

PyCaret is an open source, low-code machine learning library in Python that allows you to go preparing your data to deploying your model within minutes in your choice of notebook environment

Installing PyCaret

install pycaret

pip install pycaret # install full version of pycaret

pip install pycaret[full]

install pycaret time series module

pip install pycaret-ts-alpha

PvCaret on GPU

uninstall lightqbm CPU pip uninsstall lightgbm -y

install lightgbm GPU pip install lightgbm --install-option=--gpu --install-option="--opencl-include-dir=/usr/

/local/include/" --install-option="--opencllibrary=usr/local/cuda/lib64/libOpenCL.so"

Run PvCaret on a Docker Container

FROM python:3.7-slim WORKDIR /app

ADD . /app RUN apt-get update && apt-get install libgomp1 RUN pip install --trusted-host pypi.python.org

-r requirements.txt

CMD pytest # replace it with your entry point

PvCaret Tutorials

Binary classification (Beginner) Binary classification (Intermediate)

Multiclass classification (Beginner)

Regression

Classification

Regression (Beginner) Regression (Intermediate)

Clustering

Clustering (Beginner)

Anomaly detection

Anomaly detection (Beginner)

Natural Language Processing

NLP (Beginner) NLP (Intermediate)

Association Rule Mining

Association Rule Mining (Beginner)

Time series and forecasting (Beginner)

Loading data from PyCaret's repository

loading data from pycaret

from pycaret.datasets import get data data = get data('dataset name')

Loading data using Pandas

importing pandas

import pandas as pd df = pd.read csv(r'dir/file name.csv')

Regression and Classification

set up environment

from pycaret.regression import * from pycaret.classification import * clf1 = setup(data = df, target='column')

create and evaluate model

compare models()

model = create model(*)

model tuned = tune model(model)

ens model = ensemble model(model, method=***)

blender = blend models(top3) stacker = stack models(top3)

plot model(model, plot=**) evaluate model(model)

interpret model(model) (1) calibrate_model()

(1) optimize threshold()

make predictions

df1 = predict model(model=model, data=df)

model deployment

final model = finalize model(model)

save model(model, 'saved model') model loaded = load model('saved model')

deploy model(model=model. model name=model final,

platform = 'aws', authentication = {'bucket :

'S3-bucket-name'})

utils ()lluq

models()

get metrics() add metric()

remove metric()

get logs() get_config()

set config()

save config()

load_config()

get leaderboard()

*model:

(regression) (classification)

'rbfsvm'

'lasso' 'knn' 'ridge 'nb' 'en' 'dt' 'lar' 'svm' (regression) (classification)

'amo' 'gpc' 'br' 'mlp 'ard' 'ridge' 'rf' 'par' 'ransac' 'ada' 'tr' 'ada' 'gbc' 'huber

'kr' 'lda' 'svm' 'et' 'knn' 'xgboost'

'dt' 'lightgbm 'catboost' 'et'

'ghr' 'mlp' 'xgboost'

'lightgbm' 'catboost'

**plot=

'residuals interactive' 'residuals' 'error' 'cooks' 'rfe' 'learning' 'vc' 'manifold' 'feature' 'feature all' 'parameter' 'tree'

*** method=

'bagging' 'boosting'

(1) classification only

Time Series Analysis

set up environment from pycaret.time_series import *

exp = setup(data = df, fh = 12)

create and evaluate model

model tuned = tune model(model)

blender = blend_models(top3)

plot model(model, plot=**)

final model = finalize model(model)

make predictions

pred holdout = predict model(arima)

arima), fh=24)

model deployment

deploy model(model=model,

pull() models() get metrics() add_metric() remove metric() get_logs() get config() set config()

**plot= * model: 'naïve' 'ts' 'cv' 'grand means' 'snaive' 'acf' 'polytrend' 'acf' 'arima' 'pacf' 'exp smooth' 'decomp stl' 'ets' 'diagnostics' 'theta' 'forecast' 'tbats' 'insample' 'bats' 'residuals' 'train test split' 'prophet'

'decomp_classical

Unsupervised Learning

Clustering

'lr_cds_dt' 'en cds' 'ridge_cds_dt' 'lasso cds dt' 'lar cds dt' 'llar cds dt' 'br cds dt' 'huber cds dt' 'par cds dt'

'omp cds dt' 'knn cds dt' 'dt cds dt'

'rf cds dt' 'et cds dt'

'gbr cds dt' 'ada_cds_dt'

set up environment

clf1 = setup(data = df)

from pycaret.clustering import *

create and evaluate model

model = create model(*)

model df = assign model(*)

plot model(model, plot=**)

model tuned = tune model(model=model,

save model(model, 'saved model')

deploy model(model=model,

model_name=model final,

df1 = predict model(model=model, data = df)

model loaded = load model('saved model')

platform = 'aws', authentication = {'bucket :

supervised target = 'column name')

evaluate_model(model)

make predictions

model deployment

'S3-bucket-name'})

utils

pull()

models()

get metrics()

add_metric()

get_logs()

get config()

remove metric()

'lightgbm cds dt'

compare models()

model = create model(*)

pred_unseen = predict_model(finalize_model(

final model = finalize model(model) save model(model, 'saved model')

model loaded = load model('saved model')

model name=model final,

platform = 'aws', authentication = {'bucket :

'S3-bucket-name'}) # utils

save config()

model = create_model(*) model df = assign model(*)

evaluate model(model) model tuned = tune model(model=model,

make predictions

df1 = predict model(model=model, data=df)

save model(model, 'saved model') model loaded = load model('saved model')

platform = 'aws', authentication = {'bucket:

utils

()lluq models()

get_config() set config()

get clusters()

* model:

'cluster' 'histogram' 'knn' 'lof' 'svm' 'pca'

'mcd'

'sod'

'sos'

'abod'

**plot= 'cluster' 'tsne'

'ap' 'meanshift' 'elbow' 'sc' 'silhouette' 'hclust' 'distance' 'dbscan' 'distribution'

'birch' 'kmodes'

set config()

save config()

load config()

get clusters()

* model:

'kmeans'

'optics'

Anomaly Detection

set up environment

from pycaret.anomaly import * clf1 = setup(data = df) # create and evaluate model plot model(model, plot=**)

supervised target='column')

model deployment

deploy model(model=model. model name=model final.

get metrics() add metric() remove_metric() get logs()

'S3-bucket-name'})

save config() load_config()

**plot=

'tsne' 'umap'

Natural Language Processing

set up environment

from pycaret.nlp import *

clf1 = setup(data=df, target='colunm') # create and evaluate model

model = create model(*)

model df = assign model(*) plot model(model, plot=**) evaluate model(model)

model tuned = tune model(model=model, supervised target='column')

model deployment

save_model(model, 'saved_model') model loaded = load model('saved model')

utils pull()

models() get logs() get config()

set config() get_topics()

> * model 'lda' 'lsi' 'hdp' 'rp' 'nmf' 'frequency' 'distribution' 'bigram' 'trigram' 'sentiment' 'pos' 'tsne'

'topic model' 'topic distribution' 'wordcloud' 'umap'

** plot= 'tsne' 'umap'

Association Rule

set up environment

from pycaret.arules import * clf1 = setup(data=df, transaction id='colunm',

item id='column')

create and evaluate model model = create model() plot model(model, plot='2d')

Other Resources

PyCaret Github PvCaert Slack

Example Notebooks made by contributors Documentation 'The detailed API docs of PyCaret'

Video Tutorials

Discussions 'Have questions?'

Changelog 'Changes and version history' Roadmap of PyCaret



Parameters of setup() and its default values

pycaret.org

Clustering

```
data,
preprocess = True,
imputation_type = 'simple',
iterative_imputation_iters = 5,
categorical features = None,
categorical imputation = 'mode',
categorical iterative imputer = 'lightqbm',
ordinal features = None,
high_cardinality_features = None,
high cardinality method = 'frequency',
numeric features = None,
numeric imputation = 'mean',
numeric iterative imputer = 'lightqbm',
date features = None,
ignore features = None.
normalize = False.
normalize method = 'zscore',
transformation = False,
transformation method = 'yeo-johnson',
handle unknown categorical = True,
unknown categorical method = 'least frequent',
pca = False.
pca method = 'linear',
pca components = None,
ignore low variance = False.
combine rare levels = False,
rare level threshold = 0.1,
bin numeric features = None,
remove multicollinearity = False.
multicollinearity threshold = 0.9.
remove perfect collinearity = False,
group features = None,
group_names = None,
n jobs = -1,
use gpu = False,
custom_pipeline = None,
html = True.
session id = None,
system log = True,
log experiment = False,
experiment name = None,
log plots = False,
log profile = False,
log data = False,
silent = False.
verbose = True.
profile = False,
profile_kwargs = None
```

Anomaly Detection

data.

```
Preprocess = True,
imputation_type = 'simple',
iterative_imputation_iters = 5,
categorical features = None,
categorical imputation = 'mode',
categorical iterative imputer = 'lightqbm',
ordinal features = None,
high cardinality features = None,
high cardinality method = 'frequency',
numeric features = None,
numeric imputation = 'mean',
numeric iterative imputer = 'lightqbm',
date features = None,
ignore features = None.
normalize = False.
normalize method = 'zscore',
transformation = False,
transformation method = 'yeo-johnson',
handle unknown categorical = True,
unknown categorical method = 'least frequent'.
pca = False.
pca method = 'linear',
pca components = None,
ignore low variance = False.
combine rare levels = False.
rare level threshold = 0.1,
bin numeric features = None,
remove multicollinearity = False,
multicollinearity threshold = 0.9.
remove perfect collinearity = False.
group features = None,
group_names = None,
n iobs = -1
use gpu = False,
custom pipeline = None,
html = True.
session id = None,
system log = True,
log experiment = False,
experiment name = None,
log plots = False,
log profile = False,
log data = False,
silent = False.
verbose = True.
profile = False,
profile kwargs = None
```

Regression & Classification

```
data = DataFrame, target = 'column name',
                                                create clusters = False,
train size = 0.7,
                                                cluster iter = 20,
test data = None,
                                                polynomial features = False,
preprocess = True.
                                                polynomial degree = 2.
imputation type = 'simple',
                                                trigonometry features = False,
iterative imputation iters = 5,
                                                polynomial threshold = 0.1,
categorical features = None,
                                                group features = None,
categorical imputation = 'constant'.
categorical iterative imputer = 'lightgbm',
                                                group names = None,
ordinal features = None,
                                                feature selection = False,
high cardinality features = None,
                                                feature selection threshold = 0.8,
high cardinality method = 'frequency',
                                                feature selection method = 'classic'.
numeric features = None.
                                                feature interaction = False,
numeric imputation = 'mean',
                                                feature ratio = False,
numeric iterative imputer = 'lightqbm',
                                                interaction threshold = 0.01.
date features = None,
ignore features = None,
                                                transform target = False.
normalize = False,
                                                transform target method = 'box-cox'.
normalize method = 'zscore'.
                                                data split shuffle = True,
transformation = False,
                                                data split stratify = False,
transformation method = 'yeo-johnson',
                                                fold strategy = 'kfold',
handle unknown categorical = True,
                                                fold = 10.
unknown categorical method = 'Least frequent'.
                                                fold shuffle = False.
pca = False.
pca method = 'Linear'.
                                                fold groups = None,
pca components = None,
                                                n jobs = -1,
ignore low variance = False,
                                                use gpu = False,
combine rare levels = False.
                                                custom pipeline = None,
rare level threshold = 0.1,
                                                html = True,
bin numeric features = None,
                                                session id = None,
remove outliers = False,
outliers threshold = 0.05,
                                                log experiment = False,
remove multicollinearity = False,
                                                experiment_name = None,
multicollinearity threshold = 0.9.
                                                log plots = False.
remove perfect collinearity = True,
                                                log profile = False,
                                                log data = False,
                                                silent = False,
                                                verbose = True.
                                                profile = False.
                                                profile kwargs = None
```

Time Series

```
data = [.Series, .DataFrame],
preprocess = True,
imputation type = 'simple',
fold strategy = 'expandina'.
fold = 3.
fh = 1,
seasonal period = None,
enforce pi = False,
n jobs = -1.
use gpu = False,
custom pipeline = None.
html = True.
session id = None,
system log = True,
log experiment = False.
experiment name = None,
log plots = False,
log_profile = False,
log data = False.
verbose = True.
profile = False.
profile kwargs = None
```

Association Rule

```
data,
transaction_id ='column_name',
item_id = 'column_name',
ignore_items = None,
session_id = None
```

NLP

```
Target = 'column_name',
custom_stopwords = None,
Html = True,
session_id = None,
log_experiment = False,
experiment_name = None,
log_plots = False,
log_data = False,
Verbose = True
```

Color code

required optional