>f1</u> ||| SET (STRING) — No Doc

FIELD <u>f3</u> ||| SET (STRING) — No Doc

FIELD <u>f4</u> ||| SET ( STRING ) — No Doc

## KEY key

TestDataAndTypes \

key

No Documentation Found

RETURN ROW ( KeyLayout ) —

#### **SAMPLEDATALAYOUT** sampleDataLayout

 $TestDataAndTypes \ \setminus \\$ 

sample Data Layout

No Documentation Found

FIELD id || UNSIGNED8 — No Doc

FIELD <u>f1</u> ||| STRING — No Doc

FIELD <u>f2</u>     UNSIGNED8 — No Doc
FIELD <u>f3</u>     UNSIGNED8 — No Doc
FIELD <u>f4</u>     STRING — No Doc
SAMPLEDATA sampleData
TestDataAndTypes \
sampleData
No Documentation Found
$ \begin{array}{c} \textbf{RETURN} & \textbf{TABLE} \ ( \ \textbf{sampleDataLayout} \ ) \ \end{array} $
SAMPLEDATA2 sampleData2
SAMI ELDATAZ SampleDataZ
TestDataAndTypes \
sampleData2
No Documentation Found
$ \begin{array}{ c c c c c c c c c c c c c c c c c c c$
ENCODEDLAYOUT EncodedLayout
The ADA A The second
TestDataAndTypes \
EncodedLayout

No Documentation Found

```
FIELD id || UNSIGNED8 — No Doc
FIELD f1 || INTEGER8 — No Doc
```

#### **ENCODEDDATA1** encodedData1

TestDataAndTypes \

encodedData1

No Documentation Found

RETURN TABLE (EncodedLayout) —

#### **ENCODEDDATA2** encodedData2

TestDataAndTypes \

encodedData2

No Documentation Found

RETURN TABLE ( EncodedLayout ) —

## **DECODEDLAYOUT** DecodedLayout

TestDataAndTypes \

DecodedLayout

No Documentation Found

FIELD id || UNSIGNED8 — No Doc

FIELD <u>f1</u> ||| STRING — No Doc

FIELD <u>f2</u> ||| UNSIGNED8 — No Doc

FIELD <u>f3</u> ||| STRING — No Doc

FIELD <u>f4</u> ||| STRING — No Doc

#### **DECODEDDATA1** decodedData1

TestDataAndTypes \

decodedData1

No Documentation Found

RETURN TABLE ( DecodedLayout ) —

### **DECODEDDATA2** decodedData2

 $TestDataAndTypes \ \backslash$ 

decodedData2

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**TestData** 

Test data for testing standardScaler module

#### Children

- 1. sampleData (From TestData): No Documentation Found
- 2. key1 (From TestData): No Documentation Found
- 3. key2 (From TestData): No Documentation Found
- 4. scaledData1 (From TestData): No Documentation Found
- 5. scaledData2 (From TestData): No Documentation Found

#### **SAMPLEDATA** sampleData

TestData \

sampleData

RETURN TABLE ( NumericField ) —

# KEY1 key1

TestData \

key1

No Documentation Found

# KEY2 key2

TestData \

key2

No Documentation Found

RETURN TABLE ( KeyLayout ) —

## SCALEDDATA1 scaledData1

TestData \

scaledData1

No Documentation Found

<b>RETURN</b>	TABLE	( NumericField )	) —
---------------	-------	------------------	-----

# SCALEDDATA2 scaledData2

 ${\bf TestData}\ \backslash$ 

scaledData2

No Documentation Found

RETURN TABLE ( NumericField ) -

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- 1. sampleData (From testData): No Documentation Found
- 2. l1NormResult (From testData): No Documentation Found
- 3. l2NormResult (From testData): No Documentation Found
- 4. IInfNormResult (From testData): No Documentation Found

#### **SAMPLEDATA** sampleData

testData \

sampleData

RETURN	TABLE	( NumericField	) —

## L1NORMRESULT | I1NormResult

testData \

l1NormResult

No Documentation Found

RETURN TABLE ( NumericField ) —

## L2NORMRESULT | 12NormResult

testData \

l2NormResult

No Documentation Found

RETURN TABLE ( NumericField ) —

#### LINFNORMRESULT IInfNormResult

testData \

lInfNormResult

No Documentation Found

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# TESTDATA TestData

**TestData** 

Test Data for Testing OneHotEncoder

#### Children

- 1. validFeatureIds (From TestData): No Documentation Found
- 2. invalidFeatureIds (From TestData): No Documentation Found
- 3. key (From TestData): No Documentation Found
- 4. sample1 (From TestData): No Documentation Found
- 5. sample2 (From TestData): No Documentation Found
- 6. encodedSample1 (From TestData): No Documentation Found
- 7. encodedSample2 (From TestData): No Documentation Found

# **VALIDFEATUREIDS** validFeaturelds

TestData \

validFeatureIds No Documentation Found RETURN SET (INTEGER8)— **INVALIDFEATUREIDS** invalidFeaturelds TestData \ invalidFeatureIds No Documentation Found RETURN SET (INTEGER8)— **KEY** key TestData \ key No Documentation Found RETURN TABLE ( cFeatures ) —

# SAMPLE1 sample1

TestData \

sample 1No Documentation Found RETURN TABLE ( NumericField ) — SAMPLE2 sample2 TestData \ sample2 No Documentation Found RETURN TABLE ( NumericField ) — **ENCODEDSAMPLE1** encodedSample1 TestData \ encodedSample1 No Documentation Found

# **ENCODEDSAMPLE2** encodedSample2

RETURN TABLE ( NumericField ) —

TestData \

## ${\bf encoded Sample 2}$

No Documentation Found

RETURN TABLE ( NumericField ) -

# TestSplit

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Test data for testing split function

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# TESTDATA testData

testData

Test data for testing split function

#### Children

- 1. sampleData (From testData): No Documentation Found
- 2. trainData (From testData): No Documentation Found
- 3. testData (From testData): No Documentation Found

# **SAMPLEDATA** sampleData

testData \

sampleData

No Documentation Found

RETURN	TABLE	( NumericField	) —
		The state of the s	

# TRAINDATA trainData

testData \

trainData

No Documentation Found

RETURN TABLE ( NumericField ) —

# TESTDATA testData

testData \

testData

No Documentation Found

 $\begin{array}{c} \textbf{RETURN} & \textbf{TABLE ( NumericField )} \\ - \end{array}$ 

# **TestStandardScaler**

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# TESTDATA TestData

**TestData** 

Test data for testing standardScaler module

#### Children

- 1. key (From TestData): No Documentation Found
- 2. sampleData (From TestData): No Documentation Found
- 3. scaledData (From TestData): No Documentation Found

# KEY key

TestData \

key

No Documentation Found

RETURN	TABLE ( KeyLayout )	) —
--------	---------------------	-----

# **SAMPLEDATA** sampleData

TestData \

sampleData

No Documentation Found

RETURN TABLE ( NumericField ) —

# SCALEDDATA scaledData

TestData \

scaledData

No Documentation Found

RETURN TABLE ( NumericField ) —

# ${\bf TestStratifiedSplit}$

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- 1. Layout (From TestData): No Documentation Found
- 2. ds (From TestData): No Documentation Found
- 3. ds4 (From TestData): No Documentation Found
- 4. expTrainData (From TestData): No Documentation Found
- 5. expTrainData4 (From TestData): No Documentation Found
- 6. expTestData (From TestData): No Documentation Found
- 7. expTestData3 (From TestData): No Documentation Found
- 8. expTestData4 (From TestData) : No Documentation Found

# **LAYOUT** Layout

TestData \

Layout

No Documentation Found

FIELD id || UNSIGNED8 — No Doc

FIELD <u>f1</u> ||| UNSIGNED8 — No Doc

FIELD <u>f2</u> ||| UNSIGNED8 — No Doc

FIELD <u>f3</u> || UNSIGNED8 — No Doc

FIELD <u>f4</u> ||| UNSIGNED8 — No Doc

# DS ds

TestData \

ds

No Documentation Found

RETURN TABLE ( Layout ) —

# DS4 ds4

TestData \

ds4

No Documentation Found

RETURN	TABLE	(Layout)	) —
--------	-------	----------	-----

# **EXPTRAINDATA** expTrainData

TestData \

expTrainData

No Documentation Found

RETURN TABLE ( NumericField ) —

# **EXPTRAINDATA4** expTrainData4

TestData \

expTrainData4

No Documentation Found

RETURN TABLE ( NumericField ) —

# **EXPTESTDATA** expTestData

 $TestData \setminus$ 

expTestData

No Documentation Found

RETURN TABLE ( NumericField ) -	
---------------------------------	--

# **EXPTESTDATA3** expTestData3

TestData \

expTestData3

No Documentation Found

# **EXPTESTDATA4** expTestData4

TestData \

expTestData4

No Documentation Found

RETURN TABLE ( NumericField ) —

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# **FILES** Files

**Files** 

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- 1. pathPrefix (From Files): No Documentation Found
- 2. RawDataRec (From Files): No Documentation Found
- 3. rawDataPath (From Files): No Documentation Found
- 4. rawData (From Files): No Documentation Found
- 5. CleanDataRec (From Files): No Documentation Found
- 6. cleanDataPath (From Files): No Documentation Found
- 7. cleanData (From Files): No Documentation Found
- 8. labelEncodedDataRec (From Files): No Documentation Found
- 9. labelEncodedDataPath (From Files): No Documentation Found
- 10. labelEncodedData (From Files): No Documentation Found
- 11. MLDataPath (From Files): No Documentation Found

- 12. MLData (From Files): No Documentation Found
- 13. xTrainPath (From Files): No Documentation Found
- 14. xTrain (From Files): No Documentation Found
- 15. yTrainPath (From Files): No Documentation Found
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- 17. xTestPath (From Files): No Documentation Found
- 18. xTest (From Files): No Documentation Found
- 19. vTestPath (From Files): No Documentation Found
- 20. yTest (From Files): No Documentation Found
- 21. cleanXTrainPath (From Files): No Documentation Found
- 22. cleanXTrain (From Files): No Documentation Found
- 23. cleanXTestPath (From Files): No Documentation Found
- 24. cleanXTest (From Files): No Documentation Found
- 25. PredictionsPath (From Files): No Documentation Found

# PATHPREFIX pathPrefix

Files \

#### pathPrefix

No Documentation Found

RETURN STRING37 —

## **RAWDATAREC** RawDataRec

#### Files \

#### RawDataRec

No Documentation Found

FIELD longitude || STRING — No Doc

FIELD <u>latitude</u> ||| STRING — No Doc

FIELD totalrooms || STRING — No Doc

FIELD totalbedrooms ||| STRING — No Doc

**FIELD** population ||| STRING — No Doc

**FIELD** <u>households</u> ||| STRING — No Doc

FIELD medianincome || STRING — No Doc

FIELD medianhousevalue || STRING — No Doc

FIELD oceanproximity ||| STRING — No Doc

# RAWDATAPATH rawDataPath

## Files \

#### rawDataPath

No Documentation Found

RETURN STRING44 —

# **RAWDATA** rawData

Files \

rawData

No Documentation Found

RETURN TABLE ( RawDataRec ) —

# **CLEANDATAREC** CleanDataRec

Files \

CleanDataRec

No Documentation Found

FIELD id || UNSIGNED8 — No Doc

FIELD longitude ||| REAL4 — No Doc

FIELD <u>latitude</u> ||| REAL4 — No Doc

FIELD housingmedianage ||| REAL4 — No Doc

**FIELD** totalrooms ||| REAL4 — No Doc

FIELD totalbedrooms ||| REAL4 — No Doc

**FIELD population** ||| REAL4 — No Doc

**FIELD** <u>households</u> ||| REAL4 — No Doc

FIELD medianincome ||| REAL4 — No Doc

**FIELD** <u>medianhousevalue</u> ||| REAL8 — No Doc

**FIELD** oceanproximity ||| STRING10 — No Doc

# **CLEANDATAPATH** cleanDataPath

Files \

cleanDataPath

No Documentation Found

RETURN STRING46 —

# **CLEANDATA** cleanData

Files \

cleanData

No Documentation Found

 $\begin{array}{c} \textbf{RETURN} & \textbf{TABLE} \ ( \ \textbf{CleanDataRec} \ ) \ - \end{array}$ 

# LABELENCODEDDATAREC labelEncodedDataRec

Files \

labelEncodedDataRec

No Documentation Found

FIELD id || UNSIGNED8 — No Doc

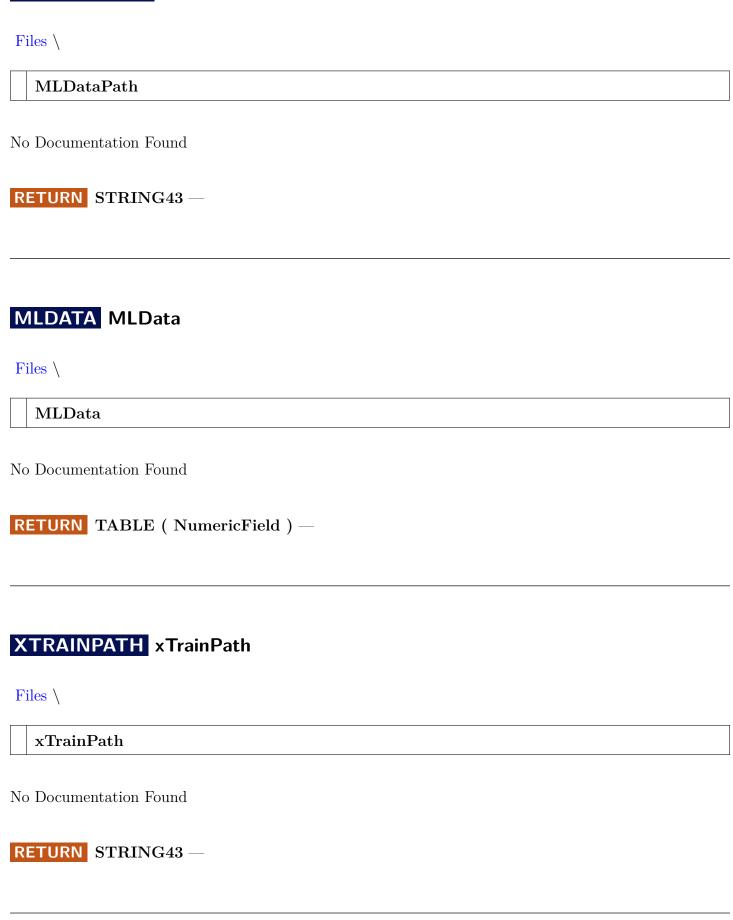
FIELD longitude ||| REAL4 — No Doc

FIELD <u>latitude</u> ||| REAL4 — No Doc

**FIELD** housingmedianage ||| REAL4 — No Doc FIELD totalrooms ||| REAL4 — No Doc **FIELD** totalbedrooms ||| REAL4 — No Doc **FIELD** population ||| REAL4 — No Doc FIELD <u>households</u> ||| REAL4 — No Doc **FIELD** medianincome ||| REAL4 — No Doc FIELD medianhousevalue ||| REAL8 — No Doc FIELD oceanproximity || INTEGER8 — No Doc LABELENCODEDDATAPATH labelEncodedDataPath Files \ labelEncodedDataPath No Documentation Found RETURN STRING53 — LABELENCODEDDATA labelEncodedData Files \ labelEncodedData No Documentation Found

RETURN TABLE ( labelEncodedDataRec ) —

# MLDATAPATH MLDataPath



XTRAIN xTrain
Files \
xTrain
No Documentation Found
${\bf RETURN} \ \ {\bf TABLE} \ ( \ {\bf NumericField} \ ) \$
YTRAINPATH yTrainPath
Files \
yTrainPath
No Documentation Found
RETURN STRING43 —
YTRAIN yTrain
Files \
yTrain
No Documentation Found
RETURN TABLE ( NumericField ) —

# XTESTPATH xTestPath

Files \

xTestPath

No Documentation Found

RETURN STRING42 —

# XTEST xTest

Files \

xTest

No Documentation Found

 $\begin{array}{c} \textbf{RETURN} & \textbf{TABLE ( NumericField )} \\ - \end{array}$ 

# YTESTPATH yTestPath

Files \

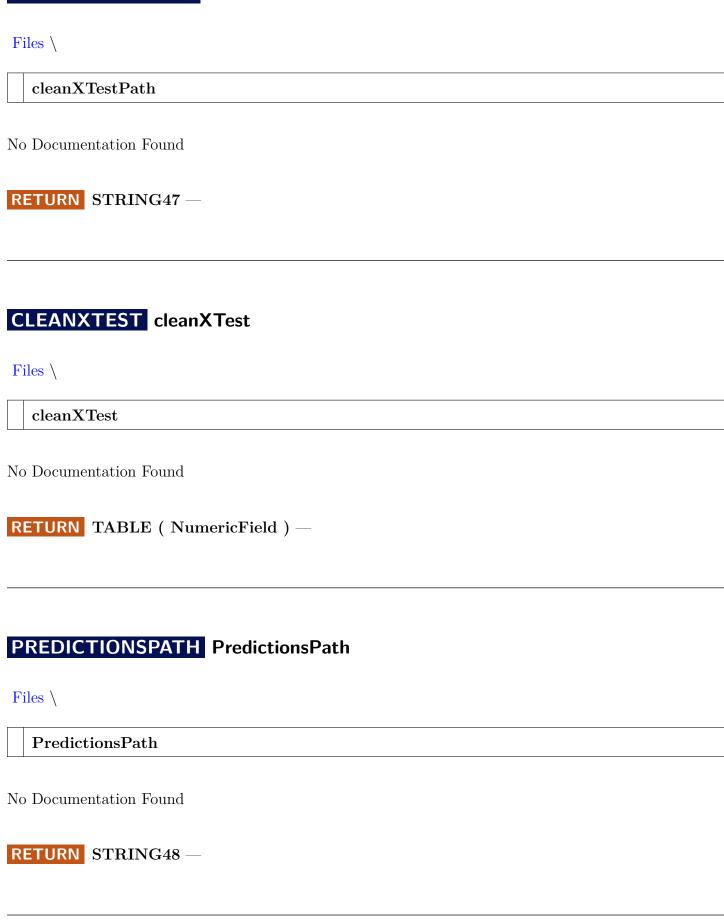
yTestPath

No Documentation Found

RETURN STRING42 —

YTEST yTest
Files \
yTest
No Documentation Found
RETURN TABLE ( NumericField ) —
CLEANXTRAINPATH cleanXTrainPath
Files \
cleanXTrainPath
No Documentation Found
RETURN STRING48 —
CLEANXTRAIN cleanXTrain
Files \
cleanXTrain
No Documentation Found
RETURN TABLE ( NumericField ) —

# **CLEANXTESTPATH** cleanXTestPath



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Merge two Numeric Field datasets ds1 and ds2 by appending ds2 to ds1  $\,$ 

GetCategories.ecl

Allows to extract all the categories of a feature from a given dataset

GetFeatureNames.ecl

Extracts the feature names from some dataset

ResetID.ecl

resets the id sequence so it starts from 1

Shuffle.ecl

shuffles a numericField dataset

 ${\bf Validate Split Input. ecl}$ 

validates input for split function

LabelEncoder

# **AppendNF**

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# **DESCRIPTIONS**

# **APPENDNF** AppendNF

#### AppendNF

(DATASET(NumericField) ds1, DATASET(NumericField) ds2)

Merge two NumericField datasets ds1 and ds2 by appending ds2 to ds1. For example, merge ds1 and ds2 as following:  $ds1 := DATASET(\{[1, 1, 1, 0.5]\}, NumericField); ds2 := DATASET(\{[1, 2, 1, 2.0]\}, NumericField); The result after merging is as below: mergedDs := DATASET(\{[1, 1, 1, 0.5], [1, 2, 2, 2.0]\}, NumericField);$ 

**PARAMETER** ds1: ||| TABLE ( NumericField ) — DATASET(NumericField) The dataset to append to

PARAMETER ds2: || TABLE ( NumericField ) — DATASET(NumericField) The dataset to be appended

**RETURN TABLE ( NumericField )** — the merged dataset with ds2 following ds1

# **GetCategories**

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# **DESCRIPTIONS**

# **GETCATEGORIES** GetCategories

GetCategories

(source, featureName)

Allows to extract all the categories of a feature from a given dataset.

**PARAMETER** source: ||| INTEGER8 — ANY. the dataset from which to extract the categories.

**PARAMETER** <u>featureName:</u> ||| INTEGER8 — STRING. the name of the feature for which to extract the categories.

**RETURN** BOOLEAN — categories: SET OF STRING. the feature's categories.

# **GetFeatureNames**

Go Up

# **DESCRIPTIONS**

# **GETFEATURENAMES** GetFeatureNames

GetFeatureNames
(dta)

Extracts the feature names from some dataset.

Note: complex record structures with child datasets are not handled.

**PARAMETER** <u>dta:</u> ||| INTEGER8 — any dataset. Dataset from which to extract the feature names

**RETURN BOOLEAN** — featureNames: SET OF STRING A set of string holding the feature names.

# ResetID

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# **IMPORTS**

std |

## **DESCRIPTIONS**

# **RESETID** ResetID

/ EXPORT ResetID

(DATASET(NumericField) ds)

resets the id sequence so it starts from 1.

**PARAMETER** <u>ds:</u> ||| TABLE ( NumericField ) — DATASET(NumericField). The dataset with unordered ids.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — dataset with ordered ids.

# $\begin{array}{c} {\bf Preprocessing/\ Utils/} \\ {\bf Shuffle} \end{array}$

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# **IMPORTS**

# **DESCRIPTIONS**

# **SHUFFLE** shuffle

/ EXPORT	shuffle
(DATASET)	(NumericField) dataToShuffle)

shuffles a numericField dataset.

**PARAMETER** <u>dataToShuffle:</u> ||| TABLE ( NumericField ) — DATASET(NumericField). the data to shuffle.

**RETURN** TABLE ( NumericField ) — shuffled data: DATASET(NumericField).

# ValidateSplitInput

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## **IMPORTS**

### **DESCRIPTIONS**

# VALIDATESPLITINPUT validateSplitInput

/ EXPORT	XPORT validateSplitInput					
(DATASET(	NumericField)	dataToSplit,	REAL4	trainSize,	REAL4	testSize)

validates input for split function.

input is valid if data is not empty, train and test sizes are not both zero, sizes are within [0.0, 1.0) with one of them being different from 0 and their sum does not exceed 1.0.

PARAMETER dataToSplit: ||| TABLE ( NumericField ) — DATASET(Types.NumericField). The data to split.

PARAMETER <u>trainSize:</u> ||| REAL4 — REAL4. The training size.

PARAMETER testSize: ||| REAL4 — REAL4. The test size.

**RETURN** STRING — 'Data is empty' if dataToSplit is empty, 'Train size and test sizes are both 0.0' if the sizes are equal to zero, 'Invalid size! valid range = [0.0, 1.0)' if one of the sizes is out of range and 'Sizes are too large! trainSize + testSize > 1.0' if the sum of sizes exceeds 1.0.

# LabelEncoder

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Builds a lookup table that maps each category to a unique number

MapCategoriesToValues.ecl

Builds a lookup table that maps each category of a feature to a unique number

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## **MapAFeatureCategories**

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#### **DESCRIPTIONS**

#### MAPAFEATURECATEGORIES MapAFeatureCategories

#### **MapAFeatureCategories**

(STRING featureName, SET OF STRING unmappedCategories)

Builds a lookup table that maps each category to a unique number. Each category is assigned its index in the category set.

PARAMETER <u>featureName:</u> ||| STRING — STRING. The name of the feature.

 $\begin{array}{c} \textbf{PARAMETER} & \textbf{unmappedCategories:} \\ & \textbf{unmapped categories.} \\ & \textbf{y} \end{array} \\ | | | \text{SET (STRING )} \\ - | \text{SET OF STRING.} \\ < \textbf{p} > \text{The feature's area of the property of the pr$ 

**RETURN TABLE ( mappingLayout )** — categoriesMapping: ROW(MappingLayout). A row the feature name mapped to its categories and each category mapped to its value.

#### Preprocessing/ Utils/ LabelEncoder/

## **MapCategoriesToValues**

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#### **DESCRIPTIONS**

#### MAPCATEGORIESTOVALUES MapCategoriesToValues

MapCategoriesToValues

(key)

Builds a lookup table that maps each category of a feature to a unique number. Each category is assigned its index in the category set.

PARAMETER <u>key:</u> ||| INTEGER8 — DATASET(KeyLayout). Mapping between feature names and categories.

**RETURN BOOLEAN** — categoriesMapping: DATASET(MappingLayout). A table with each feature name mapped to its categories and each category mapped to its value. 
//record mapping a category to its value.
 Category := RECORD
 STRING categoryName;
 INTEGER value;
 END;
//record mapping feature names to their categories.
 MappingLayout := RECORD
 STRING featureName;
 DATASET(Category) categories;
 END;
 END;
 END;
 STRING
 featureName;
 DATASET(Category) categories;

# $\frac{\text{Preprocessing/ Utils/ LabelEncoder/}}{\text{Types}}$

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#### **DESCRIPTIONS**

### **TYPES** Types

**Types** 

Utility Record Structures for LabelEncoder Module

#### Children

- 1. Category (From Types): No Documentation Found
- 2. MappingLayout (From Types): No Documentation Found
- 3. LabelLayout (From Types): No Documentation Found

#### **CATEGORY** Category

Types \

Category

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**FIELD** categoryname ||| STRING — No Doc

FIELD <u>value</u> || INTEGER8 — No Doc

## MAPPINGLAYOUT MappingLayout

Types \

MappingLayout

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FIELD <u>featurename</u> ||| STRING — No Doc

**FIELD** categories ||| TABLE ( Category ) — No Doc

## LABELLAYOUT LabelLayout

Types \

LabelLayout

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FIELD <u>label</u> ||| STRING — No Doc

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generate.ecl
test_appends.ecl
test_discrete.ecl
to_from.ecl

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### **DESCRIPTIONS**

## **FIELD\_AGGREGATES** field\_aggregates

field\_aggregates

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## **DESCRIPTIONS**

## **GENERATE** generate

generate

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# ${ m ^{Tests/}}$ ${ m test\_appends}$

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TEST\_APPENDS test\_appends

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# ${ m _{Tests/}} \ { m test\_discrete}$

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### **DESCRIPTIONS**

## TEST\_DISCRETE test\_discrete

; EXPORT test\_discrete

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# ${f to\_from}$

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## TO\_FROM to\_from

to\_from

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Make a sparse NumericField dataset dense by filling in missing values

#### FatD.ecl

Make a sparse DiscreteField dataset dense by filling in missing values

#### Gini.ecl

Create a file of pivot/target pairs with a Gini impurity value

#### SequenceInField.ecl

Assign sequence numbers within groups for a dataset

## Utils/ Fat

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#### **IMPORTS**

Types |

#### **DESCRIPTIONS**

### **FAT** Fat

```
/ EXPORT DATASET(Types.NumericField) Fat

(DATASET(Types.NumericField) d0,
Types.t_FieldReal v=0)
```

Make a sparse Numeric Field dataset dense by filling in missing values. All empty cells are set to the designated value.

**PARAMETER** <u>d0</u> ||| TABLE ( NumericField ) — They NumericField dataset to be filled.

**PARAMETER**  $\underline{\mathbf{v}} \parallel \parallel \text{REAL8}$  — The value to assign missing records.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , REAL8 value } ) — A full NumericField dataset with every field populated.

## $\mathbf{FatD}$

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#### **IMPORTS**

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#### **DESCRIPTIONS**

#### FATD FatD

```
/ EXPORT DATASET(Types.DiscreteField) FatD

(DATASET(Types.DiscreteField) d0,
Types.t_Discrete v=0)
```

Make a sparse DiscreteField dataset dense by filling in missing values. All empty cells are set to the designated value.

PARAMETER <u>d0</u> || TABLE ( DiscreteField ) — The DiscreteField dataset to be filled.

**PARAMETER**  $\underline{\mathbf{v}}$  ||| INTEGER4 — The value to assign missing records.

RETURN TABLE ( { UNSIGNED2 wi , UNSIGNED8 id , UNSIGNED4 number , INTEGER4 value } ) — A full DiscreteField dataset with every field populated.

## Utils/ Gini

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#### **DESCRIPTIONS**

#### **GINI** Gini

```
/ EXPORT Gini
(infile, pivot, target, wi_name='wi')
```

Create a file of pivot/target pairs with a Gini impurity value.

PARAMETER <u>infile</u> ||| INTEGER8 — the input file, any type with a work item field.

PARAMETER pivot || INTEGER8 — the name of the pivot field.

PARAMETER target || INTEGER8 — the name of the field used as the target.

**PARAMETER** wi\_name ||| INTEGER8 — the name of the work item field, default is "wi".

**RETURN BOOLEAN** — A table by Work Item and Pivot value giving count and Gini impurity value.

# Utils/ SequenceInField

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#### **DESCRIPTIONS**

#### **SEQUENCEINFIELD** SequenceInField

/ EXPORT	SequenceInField
(infile,infield,seq,wi_name='wi')	

Assign sequence numbers within groups for a dataset. Given a file (dataset) which is sorted by the work item identifier and INFIELD (and possibly other values), add sequence numbers within the range of each infield. Slighly elaborate code is to avoid having to partition the data to one value of infield per node and to work with very large numbers of records where a global count project would be inappropriate. This is useful for assigning rank positions with the groupings.

PARAMETER infile || INTEGER8 — the input file, any type.

PARAMETER infield ||| INTEGER8 — field name of grouping field.

PARAMETER seq || INTEGER8 — name of the field to receive the sequence number.

PARAMETER wi\_name || INTEGER8 — work item field name, default is wi.

**RETURN** BOOLEAN — a file of the same type with sequence numbers applied.