

Package ‘analysisPipelines’

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Title Compose Interoperable Analysis Pipelines & Put Them in Production

Version 1.0.2

Description Enables data scientists to compose pipelines of analysis which consist of data manipulation, exploratory analysis & reporting, as well as modeling steps. Data scientists can use tools of their choice through an R interface, and compose interoperable pipelines between R, Spark, and Python.

Credits to Mu Sigma for supporting the development of the package.

Note - To enable pipelines involving Spark tasks, the package uses the 'SparkR' package.

The SparkR package needs to be installed to use Spark as an engine within a pipeline. SparkR is distributed natively with Apache Spark and is not distributed on CRAN. The SparkR version needs to directly map to the Spark version (hence the native distribution), and care needs to be taken to ensure that this is configured properly.

To install SparkR from Github, run the following command if you know the Spark version: `devtools::install_github('apache/spark@v2.x.x', subdir='R/pkg')`.

The other option is to install SparkR by running the following terminal commands

if Spark has already been installed: `'$ export SPARK_HOME=/path/to/spark/directory && cd $SPARK_HOME/R/lib/SparkR/ && R -e ``devtools::install('.')'`.

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Suggests plotly, knitr, rmarkdown, parallel, visNetwork, rjson, DT, shiny, R.devices, corrplot, car, foreign

Enhances SparkR, reticulate

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AnalysisPipeline-class

Class for constructing Analysis Pipelines for batch/ one-time analyses

Description

Class for constructing Analysis Pipelines for batch/ one-time analyses

Details

Inherits the base class [BaseAnalysisPipeline](#) class which holds the metadata including the registry of available functions, the data on which the pipeline is to be applied, as well as the pipeline itself

Additionally, this class is meant to be used for batch/ one-time processing. Contains additional slots to hold the data frame to be used for the pipeline and associated schema

Slots

input The input dataset on which analysis is to be performed

originalSchemaDf Empty data frame representing the schema of the input

See Also

Other Package core functions for batch/one-time analyses: [checkSchema](#), [generateReport](#), [initialize](#), [BaseAnalysisPipeline](#)

analysisPipelines	<i>analysisPipelines</i>
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Description

The package aims at enabling data scientists to compose pipelines of analysis which consist of data manipulation, exploratory analysis & reporting, as well as modeling steps. It also aims to enable data scientists to use tools of their choice through an R interface, and compose interoperable pipelines between R, Spark, and Python. Credits to Mu Sigma for supporting the development of the package.

Note

To enable pipelines involving Spark tasks, the package uses the 'SparkR' package. Using Spark as an engine requires the SparkR package to be installed. SparkR is distributed natively with Apache Spark and is not distributed on CRAN. The SparkR version needs to directly map to the Spark version (hence the native distribution), and care needs to be taken to ensure that this is configured properly. To install from Github, run the following command, if you know the Spark version:

- `devtools::install_github('apache/spark@v2.x.x', subdir='R/pkg')`

The other option is to install SparkR by running the following terminal commands if Spark has already been installed:

- `$ export SPARK_HOME=/path/to/spark/directory`
- `$ cd $SPARK_HOME/R/lib/SparkR/`
- `$ R -e "devtools::install('.')"`

assessEngineSetUp	<i>Assesses engine (R, Spark, Python, Spark Structured Streaming) set up</i>
-------------------	--

Description

Assesses engine (R, Spark, Python, Spark Structured Streaming) set up

Usage

```
assessEngineSetUp(object)

## S4 method for signature 'BaseAnalysisPipeline'
assessEngineSetUp(object)
```

Arguments

object A Pipeline object

Details

Assesses whether engines required for executing functions in an AnalysisPipeline or StreamingAnalysisPipeline object have been set up

This method is implemented on the base class as it is a shared functionality across Pipeline objects

Value

Tibble containing the details of available engines, whether they are required for a pipeline, a logical value reporting whether the engine has been set up, and comments.

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>% univarCatDistPlots(uniCol = "Species", priColor = "blue",
  optionalPlots = 0) %>% assessEngineSetUp

## End(Not run)
```

BaseAnalysisPipeline-class

Base class for AnalysisPipeline and StreamingAnalysisPipeline objects

Description

Base class for AnalysisPipeline and StreamingAnalysisPipeline objects

Details

The class which holds the metadata including the registry of available functions, the data on which the pipeline is to be applied, as well as the pipeline itself, and serves as the base class for various types of Pipeline objects such as Batch and Streaming.

This base class which contains the slots related to the registry, pipeline and output can be extended to create custom class for specific scenarios if required.

In the documentation, objects of classes which are subclasses of this class are referred to as 'Pipeline' objects

Slots

pipeline A tibble which holds functions to be called

pipelineExecutor A list containing details of the execution, such as topological ordering of functions to be executed, dependency map of functions, as well as logger configuration

output A list which holds all the functions output

See Also

Other Package core functions: [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize,BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

bivarPlots

Bi-Variate Plot

Description

Bi-Variate Plot

Usage

```
bivarPlots(dataset, select_var_name_1, select_var_name_2,
  priColor = "blue", secColor = "black")
```

Arguments

dataset	the dataframe that needs to be loaded
select_var_name_1	the name of first column on which the plot needs to be generated
select_var_name_2	the name of second column on which the plot needs to be generated
priColor	the primary color for the plots
secColor	A secondary color for the plots

Details

A bivariate distribution graph on the selected columns from the dataframe. Selected two columns are on two axis' and a plot is generated

Value

Bivariate plot

See Also

Other Package EDA Utilites functions: [CheckColumnType](#), [correlationMatPlot](#), [getDatatype](#), [ignoreCols](#), [multiVarOutlierPlot](#), [outlierPlot](#), [univarCatDistPlots](#)

Examples

```
bivarPlots(dataset = iris, select_var_name_1 = "Sepal.Length",
           select_var_name_2 = "Sepal.Width")
```

castKafkaStreamAsString

Connect to a Spark session

Description

Connect to a Spark session

Usage

```
castKafkaStreamAsString(streamObj)
```

Arguments

streamObj	Spark Structured Streaming DataFrame returned by read.stream function with source = 'kafka'
-----------	---

Details

Takes in a Structured Stream from Kafka created from read.stream(source = 'kafka', ...) and returns a Structured Streaming DataFrame where the key and value from the Kafka stream are cast to string

Value

Updated Spark Structured Streaming DataFrame with key, value, topic and timestamp from the Kafka stream

See Also

Other Spark utilities: [convertKafkaValueFromJson](#), [sparkRSessionCreateIfNotPresent](#)

CheckColumnType	<i>Check for type of column</i>
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Description

Check for type of column

Usage

```
CheckColumnType(dataVector)
```

Arguments

dataVector a data vector of a column

Details

Checking for type of columns in the datavector

Value

column Type

See Also

Other Package EDA Utilites functions: [bivarPlots](#), [correlationMatPlot](#), [getDatatype](#), [ignoreCols](#), [multiVarOutlierPlot](#), [outlierPlot](#), [univarCatDistPlots](#)

Examples

```
CheckColumnType(iris$Sepal.Length)
```

checkSchemaMatch	<i>Checks the schema of the input to a Pipeline object against the original</i>
------------------	---

Description

Checks the schema of the input to a Pipeline object against the original

Usage

```
checkSchemaMatch(object, newData)
```

```
## S4 method for signature 'AnalysisPipeline'
checkSchemaMatch(object, newData)
```


Arguments

object	A Pipeline object
newData	The newData that the pipeline is to be initialized with

Details

Checks the schema of the new data frame that the pipeline is to be initialized with against the original schema that the pipeline was saved with. Provides a detailed comparison

Value

Returns a list with details on added columns, removed columns, comparison between column classes, and a logical whether the schema has remained the same from the old dataframe to the new one

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

convertKafkaValueFromJson

Connect to a Spark session

Description

Connect to a Spark session

Usage

```
convertKafkaValueFromJson(streamObj, schema)
```

Arguments

streamObj	Spark Structured Streaming DataFrame which is returned by the <code>castKafkaStreamAsString</code> function
schema	A <code>structType</code> object created from SparkR specifying the schema of the json data present in the <code>value</code> attribute of the incoming Kafka stream

Details

Takes in a Structured Stream from Kafka created from `read.stream(source = 'kafka', ...)` and returns a Structured Streaming DataFrame where the key and value from the Kafka stream are cast to string

Value

Spark Structured Streaming DataFrame with the json data in the value attribute of the Kafka stream parsed into a DataFrame format

See Also

Other Spark utilities: [castKafkaStreamAsString](#), [sparkRSessionCreateIfNotPresent](#)

correlationMatPlot *Correlation Matrix Plot*

Description

A correlation matrix is created and plotted across all the columns in the dataset

Usage

```
correlationMatPlot(dataset, methodused = "everything")
```

Arguments

dataset	the dataset that needs to be loaded
methodused	methods to be used for computing correlation

Value

Correlation Matrix graph

See Also

Other Package EDA Utilites functions: [CheckColumnType](#), [bivarPlots](#), [getDatatype](#), [ignoreCols](#), [multiVarOutlierPlot](#), [outlierPlot](#), [univarCatDistPlots](#)

Examples

```
correlationMatPlot(dataset = iris)
```

`createPipelineInstance`*Create a Pipeline object from a meta-pipeline*

Description

Create a Pipeline object from a meta-pipeline

Usage

```
createPipelineInstance(metaPipelineObj, newParams)
```

```
## S4 method for signature 'MetaAnalysisPipeline'  
createPipelineInstance(metaPipelineObj,  
  newParams)
```

Arguments

`metaPipelineObj`

A `MetaAnalysisPipeline` object

`newParams`

Either a nested named list containing all the functions in the pipeline, their arguments and corresponding values (OR) an object of class `proto` which is a pipeline prototype, with the new values of the arguments set. Refer the `getPipelinePrototype` method.

Details

This method instantiates a Pipeline object (both `AnalysisPipeline` and `StreamingAnalysisPipeline`) from a meta-pipeline as well as an object containing the new set of values for the arguments of all the functions in the pipeline.

Value

A Pipeline object

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% univarCatDistPlots(uniCol = "Species") -> pipelineObj
pipelineObj %>>% exportAsMetaPipeline -> exportedMetaPipeline
exportedMetaPipeline %>>%
  createPipelineInstance(newParams = exportedMetaPipeline %>>%
    getPipelinePrototype)

## End(Not run)
```

exportAsMetaPipeline *Method to export a meta-pipeline*

Description

Method to export a meta-pipeline

Usage

```
exportAsMetaPipeline(object)

## S4 method for signature 'BaseAnalysisPipeline'
exportAsMetaPipeline(object)
```

Arguments

object A Pipeline object

Details

This method exports a Pipeline object i.e. of the classes `AnalysisPipeline` or `StreamingAnalysisPipeline` as a meta-pipeline

Value

an object of class "MetaAnalysisPipeline"

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
#' pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% univarCatDistPlots(uniCol = "Species") %>>%
  exportAsMetaPipeline -> exportedMetaPipeline

## End(Not run)
```

generateOutput	<i>Generate a list of outputs from Pipeline objects</i>
----------------	---

Description

Generate a list of outputs from Pipeline objects

Usage

```
generateOutput(object)

## S4 method for signature 'AnalysisPipeline'
generateOutput(object)

## S4 method for signature 'StreamingAnalysisPipeline'
generateOutput(object)
```

Arguments

object object that contains input, pipeline, registry and output

Details

generateOutput is a generic function that is implemented for various types of pipeline objects such as AnalysisPipeline and StreamingAnalysisPipeline

The sequence of operations stored in the pipeline object are run and outputs generated, stored in a list

Value

Updated Pipeline object with the outputs at each step stored in the output slot.

Specific outputs can be obtained by using the [getOutputById](#) function

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize,BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

 generateReport

Generate a HTML report from an AnalysisPipeline object

Description

Generate a HTML report from an AnalysisPipeline object

Usage

```
generateReport(object, path)
```

```
## S4 method for signature 'AnalysisPipeline,character'
generateReport(object, path = ".")
```

Arguments

object	object that contains input, pipeline, registry and output
path	path on the file system, where the generated html report should be stored

Details

The sequence of operations stored in the AnalysisPipeline object are run, outputs generated, and a HTML report is generated with outputs in the same sequence as the pipeline created by the user

Value

Updated AnalysisPipeline object

See Also

Other Package core functions for batch/one-time analyses: [AnalysisPipeline-class](#), [checkSchema](#), [initialize,BaseAnalysisPipeline-method](#)

Examples

```
## Not run:
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% univarCatDistPlots(uniCol = "Species", storeOutput = T) -> pipelineObj
pipelineObj %>>% generateReport(path = ".")

## End(Not run)
```

genericPipelineException

Default exception for pipeline functions

Description

Default exception for pipeline functions

Usage

```
genericPipelineException(error)
```

Arguments

error Error encountered during the execution of a particular pipeline function

Details

This functions defines the default function which will be called in case of an exception occurring while executing any of the pipeline functions. While a function is registered, a custom function to deal with exceptions incurred during the call of the function being registered can be passed by the user. If passed, the custom function will be called instead of this function

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

`getDatatype`*Get Data Type*

Description

Get Data Type

Usage

```
getDatatype(dataset)
```

Arguments

`dataset` a dataset which needs to be loaded

Details

Based on the datatype the columns are separated into categorical and numerical columns

Value

list with `numeric_cols` and `cat_cols`

See Also

Other Package EDA Utilites functions: [CheckColumnType](#), [bivarPlots](#), [correlationMatPlot](#), [ignoreCols](#), [multiVarOutlierPlot](#), [outlierPlot](#), [univarCatDistPlots](#)

Examples

```
getDatatype(iris)
```

`getFeaturesForPyClassification`*Extracts selected columns from a data frame as a Python array*

Description

Extracts selected columns from a data frame as a Python array

Usage

```
getFeaturesForPyClassification(dataset, featureNames)
```


Arguments

dataset an R data frame
 featureNames Column names to be extracted from the R data frames. A character vector.

Details

Helper function, which when provided an R data frame and a set of column/ feature names, extracts them from the R data frame as a matrix and converts them to the equivalent Python array.

Typically this function can be used when providing a feature matrix to a Python machine learning function

See Also

Other R helper utilities for Python: [getTargetForPyClassification](#), [setPythonEnvir](#)

Examples

```
## Not run:
getFeaturesForPyClassification(dataset = iris,
  featureNames = c("Sepal.Length", "Sepal.Width"))

## End(Not run)
```

<code>getInput</code>	<i>Obtains the initializedInput</i>
-----------------------	-------------------------------------

Description

Obtains the initializedInput

Usage

```
getInput(object)

## S4 method for signature 'BaseAnalysisPipeline'
getInput(object)
```

Arguments

object The AnalysisPipeline or StreamingAnalysisPipeline object

Details

Obtains the input from the AnalysisPipeline or StreamingAnalysisPipeline object

This method is implemented on the base class as it is a shared functionality types of Analysis Pipelines which extend this class

Value

Dataframe for an AnalysisPipeline & SparkDataFrame for a StreamingAnalysisPipeline

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% getInput
```

getLoggerDetails	<i>Obtains the logger configuration for the pipeline</i>
------------------	--

Description

Obtains the logger configuration for the pipeline

Usage

```
getLoggerDetails(object)

## S4 method for signature 'BaseAnalysisPipeline'
getLoggerDetails(object)
```

Arguments

object A Pipeline object

Details

This function obtains the logger configuration for the pipeline.

Value

Logger configuration as a list

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% getLoggerDetails
```

getOutputById	<i>Obtains a specific output</i>
---------------	----------------------------------

Description

Obtains a specific output

Usage

```
getOutputById(object, reqId, includeCall = F)

## S4 method for signature 'BaseAnalysisPipeline'
getOutputById(object, reqId,
  includeCall = F)
```

Arguments

object	The AnalysisPipeline or StreamingAnalysisPipeline object
reqId	The position of the function for which the output is desired in the sequence of operations in the pipeline.
includeCall	Logical which defines whether the call used to generate the output should be returned. By default this is false

Details

Obtains a specific output from the AnalysisPipeline or StreamingAnalysisPipeline object by passing the position of the function for which the output is desired, in the sequence of operations in the pipeline. This can be obtained by passing the number under the 'id' column in the pipeline table corresponding to the required function

This method is implemented on the base class as it is a shared functionality types of Analysis Pipelines which extend this class

Value

If includeCall = F, the output object generated by the function is returned

If includeCall = T, it is a list containing two elements - call: tibble with 1 row containing the function call for the output desired - output: output generated

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize,BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
getNumRows <- function(dataset){
  return(nrow(dataset))
}
registerFunction("getNumRows")
pipelineObj %>>% getNumRows(storeOutput = TRUE) -> pipelineObj
pipelineObj %>>% generateOutput %>>% getOutputById("1")

## End(Not run)
```

getPipeline

Obtain the pipeline

Description

Obtain the pipeline

Usage

```
getPipeline(object)
```

```
## S4 method for signature 'BaseAnalysisPipeline'
getPipeline(object)
```

Arguments

object The AnalysisPipeline or StreamingAnalysisPipeline object

Details

Obtains the pipeline from the AnalysisPipeline or StreamingAnalysisPipeline object as a tibble

This method is implemented on the base class as it is a shared functionality types of Analysis Pipelines which extend this class

Value

Tibble describing the pipeline

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
getNumRows <- function(dataset){
  return(nrow(dataset))
}
registerFunction("getNumRows")
pipelineObj %>>% getNumRows %>>% getPipeline

## End(Not run)
```

getPipelinePrototype *Obtain the prototype of the functions in the pipeline*

Description

Obtain the prototype of the functions in the pipeline

Usage

```
getPipelinePrototype(metaPipelineObj)

## S4 method for signature 'MetaAnalysisPipeline'
getPipelinePrototype(metaPipelineObj)
```

Arguments

metaPipelineObj
A MetaAnalysisPipeline object

Details

This method returns the prototype of functions in the pipeline and their respective arguments as proto object. Functions in the pipeline can be accessed easily by using the '\$' operator, and within the functions the arguments can be accessed the same way. These can be accessed and set to new values. This pipeline prototype can then be passed to the createPipelineInstance method which will instantiate an executable pipeline with the inputs set in the prototype

Value

An object of class proto from the 'proto' package

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% univarCatDistPlots(uniCol = "Species") %>>%
  exportAsMetaPipeline %>>% getPipelinePrototype

## End(Not run)
```

getRegistry

Obtains the function registry

Description

Obtains the function registry

Usage

```
getRegistry()
```

Details

Obtains the function registry as a tibble, including both predefined and user defined functions

Value

Tibble describing the registry

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
getRegistry()
```

getResponse	<i>Obtains the response term from the formula</i>
-------------	---

Description

Obtains the response term from the formula

Usage

```
getResponse(f)
```

Arguments

f formula from which term is to be extracted.

Details

This is a helper function to extract the response variable from a formula

Value

The response variable in the formula as a string

Examples

```
library(analysisPipelines)
getResponse(y ~ x1 + x2)
```

getTargetForPyClassification

Extracts selected column from a data frame a binary class Python array

Description

Extracts selected column from a data frame a binary class Python array

Usage

```
getTargetForPyClassification(dataset, targetVarName, positiveClass)
```

Arguments

dataset	an R data frame
targetVarName	Name of the target variable for classification. Should be a categorical variable.
positiveClass	Name of the class of the target variable which should be coded as '1'

Details

Helper function, which when provided an R dataframe and a binary categorical column, extracts it from the R data frame, converts it to 1/0 class coding, and converts it to a Python array

Typically this function can be used to extract a target variable for a classifier to be provided to a Python machine learning function

See Also

Other R helper utilities for Python: [getFeaturesForPyClassification](#), [setPythonEnvir](#)

Examples

```
## Not run:  
getTargetForPyClassification(dataset = iris,  
  targetVarName = "Species", positiveClass = "setosa")  
  
## End(Not run)
```

getTerm	<i>Obtains the dependency term from the formula</i>
---------	---

Description

Obtains the dependency term from the formula

Usage

```
getTerm(f)
```

Arguments

f formula from which term is to be extracted.

Details

This is a helper function to extract the terms from a formula

Value

String with the terms

Examples

```
library(analysisPipelines)
getTerm(y ~ x)
```

ignoreCols	<i>Ignores the columns in the loaded dataframe object</i>
------------	---

Description

Ignores the columns in the loaded dataframe object

Usage

```
ignoreCols(data, columns)
```

Arguments

data the dataframe object that needs to be loaded
columns the names of columns to be ignored from dataframe object

Details

The columns selected are removed from the object

Value

Updated dataframe object

See Also

Other Package EDA Utilites functions: [CheckColumnType](#), [bivarPlots](#), [correlationMatPlot](#), [getDatatype](#), [multiVarOutlierPlot](#), [outlierPlot](#), [univarCatDistPlots](#)

Examples

```
ignoreCols(data = iris, columns = "Species")
```

isDependencyParam	<i>Checks if the parameter is the dependency parameter</i>
-------------------	--

Description

Checks if the parameter is the dependency parameter

Usage

```
isDependencyParam(f)
```

Arguments

f formula from which term is to be extracted.

Details

This is a helper function to check if the formula provided is a dependency parameter, as per the package's formula semantics, capturing function dependencies

Value

Logical as to whether it is a dependency parameter

Examples

```
library(analysisPipelines)
isDependencyParam(~f1)
```

loadMetaPipeline	<i>Load a meta-pipeline</i>
------------------	-----------------------------

Description

Load a meta-pipeline

Usage

```
loadMetaPipeline(path)
```

Arguments

path the path at which the .Rds file containing the pipeline is located

Details

This function loads a meta-pipeline from a file system, and returns the meta-pipeline object, which can be assigned to an object in the environment.

Note - When a meta-pipeline is loaded, the existing registry is overwritten with the registry saved with the meta-pipeline

Value

An MetaAnalysisPipeline object

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-methods](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:  
loadMetaPipeline(path = "./metaPipeline.RDS")  
  
## End(Not run)
```

loadPipeline	<i>Loads the AnalysisPipeline or StreamingAnalysisPipeline object from the file system</i>
--------------	--

Description

Loads the AnalysisPipeline or StreamingAnalysisPipeline object from the file system

Usage

```
loadPipeline(path, input = data.frame(), filePath = "")
```

Arguments

path	the path at which the .Rds file containing the pipeline is located
input	(optional) data frame with which the pipeline object should be initialized
filePath	(optional) path where a dataset in .CSV format is present which is to be loaded

Details

The AnalysisPipeline or StreamingAnalysisPipeline object is loaded into the file system from the file system based on the path specified.

Optionally, the input parameter can be provided to initialize the AnalysisPipeline or StreamingAnalysisPipeline object with an R data frame or Streaming Spark DataFrame (in case of StreamingAnalysisPipeline object) present in the R session.

Another provided option, is to specify a filePath where the input dataset is present (in a .CSV format) and the object will be initialized with this data frame. The filePath parameter takes precedence over input parameter. This is applicable only from AnalysisPipeline objects

Note - When a pipeline is loaded, the existing registry is overwritten with the registry saved with the pipeline

Value

An AnalysisPipeline or StreamingAnalysisPipeline object, optionally initialized with the data frame provided

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize,BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:  
library(analysisPipelines)  
loadPipeline(path = "../pipeline.RDS")  
  
## End(Not run)
```

```
loadPredefinedFunctionRegistry  
    Loading the registry of predefined functions
```

Description

Loading the registry of predefined functions

Usage

```
loadPredefinedFunctionRegistry()
```

Details

Loads the registry of predefined functions

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-methods](#), [loadMetaPipeline](#), [loadPipeline](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:  
library(analysisPipelines)  
loadPredefinedFunctionRegistry()  
  
## End(Not run)
```

loadRegistry	<i>Loads a function registry from a file</i>
--------------	--

Description

Loads a function registry from a file

Usage

```
loadRegistry(path)
```

Arguments

path path on the file system, where the registry is to be loaded from

Details

This function loads a function registry and associated function definition stored in an RDS file into the environment. The existing registry is overwritten with the newly loaded registry

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
  library(analysisPipelines)
  loadRegistry(path = "../registry.RDS")

## End(Not run)
```

MetaAnalysisPipeline-class

Class for creating and working with meta-pipelines

Description

Class for creating and working with meta-pipelines

Details

This class works with the `AnalysisPipeline` and `StreamingAnalysisPipeline` classes, and allows the pipeline to be exported as meta-pipeline. A meta-pipeline is a construct, where the input dataset as well as the arguments to functions in the pipeline are not defined. Only the analysis flow and dependencies are stored.

Slots

`pipeline` A tibble which holds functions to be called in the pipeline

`pipelinePrototype` An object of class `proto` from the 'proto' package which maintains the prototype of the functions in the pipeline and their respective arguments

`type` A string defining whether it is a batch or streaming pipeline. Acceptable values are 'batch' & 'streaming'

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

`multiVarOutlierPlot` *Multi-Variate Outlier Plot*

Description

Multi-Variate Outlier Plot

Usage

```
multiVarOutlierPlot(data, depCol, indepCol, sizeCol, priColor = "blue",
  optionalPlots = 0, cutoffValue = 0.05)
```

Arguments

<code>data</code>	the dataframe that needs to be loaded
<code>depCol</code>	the name of column which is to be identified as dependent column
<code>indepCol</code>	the name of an independent column
<code>sizeCol</code>	the name of column used to define the size of point in plots
<code>priColor</code>	the primary color for the plots
<code>optionalPlots</code>	A Flag for optional plots
<code>cutoffValue</code>	A p-value cutoff for detecting outliers

Details

Multivariate outlier plot using the selected columns from the dataframe

Value

Outliers plot

See Also

Other Package EDA Utilities functions: [CheckColumnType](#), [bivarPlots](#), [correlationMatPlot](#), [getDatatype](#), [ignoreCols](#), [outlierPlot](#), [univarCatDistPlots](#)

Examples

```
## Not run:
multiVarOutlierPlot(data = iris, depCol = "Sepal.Length",
  indepCol = "Sepal.Width", sizeCol = "Petal.Length")

## End(Not run)
```

outlierPlot	<i>Outlier detection plot</i>
-------------	-------------------------------

Description

Outlier detection plot

Usage

```
outlierPlot(data, method = "iqr", columnName, cutoffValue = 0.05,
  priColor = "blue", optionalPlots = 0)
```

Arguments

data	the dataframe that needs to be loaded
method	the method on which outliers are to be identified
columnName	the name of column for which the outliers are identified
cutoffValue	the cut off value to define the threshold for outliers
priColor	the primary color for the plots
optionalPlots	A Flag for optional plots

Details

Outlier are to be identified on the selected column from the dataframe

Value

Outliers plot object

See Also

Other Package EDA Utilites functions: [CheckColumnType](#), [bivarPlots](#), [correlationMatPlot](#), [getDatatype](#), [ignoreCols](#), [multiVarOutlierPlot](#), [univarCatDistPlots](#)

Examples

```
## Not run:
outlierPlot(data = iris, columnName = "Sepal.Length")

## End(Not run)
```

prepExecution	<i>Prepare the pipeline for execution</i>
---------------	---

Description

Prepare the pipeline for execution

Usage

```
prepExecution(object)

## S4 method for signature 'BaseAnalysisPipeline'
prepExecution(object)
```

Arguments

object A Pipeline object

Details

The pipeline is prepared for execution by identifying the graph of the pipeline as well as its topological ordering, and dependency map in order to prepare for execution

Value

Updated AnalysisPipeline StreamingAnalysisPipeline object

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize,BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% univarCatDistPlots(uniCol = "Species",
  priColor = "blue", optionalPlots = 0, storeOutput = T) %>>%
  prepExecution -> pipelineObj

## End(Not run)
```

registerFunction	<i>Register a user-defined function to be used with a AnalysisPipeline or StreamingAnalysisPipeline object</i>
------------------	--

Description

Register a user-defined function to be used with a AnalysisPipeline or StreamingAnalysisPipeline object

Usage

```
registerFunction(functionName, heading = "", functionType = "batch",
  engine = "r",
  exceptionFunction = as.character(substitute(genericPipelineException)),
  isDataFunction = T, firstArgClass = "", loadPipeline = F,
  userDefined = T)
```

Arguments

functionName	name of function to be registered
heading	heading of that section in report
functionType	type of function - 'batch' for AnalysisPipeline objects, 'streaming' for StreamingAnalysisPipeline objects
engine	specifies which engine the function is to be run on. Available engines include "r", "spark", and "python"
exceptionFunction	R object corresponding to the exception function
isDataFunction	logical parameter which defines whether the function to be registered operates on data i.e. the first parameter is a dataframe
firstArgClass	character string with the class of the first argument to the function, if it is a non-data function
loadPipeline	logical parameter to see if function is being used in loadPipeline or not. This is for internal working
userDefined	logical parameter defining whether the function is user defined. By default, set to true

Details

The specified operation along with the heading and engine details is stored in the registry, after which it can be added to a pipeline.

If the function already exists in the registry, registration will be skipped. In order to change the definition, the function needs to be reassigned in the Global Environment and then the registerFunction called again.

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
library(analysisPipelines)
getNumRows <- function(dataset){
  return(nrow(dataset))
}

registerFunction("getNumRows")

## End(Not run)
```

savePipeline	<i>Saves the AnalysisPipeline or StreamingAnalysisPipeline object to the file system without outputs</i>
--------------	--

Description

Saves the AnalysisPipeline or StreamingAnalysisPipeline object to the file system without outputs

Usage

```
savePipeline(object, path)

## S4 method for signature 'BaseAnalysisPipeline'
savePipeline(object, path)

## S4 method for signature 'MetaAnalysisPipeline'
savePipeline(object, path)
```

Arguments

object	object that contains input, pipeline, registry and output
path	the path at which the .Rda file containing the pipeline should be stored, along with the name of the file including a .Rda extension

Details

The AnalysisPipeline or StreamingAnalysisPipeline object is saved to the file system in the paths specified

This method is implemented on the base class as it is a shared functionality types of Analysis Pipelines which extend this class

Value

Does not return a value

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% savePipeline(path = "./test.RDS")

## End(Not run)
```

saveRegistry

Saves the registry to the file system

Description

Saves the registry to the file system

Usage

```
saveRegistry(path)
```

Arguments

path path on the file system, where the registry is to be saved to

Details

This function saves the existing function registry and associated function definition loaded in the environment into a file.

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [setInput](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
## Not run:
  library(analysisPipelines)
  saveRegistry(path = "./registry.RDS")

## End(Not run)
```

setInput	<i>Sets the input for an AnalysisPipeline or StreamingAnalysisPipeline object</i>
----------	---

Description

Sets the input for an AnalysisPipeline or StreamingAnalysisPipeline object

Usage

```
setInput(object, input, filePath = "")

## S4 method for signature 'BaseAnalysisPipeline'
setInput(object, input, filePath = "")
```

Arguments

object	object that contains input, pipeline, registry and output
input	the input data frame
filePath	path to the file which needs to be read (currently supports .csv files)

Details

Assigns the input to the pipeline for an AnalysisPipeline or StreamingAnalysisPipeline object

This method is implemented on the base class as it is a shared functionality types of Analysis Pipelines which extend this class

Value

Updated AnalysisPipeline StreamingAnalysisPipeline object

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setLoggerDetails](#), [updateObject](#), [visualizePipeline](#)

Examples

```
library(analysisPipelines)
pipelineObj <- AnalysisPipeline()
pipelineObj %>>% setInput(input = iris) -> pipelineObj
```

setLoggerDetails	<i>Sets the logger configuration for the pipeline</i>
------------------	---

Description

Sets the logger configuration for the pipeline

Usage

```
setLoggerDetails(object, target = "console",
  targetFile = "pipelineExecution.out", layout = "layout.simple")

## S4 method for signature 'BaseAnalysisPipeline'
setLoggerDetails(object,
  target = "console", targetFile = "pipelineExecution.out",
  layout = "layout.simple")
```

Arguments

object	A Pipeline object
target	A string value. 'console' for appending to console, 'file' for appending to a file, or 'console&file' for both
targetFile	File name of the log file in case the target is 'file'
layout	Specify the layout according to 'futile.logger' package convention

Details

This function sets the logger configuration for the pipeline.

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [updateObject](#), [visualizePipeline](#)

Examples

```
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% setLoggerDetails(target = "file",
  targetFile = "pipeline.out") -> pipelineObj
```

<code>setPythonEnvir</code>	<i>Sets the python environment to be used</i>
-----------------------------	---

Description

Sets the python environment to be used

Usage

```
setPythonEnvir(type = "conda", pathOrEnvirName = "base")
```

Arguments

<code>type</code>	Type of python environment. Takes three possible vales - 'conda' for Anaconda environments, 'virtualenv' for Virtual environments, and 'python' to manually set the python path to use
<code>pathOrEnvirName</code>	Name of the environment for Anaconda and Virtual environments, or the Python path when type is 'python'

Details

Wrapper function over reticulate functions to set a python environment to be used

See Also

Other R helper utilities for Python: [getFeaturesForPyClassification](#), [getTargetForPyClassification](#)

Examples

```
## Not run:
setPythonEnvir()

## End(Not run)
```

```
sparkRSessionCreateIfNotPresent  
Connect to a Spark session
```

Description

Connect to a Spark session

Usage

```
sparkRSessionCreateIfNotPresent(...)
```

Arguments

```
... Arguments to sparkR.session
```

Details

Loads the SparkR package and initializes a Spark session from R

See Also

Other Spark utilities: [castKafkaStreamAsString](#), [convertKafkaValueFromJson](#)

Examples

```
## Not run:  
sparkHome <- "/Users/naren/software/spark-2.3.1-bin-hadoop2.7/"  
sparkMaster <- "local[1]"  
sparkPackages <- c("org.apache.spark:spark-sql-kafka-0-10_2.11:2.3.1")  
sparkRSessionCreateIfNotPresent(master = sparkMaster,  
  sparkPackages = sparkPackages)  
  
## End(Not run)
```

```
StreamingAnalysisPipeline-class  
Class for constructing Analysis Pipelines for streaming analyses
```

Description

Class for constructing Analysis Pipelines for streaming analyses

Details

Inherits the base class [BaseAnalysisPipeline](#) class which holds the metadata including the registry of available functions, the data on which the pipeline is to be applied, as well as the pipeline itself

This class currently only supports Apache Spark Structured Streaming, implemented through the SparkR interface

Slots

input The input Spark DataFrame on which analysis is to be performed

originalSchemaDf Empty Spark DataFrame representing the schema of the input

univarCatDistPlots *Univariate Categorical Distribution*

Description

Univariate Categorical Distribution

Usage

```
univarCatDistPlots(data, uniCol, priColor = "blue", optionalPlots = 0)
```

Arguments

data the dataset where the column on which the plot is to be generated is present

uniCol the name of column on which the plot needs to be generated

priColor the primary color for the plots

optionalPlots A Flag for optional plots

Details

A univariate distribution graph on the selected categorical columns from the dataframe

Value

A univariate categorical distribution plot

See Also

Other Package EDA Utilites functions: [CheckColumnType](#), [bivarPlots](#), [correlationMatPlot](#), [getDatatype](#), [ignoreCols](#), [multiVarOutlierPlot](#), [outlierPlot](#)

Examples

```
univarCatDistPlots(data = iris, uniCol = "Species")
```

updateObject	<i>Update the AnalysisPipeline or StreamingAnalysisPipeline object by adding an operation to the pipeline</i>
--------------	---

Description

Update the AnalysisPipeline or StreamingAnalysisPipeline object by adding an operation to the pipeline

Usage

```
updateObject(object, operation, heading = "", parameters, outAsIn = F,
             storeOutput = F)
```

```
## S4 method for signature 'BaseAnalysisPipeline'
updateObject(object, operation,
             heading = "", parameters, outAsIn = F, storeOutput = F)
```

Arguments

object	object that contains input, pipeline, registry and output
operation	function name to be updated in tibble
heading	heading of that section in report
parameters	parameters passed to that function
outAsIn	whether to use original input or output from previous function
storeOutput	whether the output of this operation is to be stored

Details

The specified operation along with the heading and parameters is updated in the pipeline slot of the AnalysisPipeline or StreamingAnalysisPipeline object, where the sequence of operations to be performed is stored

This method is implemented on the base class as it is a shared functionality types of Analysis Pipelines which extend this class

Value

Updated AnalysisPipeline StreamingAnalysisPipeline object

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize,BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [visualizePipeline](#)

visualizePipeline	<i>Visualizes the pipeline as a graph</i>
-------------------	---

Description

Visualizes the pipeline as a graph

Usage

```
visualizePipeline(object)

## S4 method for signature 'BaseAnalysisPipeline'
visualizePipeline(object)

## S4 method for signature 'MetaAnalysisPipeline'
visualizePipeline(object)
```

Arguments

object The AnalysisPipeline or StreamingAnalysisPipeline object

Details

Indicates dependencies amongst functions as well as functions for which output needs to be stored

Value

A graph object which can be printed (or) plotted to visualize the pipeline

See Also

Other Package core functions: [BaseAnalysisPipeline-class](#), [MetaAnalysisPipeline-class](#), [assessEngineSetUp](#), [checkSchemaMatch](#), [createPipelineInstance](#), [exportAsMetaPipeline](#), [generateOutput](#), [genericPipelineException](#), [getInput](#), [getLoggerDetails](#), [getOutputById](#), [getPipelinePrototype](#), [getPipeline](#), [getRegistry](#), [initDfBasedOnType](#), [initialize](#), [BaseAnalysisPipeline-method](#), [loadMetaPipeline](#), [loadPipeline](#), [loadPredefinedFunctionRegistry](#), [loadRegistry](#), [prepExecution](#), [registerFunction](#), [savePipeline](#), [saveRegistry](#), [setInput](#), [setLoggerDetails](#), [updateObject](#)

Examples

```
## Not run:
library(analysisPipelines)
pipelineObj <- AnalysisPipeline(input = iris)
pipelineObj %>>% univarCatDistPlots(uniCol = "Species",
  priColor = "blue", optionalPlots = 0, storeOutput = T) %>>%
visualizePipeline

## End(Not run)
```

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