Package ‘mlr3viz’

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Title  Visualizations for 'mlr3'

Version  0.7.0

Description  Visualization package of the 'mlr3' ecosystem. It features plots for mlr3 objects such as tasks, learners, predictions, benchmark results, tuning instances and filters via the 'autoplot()' generic of 'ggplot2'. The package draws plots with the 'viridis' color palette and applies the minimal theme. Visualizations include barplots, boxplots, histograms, ROC curves, and Precision-Recall curves.

License  LGPL-3


BugReports  https://github.com/mlr-org/mlr3viz/issues

Depends  R (>= 3.1.0)

Imports  checkmate, data.table, ggplot2 (>= 3.3.0), mlr3misc (>= 0.7.0), scales, utils, viridis

Suggests  bbotk (>= 0.7.3), cluster, GGally, ggdendro, ggfortify (>= 0.4.11), ggparty, glmnet, knitr, lgr, mlr3 (>= 0.6.0), mlr3cluster, mlr3filters, mlr3learners, mlr3tuning (>= 0.9.0), paradox, partykit, patchwork (>= 1.1.1), precrec, ranger, rpart, stats, testthat (>= 3.0.0), vdiffr (>= 1.0.2), xgboost

Config/testthat/edition  3

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Encoding  UTF-8

NeedsCompilation  no

RoxygenNote  7.2.3.9000

Collate  'BenchmarkResult.R' 'Filter.R' 'LearnerClassif.R'
          'LearnerClassifCVGlmnet.R' 'LearnerClassifGlmnet.R'
          'LearnerClassifRpart.R' 'LearnerClustHierarchical.R'
          'LearnerRegr.R' 'LearnerRegrCVGlmnet.R' 'LearnerRegrGlmnet.R'
          'LearnerRegrRpart.R' 'OptimInstanceSingleCrit.R' 'Prediction.R'
          'PredictionClassif.R' 'PredictionClust.R' 'PredictionRegr.R'
          'ResampleResult.R' 'Task.R' 'TaskClassif.R' 'TaskClust.R'
mlr3viz-package

mlr3viz: Visualizations for 'mlr3'

Description

Visualization package of the 'mlr3' ecosystem. It features plots for mlr3 objects such as tasks, learners, predictions, benchmark results, tuning instances and filters via the 'autoplot()' generic of 'ggplot2'. The package draws plots with the 'viridis' color palette and applies the minimal theme. Visualizations include barplots, boxplots, histograms, ROC curves, and Precision-Recall curves.
as_precrec

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See Also

Useful links:

- https://mlr3viz.mlr-org.com
- https://github.com/mlr-org/mlr3viz
- Report bugs at https://github.com/mlr-org/mlr3viz/issues

Description

Converts to a format which is understood by `precrec::evalmod()` of package `precrec`.

Usage

```r
as_precrec(object)
```

## S3 method for class 'PredictionClassif'
as_precrec(object)

## S3 method for class 'ResampleResult'
as_precrec(object)

## S3 method for class 'BenchmarkResult'
as_precrec(object)

Arguments

- `object` (any)
  - Object to convert.
**Value**

Object as created by `precrec::mmdata()`.

**References**


---

**autoplot.BenchmarkResult**

*Plots for Benchmark Results*

**Description**

Visualizations for `mlr3::BenchmarkResult`. The argument `type` controls what kind of plot is drawn. Possible choices are:

- "boxplot" (default): Boxplots of performance measures, one box per `mlr3::Learner` and one facet per `mlr3::Task`.
- "roc": ROC curve (1 - specificity on x, sensitivity on y). The `mlr3::BenchmarkResult` may only have a single `mlr3::Task` and a single `mlr3::Resampling`. Note that you can subset any `mlr3::BenchmarkResult` with its `$filter()` method (see examples). Requires package `precrec`.
- "prc": Precision recall curve. See "roc".

**Usage**

```r
## S3 method for class 'BenchmarkResult'
autoplot(
  object,
  type = "boxplot",
  measure = NULL,
  theme = theme_minimal(),
  ...
)
```

**Arguments**

- `object` *(mlr3::BenchmarkResult)*.
- `type` *(character(1)):
  Type of the plot. See description.
- `measure` *(mlr3::Measure)*
  Performance measure to use.
- `theme` *(ggplot2::theme())*
  The `ggplot2::theme_minimal()` is applied by default to all plots.
- `...` *(ignored).*
Value

`ggplot2::ggplot()`.

References


Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  tasks = tsk(c("pima", "sonar"))
  learner = lrm(c("classif.featureless", "classif.rpart"),
               predict_type = "prob")
  resampling = rmp(c("cv")
  object = benchmark(benchmark_grid(tasks, learner, resampling))

  head(fortify(object))
  autoplot(object)
  autoplot(object$clone(deep = TRUE)$filter(task_ids = "pima"), type = "roc")
}
```

Description

Visualizations for `mlr3filters::Filter`. The argument `type` controls what kind of plot is drawn. Possible choices are:

- "barplot" (default): Bar plot of filter scores.

Usage

```r
## S3 method for class 'Filter'
autoplot(object, type = "boxplot", n = Inf, theme = theme_minimal(), ...)
```

Arguments

- `object` (*mlr3filters::Filter*).
- `type` (character(1)): Type of the plot. See description.
- `n` (integer(1)): Only include the first n features with the highest importance. Defaults to all features.
theme: (ggplot2::theme())

The ggplot2::theme_minimal() is applied by default to all plots.

Value

ggplot2::ggplot().

Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)
  library(mlr3filters)

  task = tsk("mtcars")
  f = flt("correlation")
  f$calculate(task)

  head(fortify(f))
  autoplot(f)
  autoplot(f, n = 5)
}
```

**Description**

Visualizations for mlr3::LearnerClassif. The argument type controls what kind of plot is drawn. Possible choices are:

- "prediction" (default): Decision boundary of the learner and the true class labels.

**Usage**

```r
# S3 method for class 'LearnerClassif'
autoplot(
  object,
  type = "prediction",
  task,
  grid_points = 100L,
  expand_range = 0,
  theme = theme_minimal(),
  ...
)
```
Arguments

- **object** *(mlr3::LearnerClassif)*
- **type** *(character(1)): Type of the plot. See description.*
- **task** *(mlr3::Task)*
  - Train task.
- **grid_points** *(integer(1)) Number of grid points per feature dimension.*
- **expand_range** *(numeric(1)) Expand the range of the grid.*
- **theme** *(ggplot2::theme())*
  - The `ggplot2::theme_minimal()` is applied by default to all plots.
- **...** *(ignored).*

Value

`ggplot2::ggplot()`

Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  task = tsk("pima")$select(c("age", "pedigree"))
  learner = lrn("classif.rpart", predict_type = "prob"
  learner$train(task)

  autoplot(learner, type = "prediction", task)
}
```

Description

Visualizations for `mlr3learners::LearnerClassifGlmnet`. The argument `type` controls what kind of plot is drawn. Possible choices are:

- "prediction" (default): Decision boundary of the learner and the true class labels.
- "ggfortify": Visualizes the model using the package `ggfortify`.
Usage

```r
## S3 method for class 'LearnerClassifCVGlmnet'
autoplot(
  object,
  type = "prediction",
  task = NULL,
  grid_points = 100L,
  expand_range = 0,
  theme = theme_minimal(),
  ...
)

## S3 method for class 'LearnerClassifGlmnet'
autoplot(
  object,
  type = "prediction",
  task = NULL,
  grid_points = 100L,
  expand_range = 0,
  theme = theme_minimal(),
  ...
)

## S3 method for class 'LearnerRegrCVGlmnet'
autoplot(
  object,
  type = "prediction",
  task = NULL,
  grid_points = 100L,
  expand_range = 0,
  theme = theme_minimal(),
  ...
)

## S3 method for class 'LearnerRegrGlmnet'
autoplot(
  object,
  type = "prediction",
  task = NULL,
  grid_points = 100L,
  expand_range = 0,
  theme = theme_minimal(),
  ...
)
```
Arguments

object (mlr3learners::LearnerClassifGlmnet | mlr3learners::LearnerRegrGlmnet | mlr3learners::LearnerRegrCVGlmnet | mlr3learners::LearnerRegrCVGlmnet)

type (character(1))
Type of the plot. See description.

task (mlr3::Task)
Train task.

grid_points (integer(1))
Number of grid points per feature dimension.

expand_range (numeric(1))
Expand the range of the grid.

theme (ggplot2::theme())
The ggplot2::theme_minimal() is applied by default to all plots.

... (ignored).

Value

ggplot2::ggplot().

References


Examples

## Not run:
library(mlr3)
library(mlr3viz)
library(mlr3learners)

# classification
task = tsk("sonar")
learner = lrn("classif.glmnet")
learner$train(task)
autoplot(learner, type = "ggfortify")

# regression
task = tsk("mtcars")
learner = lrn("regr.glmnet")
learner$train(task)
autoplot(learner, type = "ggfortify")

## End(Not run)
Description

Visualizations for `mlr3::LearnerClassifRpart`. The argument `type` controls what kind of plot is drawn. Possible choices are:

- "prediction" (default): Decision boundary of the learner and the true class labels.
- "ggparty": Visualizes the tree using the package `ggparty`.

Usage

```r
## S3 method for class 'LearnerClassifRpart'
autoplot(
  object,
  type = "prediction",
  task = NULL,
  grid_points = 100L,
  expand_range = 0,
  theme = theme_minimal(),
  ...
)

## S3 method for class 'LearnerRegrRpart'
autoplot(
  object,
  type = "prediction",
  task = NULL,
  grid_points = 100L,
  expand_range = 0,
  theme = theme_minimal(),
  ...
)
```

Arguments

- `object` (`mlr3::LearnerClassifRpart | mlr3::LearnerRegrRpart`): 
- `type` (character(1)):
  Type of the plot. See description.
- `task` (`mlr3::Task`):
  Train task.
- `grid_points` (integer(1)):
  Number of grid points per feature dimension.
autoplot.LearnerClustHierarchical

Description

Visualizations for hierarchical clusters. The argument type controls what kind of plot is drawn. Possible choices are:

- "dend" (default): Dendrograms using ggdendro package.
- "scree": Scree plot that shows the number of possible clusters on the x-axis and the height on the y-axis.

Usage

```r
## S3 method for class 'LearnerClustHierarchical'
autoplot(
  object,
  type = "dend",
```

expand_range  (numeric(1))
   Expand the range of the grid.

theme  (ggplot2::theme())
   The ggplot2::theme_minimal() is applied by default to all plots.

Value

ggplot2::ggplot().

Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  # classification
  task = tsk("iris")
  learner = lrn("classif.rpart", keep_model = TRUE)
  learner$train(task)
  autoplot(learner, type = "ggparty")

  # regression
  task = tsk("mtcars")
  learner = lrn("regr.rpart", keep_model = TRUE)
  learner$train(task)
  autoplot(learner, type = "ggparty")
}
```
autoplot.LearnerClustHierarchical

```r
  task = NULL,
  theme = theme_minimal(),
  theme_dendro = TRUE,
  ...
)
```

**Arguments**

- `object` *(mlr3cluster::LearnerClustAgnes | mlr3cluster::LearnerClustDiana | mlr3cluster::LearnerClustHclust)*
- `type` *(character(1))*
  - Type of the plot. See description.
- `task` *(mlr3::Task)*
  - Optionally, pass the task to add labels of observations to a `hclust` dendrogram. Labels are set via the row names of the task.
- `theme` *(ggplot2::theme())*
  - The `ggplot2::theme_minimal()` is applied by default to all plots.
- `theme_dendro` *(logical(1))*
  - If TRUE (default), the special dendrogram theme from `ggdendro` package is used in plot "dend". Set to FALSE to use the theme passed in `theme`.
  - (ignored).

**Value**

`ggplot2::ggplot()`

**Examples**

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3cluster)
  library(mlr3viz)

  task = tsk("usarrests")

  # agnes clustering
  learner = lrn("clust.agnes")
  learner$train(task)
  autoplot(learner)

  # diana clustering
  learner = lrn("clust.diana")
  learner$train(task)
  autoplot(learner)

  # hclust clustering
  learner = lrn("clust.hclust")
  learner$train(task)
  autoplot(learner, type = "scree")
}
```
Description

Visualizations for mlr3::LearnerRegr. The argument type controls what kind of plot is drawn. Possible choices are:

- "prediction" (default): Decision boundary of the learner and the true class labels.

Usage

```r
## S3 method for class 'LearnerRegr'
autoplot(
  object,  
  type = "prediction",  
  task,  
  grid_points = 100L,  
  expand_range = 0,  
  theme = theme_minimal(),  
  ...  
)
```

Arguments

- `object` (mlr3::LearnerRegr).
- `type` (character(1)): Type of the plot. See description.
- `task` (mlr3::Task) Train task.
- `grid_points` (integer(1)) Number of grid points per feature dimension.
- `expand_range` (numeric(1)) Expand the range of the grid.
- `theme` (ggplot2::theme()) The ggplot2::theme_minimal() is applied by default to all plots.
- `...` (ignored).

Value

`ggplot2::ggplot()`.
Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  task = tsk("mtcars")$select(c("am", "carb"))
  learner = lrn("regr.rpart")
  learner$train(task)

  autoplot(learner, type = "prediction", task)
}
```

### autoplot.OptimInstanceSingleCrit

*Plots for Optimization Instances*

**Description**

Visualizations for `bbotk::OptimInstanceSingleCrit`. The argument `type` controls what kind of plot is drawn. Possible choices are:

- "marginal" (default): Scatter plots of x versus y. The color of the points shows the batch number.
- "performance": Scatter plots of batch number versus y
- "parameter": Scatter plots of batch number versus input. The color of the points shows the y values.
- "parallel": Parallel coordinates plot. x values are rescaled by \( (x - \text{mean}(x)) / \text{sd}(x) \).
- "points": Scatter plot of two x dimensions versus. The color of the points shows the y values.
- "surface": Surface plot of two x dimensions versus y values. The y values are interpolated with the supplied `mlr3::Learner`.
- "pairs": Plots all x and y values against each other.
- "incumbent": Plots the incumbent versus the number of configurations.

**Usage**

```r
## S3 method for class 'OptimInstanceSingleCrit'
autoplot(
  object,
  type = "marginal",
  cols_x = NULL,
  trafo = FALSE,
  learner = mlr3::lrn("regr.ranger"),
  grid_resolution = 100,
)```
autoplot.OptimInstanceSingleCrit

batch = NULL,
theme = theme_minimal(),
...
)

Arguments

object (bbotk::OptimInstanceSingleCrit).
type (character(1)):
Type of the plot. See description.
cols_x (character())
Column names of x values. By default, all untransformed x values from the
search space are plotted. Transformed hyperparameters are prefixed with x_domain_.
trafo (logical(1))
If FALSE (default), the untransformed x values are plotted. If TRUE, the trans-
formed x values are plotted.
learner (mlr3::Learner)
Regression learner used to interpolate the data of the surface plot.
grid_resolution (numeric())
Resolution of the surface plot.
batch (integer())
The batch number(s) to limit the plot to. The default is all batches.
theme (ggplot2::theme())
The ggplot2::theme_minimal() is applied by default to all plots.
...(ignored).

Value

ggplot2::ggplot().

Examples

if (requireNamespace("mlr3") && requireNamespace("bbotk") && requireNamespace("patchwork")) {
library(bbotk)
library(paradox)

fun = function(xs) {
  c(y = -(xs[[1]] - 2)^2 - (xs[[2]] + 3)^2 + 10)
}
domain = ps(
  x1 = p_dbl(-10, 10),
  x2 = p_dbl(-5, 5)
)
codomain = ps(
  y = p_dbl(tags = "maximize")
)
obfun = ObjectiveRFun$new(
  fun = fun,
  ...
fun = fun,
domain = domain,
codomain = codomain
)

instance = OptimInstanceSingleCrit$new(objective = obfun, terminator = trm("evals", n_evals = 20))

optimizer = opt("random_search", batch_size = 2)
optimizer$optimize(instance)

# plot y versus batch number
print(autoplot(instance, type = "performance"))

# plot x1 values versus performance
print(autoplot(instance, type = "marginal", cols_x = "x1"))

# plot parallel coordinates plot
print(autoplot(instance, type = "parallel"))

# plot pairs
print(autoplot(instance, type = "pairs"))

# plot incumbent
print(autoplot(instance, type = "incumbent"))

autoplot.PredictionClassif

Plots for Classification Predictions

Description

Visualizations for mlr3::PredictionClassif. The argument type controls what kind of plot is drawn. Possible choices are:

- "stacked" (default): Stacked barplot of true and estimated class labels.
- "roc": ROC curve (1 - specificity on x, sensitivity on y). Requires package precrec.
- "prc": Precision recall curve. Requires package precrec.
- "threshold": Systematically varies the threshold of the mlr3::PredictionClassif object and plots the resulting performance as returned by measure.

Usage

## S3 method for class 'PredictionClassif'
autoplot(
  object,
  type = "stacked",
  measure = NULL,
)
autoplot.PredictionClassif

```r
theme = theme_minimal(),
...
)
```

**Arguments**

- `object` *(mlr3::PredictionClassif).*
- `type` *(character(1)): Type of the plot. See description.*
- `measure` *(mlr3::Measure) Performance measure to use.*
- `theme` *(ggplot2::theme()) The ggplot2::theme_minimal() is applied by default to all plots.*
- *(ignored).*

**Value**

`ggplot2::ggplot()`

**References**


**Examples**

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  task = tsk("spam")
  learner = lrn("classif.rpart", predict_type = "prob")
  object = learner$train(task)$predict(task)

  head(fortify(object))
  autoplot(object)
  autoplot(object, type = "roc")
  autoplot(object, type = "prc")
}
```
Description

Visualizations for mlr3cluster::PredictionClust. The argument type controls what kind of plot is drawn. Possible choices are:

- "scatter" (default): scatterplot with correlation values and colored cluster assignments.
- "sil": Silhouette plot with mean silhouette value as the reference line. Requires package ggfortify.
- "pca": Perform PCA on data and color code cluster assignments. Inspired by and uses ggfortify::autoplot.kmeans.

Usage

```r
## S3 method for class 'PredictionClust'
autoplot(
  object,
  task,
  row_ids = NULL,
  type = "scatter",
  theme = theme_minimal(),
  ...
)
```

Arguments

- `object` (mlr3cluster::PredictionClust).
- `task` (mlr3cluster::TaskClust).
- `row_ids` (integer()) Row ids to subset task data to ensure that only the data used to make predictions are shown in plots.
- `type` (character(1)): Type of the plot. See description.
- `theme` (ggplot2::theme()): The ggplot2::theme_minimal() is applied by default to all plots.

Value

ggplot2::ggplot().

References

Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3cluster)
  library(mlr3viz)

  task = tsk("usarrests")
  learner = lrn("clust.kmeans", centers = 3)
  object = learner$train(task)$predict(task)

  head(fortify(object))
  autoplot(object, task)
}
```

**Description**

Visualizations for `mlr3::PredictionRegr`. The argument `type` controls what kind of plot is drawn. Possible choices are:

- "xy" (default): Scatterplot of "true" response vs. "predicted" response. By default a linear model is fitted via `geom_smooth(method = "lm")` to visualize the trend between x and y (by default colored blue). In addition `geom_abline()` with `slope = 1` is added to the plot. Note that `geom_smooth()` and `geom_abline()` may overlap, depending on the given data.
- "histogram": Histogram of residuals: \( r = y - \hat{y} \).
- "residual": Plot of the residuals, with the response \( \hat{y} \) on the "x" and the residuals on the "y" axis. By default a linear model is fitted via `geom_smooth(method = "lm")` to visualize the trend between x and y (by default colored blue).
- "confidence": Scatterplot of "true" response vs. "predicted" response with confidence intervals. Error bars calculated as `object$response + quantile * object$se` and so only possible with `predict_type = "se"`. `geom_abline()` with `slope = 1` is added to the plot.

**Usage**

```r
## S3 method for class 'PredictionRegr'
autoplot(object, 
  type = "xy", 
  binwidth = NULL, 
  theme = theme_minimal(), 
  quantile = 1.96, 
  ...)
```
Arguments

**object**  
(mlr3::PredictionRegr).

**type**  
(character(1)):  
Type of the plot. See description.

**binwidth**  
(integer(1))  
Width of the bins for the histogram.

**theme**  
(ggplot2::theme())  
The ggplot2::theme_minimal() is applied by default to all plots.

**quantile**  
(numeric(1))  
Quantile multiplier for standard errors for type="confidence". Default 1.96.

...  
(ignored).

Value

ggplot2::ggplot().

Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  task = tsk("boston_housing")
  learner = lrn("regr.rpart")
  object = learner$train(task)$predict(task)

  head(fortify(object))
  autoplot(object)
  autoplot(object, type = "histogram", binwidth = 1)
  autoplot(object, type = "residual")

  if (requireNamespace("mlr3learners")) {
    library(mlr3learners)
    learner = lrn("regr.ranger", predict_type = "se")
    object = learner$train(task)$predict(task)
    autoplot(object, type = "confidence")
  }
}
```

---

**autoplot.ResampleResult**

Plots for Resample Results
Description

Visualizations for `mlr3::ResampleResult`. The argument `type` controls what kind of plot is drawn. Possible choices are:

- "boxplot" (default): Boxplot of performance measures.
- "histogram": Histogram of performance measures.
- "roc": ROC curve (1 - specificity on x, sensitivity on y). The predictions of the individual `mlr3::Resamplings` are merged prior to calculating the ROC curve (micro averaged). Requires package `precrec`.
- "prc": Precision recall curve. See "roc".
- "prediction": Plots the learner prediction for a grid of points. Needs models to be stored. Set `store_models = TRUE` for `[mlr3::resample]`. For classification, we support tasks with exactly two features and learners with `predict_type` set to "response" or "prob". For regression, we support tasks with one or two features. For tasks with one feature we can print confidence bounds if the predict type of the learner was set to "se". For tasks with two features the predict type will be ignored.

Usage

```r
## S3 method for class 'ResampleResult'
autoplot(
  object,
  type = "boxplot",
  measure = NULL,
  predict_sets = "test",
  binwidth = NULL,
  theme = theme_minimal(),
  ...
)
```

Arguments

- `object` (mlr3::ResampleResult).
- `type` (character(1)): Type of the plot. See description.
- `measure` (mlr3::Measure): Performance measure to use.
- `predict_sets` (character()): For `type` set to "prediction". Which points should be shown in the plot? Can be a subset of ("train", "test") or empty.
- `binwidth` (integer(1)): Width of the bins for the histogram.
- `theme` (ggplot2::theme()): The `ggplot2::theme_minimal()` is applied by default to all plots.
- `...` (ignored).
Value

`ggplot2::ggplot()`.

References


Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  task = tsk("sonar")
  learner = lrn("classif.rpart", predict_type = "prob")
  resampling = rsmp("cv", folds = 3)
  object = resample(task, learner, resampling)

  head(fortify(object))

  # Default: boxplot
  autoplot(object)

  # Histogram
  autoplot(object, type = "histogram", bins = 30)

  # ROC curve, averaged over resampling folds:
  autoplot(object, type = "roc")

  # ROC curve of joint prediction object:
  autoplot(object$prediction(), type = "roc")

  # Precision Recall Curve
  autoplot(object, type = "prc")

  # Prediction Plot
  task = tsk("iris")$select(c("Sepal.Length", "Sepal.Width"))
  resampling = rsmp("cv", folds = 3)
  object = resample(task, learner, resampling, store_models = TRUE)
  autoplot(object, type = "prediction")
}
```
Description

Visualizations for `mlr3::TaskClassif`. The argument type controls what kind of plot is drawn. Possible choices are:

- "target" (default): Bar plot of the target variable (default).
- "duo": Passes data to `GGally::ggduo()`. `columnsX` is the target and `columnsY` are the features.
- "pairs": Passes data to `GGally::ggpairs()`. Color is set to target column.

Usage

```r
## S3 method for class 'TaskClassif'
autoplot(object, type = "target", theme = theme_minimal(), ...)
```

Arguments

- **object** (mlr3::TaskClassif).
- **type** (character(1)): Type of the plot. See description.
- **theme** (ggplot2::theme()): The `ggplot2::theme_minimal()` is applied by default to all plots.
- **...** (ignored).

Value

`ggplot2::ggplot()`.

Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  task = tsk("iris")

  head(fortify(task))
  autoplot(task)
  autoplot(task$clone()$select(c("Sepal.Length", "Sepal.Width")),
    type = "pairs")
  autoplot(task, type = "duo")
}
```
autoplot.TaskClust  Plots for Clustering Tasks

Description
Visualizations for mlr3cluster::TaskClust. The argument type controls what kind of plot is drawn. Possible choices are:

- "pairs" (default): Passes data GGally::ggpairs().

Usage
## S3 method for class 'TaskClust'
autoplot(object, type = "pairs", theme = theme_minimal(), ...)

Arguments
object  (mlr3cluster::TaskClust).
type  (character(1)): Type of the plot. See description.
theme  (ggplot2::theme())
The ggplot2::theme_minimal() is applied by default to all plots.
...
(ignored).

Value

ggplot2::ggplot().

Examples

if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3cluster)
  library(mlr3viz)

  task = mlr_tasks$get("usarrests")

  head(fortify(task))
  autoplot(task)
  }

autoplot.TaskRegr  Plots for Regression Tasks

Description

Visualizations for mlr3::TaskRegr. The argument type controls what kind of plot is drawn. Possible choices are:

- "target" (default): Box plot of the target variable.
- "pairs": Passes data to GGally::ggpairs(). Color is set to target column.

Usage

```r
## S3 method for class 'TaskRegr'
autoplot(object, type = "target", theme = theme_minimal(), ...)
```

Arguments

- `object` (mlr3::TaskRegr).
- `type` (character(1)): Type of the plot. See description.
- `theme` (ggplot2::theme()): The ggplot2::theme_minimal() is applied by default to all plots.
- `...` (ignored).

Value

ggplot2::ggplot().

Examples

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  task = tsk("mtcars")
  task$select(c("am", "carb"))

  head(fortify(task))
  autoplot(task)
  autoplot(task, type = "pairs")
}
```
Description

Visualizations for `mlr3tuning::TuningInstanceSingleCrit`. The argument `type` controls what kind of plot is drawn. Possible choices are:

- "marginal" (default): Scatter plots of x versus y. The color of the points shows the batch number.
- "performance": Scatter plots of batch number versus y
- "parameter": Scatter plots of batch number versus input. The color of the points shows the y values.
- "parallel": Parallel coordinates plot. Hyperparameters are rescaled by \( (x - \text{mean}(x)) / \text{sd}(x) \).
- "points": Scatter plot of two x dimensions versus. The color of the points shows the y values.
- "surface": Surface plot of two x dimensions versus y values. The y values are interpolated with the supplied `mlr3::Learner`.
- "pairs": Plots all x and y values against each other.
- "incumbent": Plots the incumbent versus the number of configurations.

Usage

```r
## S3 method for class 'TuningInstanceSingleCrit'
autoplot(
  object,
  type = "marginal",
  cols_x = NULL,
  trafo = FALSE,
  learner = mlr3::lrn("regr.ranger"),
  grid_resolution = 100,
  theme = theme_minimal(),
  ...
)
```

Arguments

- `object` (mlr3tuning::TuningInstanceSingleCrit): Object of class `TuningInstanceSingleCrit`.
- `type` (character(1)): Type of the plot. See description.
- `cols_x` (character(vector)): Column names of hyperparameters. By default, all untransformed hyperparameters are plotted. Transformed hyperparameters are prefixed with `x_domain_`. 
trafo (logical(1))
If FALSE (default), the untransformed hyperparameters are plotted. If TRUE, the
transformed hyperparameters are plotted.

learner (mlr3::Learner)
Regression learner used to interpolate the data of the surface plot.

grid_resolution (numeric())
Resolution of the surface plot.

theme (ggplot2::theme())
The ggplot2::theme_minimal() is applied by default to all plots.

Value

ggplot2::ggplot().

Examples

if (requireNamespace("mlr3tuning") && requireNamespace("patchwork")) {
  library(mlr3tuning)

  learner = lrn("classif.rpart")
  learner$param_set$values$cp = to_tune(0.001, 0.1)
  learner$param_set$values$minsplit = to_tune(1, 10)

  instance = TuningInstanceSingleCrit$new(
    task = tsk("iris"),
    learner = learner,
    resampling = rsmp("holdout"),
    measure = msr("classif.ce"),
    terminator = trm("evals", n_evals = 10))

  tuner = tnr("random_search")

  tuner$optimize(instance)

  # plot performance versus batch number
  autoplot(instance, type = "performance")

  # plot cp values versus performance
  autoplot(instance, type = "marginal", cols_x = "cp")

  # plot transformed parameter values versus batch number
  autoplot(instance, type = "parameter", trafo = TRUE)

  # plot parallel coordinates plot
  autoplot(instance, type = "parallel")

  # plot pairs
  autoplot(instance, type = "pairs")
}

**plot_learner_prediction**

*Plots for Learner Predictions*

---

**Description**

Visualizations for the `mlr3::Prediction` of a single `mlr3::Learner` on a single `mlr3::Task`.

- For classification we support tasks with exactly two features and learners with `predict_type` set to "response" or "prob".
- For regression we support tasks with one or two features. For tasks with one feature we print confidence bounds if the predict type of the learner was set to "se". For tasks with two features the predict type will be ignored.

Note that this function is a wrapper around `autoplot.ResampleResult()` for a temporary `mlr3::ResampleResult` using `mlr3::mlr_resamplings_holdout` with ratio 1 (all observations in the training set).

**Usage**

```r
plot_learner_prediction(learner, task, grid_points = 100L, expand_range = 0)
```

**Arguments**

- **learner** *(mlr3::Learner)*.
- **task** *(mlr3::Task)*.
- **grid_points** *(integer(1))*
  Resolution of the grid. For factors, ordered and logicals this value is ignored.
- **expand_range** *(numeric(1))*
  Expand the prediction range for numerical features.

**Value**

`ggplot2::ggplot()`.

**Examples**

```r
if (requireNamespace("mlr3")) {
  library(mlr3)
  library(mlr3viz)

  task = mlr3::tsk("pima")$select(c("age", "glucose"))
  learner = lrn("classif.rpart", predict_type = "prob")
  p = plot_learner_prediction(learner, task)
  print(p)
}
```
**predict_grid**

Generates a data.table of evenly distributed points.

**Description**

For each point we have the predicted class / regression value in column `response`. If the learner predicts probabilities, a column `.prob.response` is added that contains the probability of the predicted class.

**Usage**

```r
predict_grid(learners, task, grid_points, expand_range)
```

**Arguments**

- `learners`: list of trained learners, each learner belongs to one resampling iteration
- `task`: the task all learners are trained on
- `grid_points`: (int): see sequenize
- `expand_range`: see sequenize
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