Applied Machine Learning

Feature Engineering: mlr3pipelines - Linear Pipelines



Learning goals

Linear pipelines in action

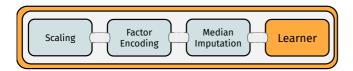


mlr3pipelines: Linear Pipelines

Linear Preprocessing Pipeline

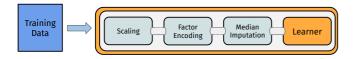
```
graph = po("scale") %>>%
  po("encode") %>>%
  po("imputemedian") %>>%
  lrn("classif.rpart")
```





Linear Preprocessing Pipeline

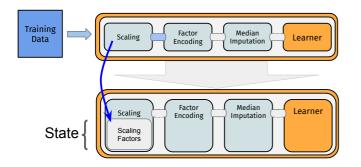
```
glrn = as_learner(graph) # or: GraphLearnerfnew(graph)
glrn$train(task)
```





Linear Preprocessing Pipeline

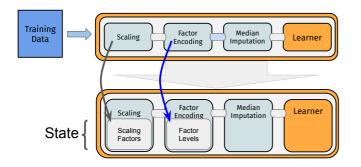
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Linear Preprocessing Pipeline

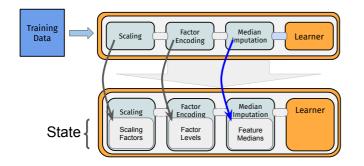
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Linear Preprocessing Pipeline

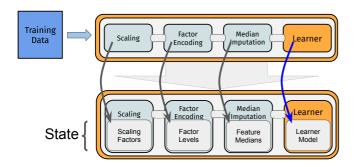
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Linear Preprocessing Pipeline

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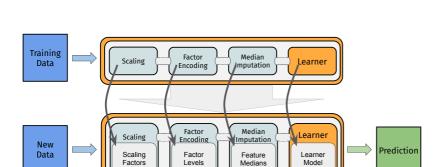




Linear Preprocessing Pipeline

- train()ing: Data propagates and creates \$states
- predict()tion: Data propagates, uses \$states

glrn\$predict(task)





Linear Preprocessing Pipeline scale %>>% encode %>>% impute %>>% rpart

• Setting / retrieving parameters: \$param_set

```
{\tt graph\$pipeops\$scale\$param\_set\$values\$center} = {\tt FALSE}
```

• Retrieving state: \$state of individual PipeOps (after \$train())

```
graph$pipeops$scale$state$scale
#> Petal.Length Petal.Width Sepal.Length Sepal.Width
#> 4.2 1.4 5.9 3.1
```

Retrieving intermediate results: \$.result (set debug option before)

graph\$pipeops\$scale\$.result[[1]]\$head(3) Species Petal.Length Petal.Width Sepal.Length Sepal.Width #> #> <fctr> <n11m> <n11m> <n11m> <n11m> 0.34 0.14 0.86 1.13 #> 1: setosa 0.34 0.14 0.83 0.97 #> 2: setosa 0.31 0.14 0.79 #> 3: setosa 1.03

