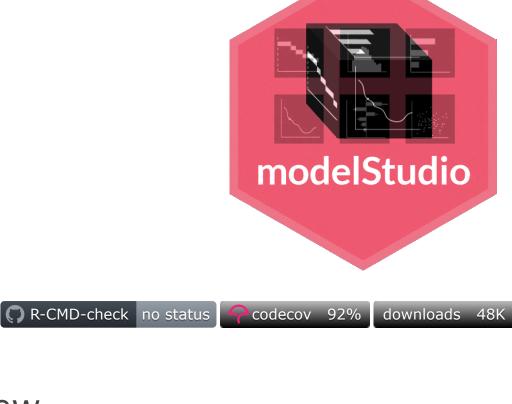
Articles ▼

JOSS 10.21105/joss.01798

Interactive Studio for Explanatory Model **Analysis**



The modelStudio package automates the explanatory analysis of machine learning predictive models. It generates advanced interactive model explanations in the form of a serverless HTML site with

only one line of code. This tool is model-agnostic, therefore compatible with most of the black-box predictive models and frameworks (e.g. mlr/mlr3, xgboost, caret, h2o, parsnip, tidymodels,

scikit-learn, lightgbm, keras/tensorflow).

CRAN 3.1.2

Overview

The main modelStudio() function computes various (instance and model-level) explanations and produces a **customisable dashboard**, which consists of multiple panels for plots with their short descriptions. It is possible to easily **save** the dashboard and **share** it with others. Tools for Explanatory Model Analysis unite with tools for Exploratory Data Analysis to give a broad overview of the model behavior.

explain COVID-19 **R & Python examples** More resources **Interactive EMA** id: United Arab Emirates | y: 6.825 perceptions_of_corruption > **Interactive Studio for Random Forest Model** D X **Ceteris Paribus Partial Dependence** average prediction

mse: 0.323 | rmse: 0.569 | r2: 0.737 | mad: 0.395 Site built with modelStudio v3.0.0 on 2021-07-09 10:33:48 The modelStudio package is a part of the **DrWhy.Al** universe. Installation

0.00

perceptions_of_corruption = 0.182

Site built with modelStudio v3.0.0 on 2021-07-08 19:40:12

0.45

model <- ranger(score ~., data = happiness_train)</pre>

Install the development version from GitHub:

devtools::install_github("ModelOriented/modelStudio")

Install from CRAN:

Simple demo

library("DALEX")

library("ranger")

fit a model

library("modelStudio")

install.packages("modelStudio")

perceptions_of_corruption

```
# create an explainer for the model
  explainer <- explain(model,</pre>
                             data = happiness_test,
                             y = happiness_test$score,
                             label = "Random Forest")
  # make a studio for the model
  modelStudio(explainer)
Save the output in the form of a HTML file - Demo Dashboard.
                                                                               id: Central African Republic | y: 3.083 V
                                                            gdp_per_capita
  Interactive Studio for Random Forest Model
                                                 D X
                Feature Importance
        gdp_per_capita
                                                                          Break Down [Local]
                                                                         Shapley Values [Local]
```

R & Python examples more

packages for the explainer objects

mse: 0.317 | rmse: 0.563 | r2: 0.743 | mad: 0.395

DALEXtra::explain_*().

mlr dashboard

library(mlr)

library(DALEXtra)

split the data

fit a model

train <- data[index,]</pre>

test <- data[-index,]</pre>

pick observations

explainer <- explain_mlr(model,</pre>

new_observation <- test[1:2,]</pre>

make a studio for the model

rownames(new_observation) <- c("id1", "id2")</pre>

library(modelStudio)

▼ code

install.packages("DALEX")

load packages and data

install.packages("DALEXtra")

healthy_life_expectancy **Ceteris Paribus [Local]** Partial Dependence [Global] freedom_life_choices **Accumulated Dependence [Global] Residuals vs Feature [Global]** social_support Feature Distribution [EDA] Target vs Feature [EDA] perceptions_of_corruption **Average Target vs Feature [EDA]** generosity 0.85 Root mean square error (RMSE) +

The modelStudio() function uses DALEX explainers created with DALEX::explain() or

Make a studio for the regression ranger model on the apartments data.

data <- DALEX::apartments</pre>

index <- sample(1:nrow(data), 0.7*nrow(data))</pre>

```
task <- makeRegrTask(id = "apartments", data = train, target = "m2.price")</pre>
learner <- makeLearner("regr.ranger", predict.type = "response")</pre>
model <- train(learner, task)</pre>
# create an explainer for the model
```

y = test\$m2.price,

label = "mlr")

```
modelStudio(explainer, new_observation)
xgboost dashboard
Make a studio for the classification xgboost model on the titanic data.
▶ code
The modelStudio() function uses dalex explainers created with dalex.Explainer().
 :: package for the Explainer object
 pip install dalex -U
Use pickle Python module and reticulate R package to easily make a studio for a model.
 # package for pickle load
 install.packages("reticulate")
scikit-learn dashboard
Make a studio for the regression Pipeline SVR model on the fifa data.
▶ code
lightgbm dashboard
Make a studio for the classification Pipeline LGBMClassifier model on the titanic data.
▶ code
```

Citations If you use modelStudio, please cite our JOSS article:

@article{baniecki2019modelstudio,

 $= \{2019\},$

 $= \{2023\},$

pages = $\{1--37\}$,

author = {Hubert Baniecki and Przemyslaw Biecek},

= {https://doi.org/10.21105/joss.01798}

journal = {Journal of Open Source Software},

Save & share

RStudio Viewer

year

url

year

url

Models

volume = $\{4\}$,

number = $\{43\}$,

pages = $\{1798\}$,

r2d3::save_d3_html().

☐ Save As Publish ▼

```
}
For a description and evaluation of the Interactive EMA process, refer to our DAMI article:
  @article{baniecki2023grammar,
```

= {The grammar of interactive explanatory model analysis},

author = {Hubert Baniecki and Dariusz Parzych and Przemyslaw Biecek},

Save modelStudio as a HTML file using buttons on the top of the RStudio Viewer or with

Interactive Studio for Random Forest Model

= {{modelStudio: Interactive Studio with Explanations for ML Predictive M

```
More resources
  • Introduction to the plots: Explanatory Model Analysis: Explore, Explain, and Examine Predictive
```

= {https://doi.org/10.1007/s10618-023-00924-w}

journal = {Data Mining and Knowledge Discovery},

• Changelog: NEWS

• Conference poster: ML in PL 2019

• Vignettes: perks and features, R & Python examples, modelStudio in R Markdown HTML

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0328/17.

Developed by Hubert Baniecki, Przemyslaw Biecek.