# Classification using PyCaret

- It is used for Binary or Multi-class classification problems
- It has over 18 algorithms and 14 plots to analyze the performance of models.
- It has hyper-parameter tuning, ensembling or advanced techniques like stacking

### Steps:

- 1. Data Acquisition: import data from PyCaret or any other repository
- 2. **Setting up Environment**: How to setup an experiment in PyCaret and get started with building classification models
- 3. **Model Creation**: How to create a model, perform stratified cross validation and evaluate classification metrics
- 4. Model Tuning: How to automatically tune the hyper-parameters of a classification model
- 5. **Performance Analysis**: How to analyze model performance using various plots
- 6. **Finalize Model**: How to finalize the best model at the end of the experiment
- 7. **Predict Model**: How to make predictions on new / unseen data
- 8. Save / Load Model: How to save / load a model for future use

#### ▼ Install PyCaret

```
!pip install pycaret
```

```
Bullaing wheel for alembic (setup.py) ... done
 Created wheel for alembic: filename=alembic-1.4.1-py2.py3-none-any.whl size=1581
 Stored in directory: /root/.cache/pip/wheels/be/5d/0a/9e13f53f4f5dfb67cd8d245bb7
 Building wheel for databricks-cli (setup.py) ... done
 Created wheel for databricks-cli: filename=databricks cli-0.16.2-py3-none-any.wh
 Stored in directory: /root/.cache/pip/wheels/f4/5c/ed/e1ce20a53095f63b27b4964abb
 Building wheel for pyLDAvis (setup.py) ... done
 Created wheel for pyLDAvis: filename=pyLDAvis-3.2.2-py2.py3-none-any.whl size=13
 Stored in directory: /root/.cache/pip/wheels/f8/b1/9b/560ac1931796b7303f7b517b949
 Building wheel for pyod (setup.py) ... done
 Created wheel for pyod: filename=pyod-0.9.5-py3-none-any.whl size=132699 sha256=
 Stored in directory: /root/.cache/pip/wheels/3d/bb/b7/62b60fb451b33b0df1ab800669
 Building wheel for umap-learn (setup.py) ... done
 Created wheel for umap-learn: filename=umap learn-0.5.2-py3-none-any.whl size=82
 Stored in directory: /root/.cache/pip/wheels/84/1b/c6/aaf68a748122632967cef4dffe-
 Building wheel for pynndescent (setup.py) ... done
 Created wheel for pynndescent: filename=pynndescent-0.5.5-py3-none-any.whl size=
 Stored in directory: /root/.cache/pip/wheels/af/e9/33/04db1436df0757c42fda8ea679
Successfully built htmlmin imagehash alembic databricks-cli pyLDAvis pyod umap-lea
Installing collected packages: threadpoolctl, tangled-up-in-unicode, smmap, scipy,
 Attempting uninstall: scipy
    Found existing installation: scipy 1.4.1
    Uninstalling scinv-1 4 1.
```

```
OHITHS CALLTING SCIPY I.T.I.
      Successfully uninstalled scipy-1.4.1
  Attempting uninstall: scikit-learn
    Found existing installation: scikit-learn 0.22.2.post1
    Uninstalling scikit-learn-0.22.2.post1:
      Successfully uninstalled scikit-learn-0.22.2.post1
  Attempting uninstall: requests
    Found existing installation: requests 2.23.0
    Uninstalling requests-2.23.0:
      Successfully uninstalled requests-2.23.0
  Attempting uninstall: PyYAML
    Found existing installation: PyYAML 3.13
    Uninstalling PyYAML-3.13:
      Successfully uninstalled PyYAML-3.13
  Attempting uninstall: yellowbrick
    Found existing installation: yellowbrick 0.9.1
    Uninstalling yellowbrick-0.9.1:
      Successfully uninstalled yellowbrick-0.9.1
  Attempting uninstall: pandas-profiling
    Found existing installation: pandas-profiling 1.4.1
    Uninstalling pandas-profiling-1.4.1:
      Successfully uninstalled pandas-profiling-1.4.1
  Attempting uninstall: mlxtend
    Found existing installation: mlxtend 0.14.0
    Uninstalling mlxtend-0.14.0:
      Successfully uninstalled mlxtend-0.14.0
  Attempting uninstall: lightgbm
    Found existing installation: lightgbm 2.2.3
    Uninstalling lightgbm-2.2.3:
      Successfully uninstalled lightgbm-2.2.3
  Attempting uninstall: imbalanced-learn
    Found existing installation: imbalanced-learn 0.4.3
    Uninstalling imbalanced-learn-0.4.3:
      Successfully uninstalled imbalanced-learn-0.4.3
ERROR: pip's dependency resolver does not currently take into account all the pack
google-colab 1.0.0 requires requests~=2.23.0, but you have requests 2.26.0 which i
datascience 0.10.6 requires folium==0.2.1. but vou have folium 0.8.3 which is inco

▼
```

#### Dataset source

https://archive.ics.uci.edu/ml/datasets/Heart+failure+clinical+records

#### Upload dataset from user-system

```
from google.colab import files
files.upload()
```

```
Choose files No file chosen Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to enable.

Saving heart_failure_dataset.csv to heart_failure_dataset.csv {'heart_failure_dataset.csv': b'age,anaemia,creatinine_phosphokinase,diabetes,ejec
```

#### Read dataset

```
import pandas as pd
dataset = pd.read_csv('heart_failure_dataset.csv')
dataset.head()
```

	age	anaemia	creatinine_phosphokinase	diabetes	ejection_fraction	high_bloo
0	75.0	0	582	0	20	
1	55.0	0	7861	0	38	
2	65.0	0	146	0	20	
3	50.0	1	111	0	20	
4	65.0	1	160	1	20	

### Upload dataset to google drive,

### mount google drive and read dataset

```
from google.colab import drive
drive.mount('/content/drive')
Dataset = pd.read_csv('/content/heart_failure_dataset.csv')

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.m
```

# Load Datasets from PyCaret

- 1. from pycaret.datasets import get\_data
- 2. datasets = get\_data('index')
- 3. dataset = get\_data("diabetes")

# import all classification methods

from pycaret.classification import \*

# **▼** Setup function initializes the training environment

s= setup(data=dataset, target='DEATH\_EVENT', train\_size=0.75, silent=True)

	Description	Value
0	session_id	8441
1	Target	DEATH_EVENT
2	Target Type	Binary
3	Label Encoded	None
4	Original Data	(291, 13)
5	Missing Values	False
6	Numeric Features	6
7	Categorical Features	6
8	Ordinal Features	False
9	High Cardinality Features	False
10	High Cardinality Method	None
11	Transformed Train Set	(218, 26)
12	Transformed Test Set	(73, 26)
13	Shuffle Train-Test	True
14	Stratify Train-Test	False
15	Fold Generator	StratifiedKFold
16	Fold Number	10
17	CPU Jobs	-1
18	Use GPU	False
19	Log Experiment	False
20	Experiment Name	clf-default-name
21	USI	f80e
22	Imputation Type	simple
23	Iterative Imputation Iteration	None
24	Numeric Imputer	mean
25	Iterative Imputation Numeric Model	None
26	Categorical Imputer	constant
27	Iterative Imputation Categorical Model	None
28	Unknown Categoricals Handling	least_frequent
29	Normalize	False
30	Normalize Method	None

# ▼ Run and Compare all ML models

cm = compare\_models()

	Model	Accuracy	AUC	Recall	Prec.	F1	Карра	MCC	(Se
rf	Random Forest Classifier	0.8530	0.8942	0.6238	0.8614	0.7143	0.6206	0.6407	0.4
lightgbm	Light Gradient Boosting Machine	0.8258	0.8605	0.6548	0.7556	0.6880	0.5703	0.5818	0.0
gbc	Gradient Boosting Classifier	0.8115	0.8783	0.6071	0.7302	0.6523	0.5264	0.5379	0.0
lr	Logistic Regression	0.8071	0.8283	0.5952	0.7414	0.6506	0.5206	0.5331	0.3
ada	Ada Boost Classifier	0.8024	0.8260	0.5929	0.7073	0.6404	0.5070	0.5130	0.0
ridge	Ridge Classifier	0.7883	0.0000	0.5881	0.6630	0.6045	0.4655	0.4797	0.0
nb	Naive Bayes	0.7840	0.8131	0.4738	0.7267	0.5597	0.4291	0.4522	0.0
49	Tria	tures	False						

# ▼ Create a model - Random Forest model

m = create\_model('rf')
print (m)

	Accuracy	AUC	Recall	Prec.	F1	Карра	MCC
0	0.9091	0.9688	0.6667	1.0000	0.8000	0.7442	0.7698
1	0.9091	0.8857	0.8571	0.8571	0.8571	0.7905	0.7905
2	0.8182	0.8571	0.5714	0.8000	0.6667	0.5464	0.5610
3	0.8636	0.8667	0.5714	1.0000	0.7273	0.6452	0.6901
4	0.7727	0.8905	0.4286	0.7500	0.5455	0.4086	0.4370
_						· ·	

#### Tune the Model

```
tuned_m = tune_model(m,n_iter = 50)
print(tuned_m)
```

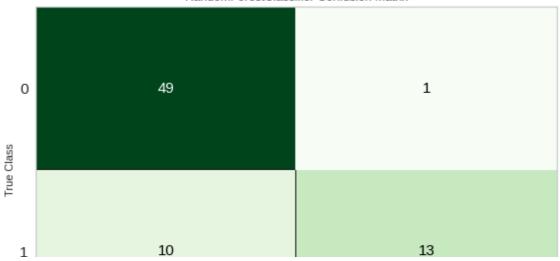
	Accuracy	AUC	Recall	Prec.	F1	Карра	MCC
0	0.9091	0.9583	0.6667	1.0000	0.8000	0.7442	0.7698
1	0.9091	0.8857	0.8571	0.8571	0.8571	0.7905	0.7905
2	0.7273	0.8476	0.2857	0.6667	0.4000	0.2584	0.2973
3	0.8636	0.8952	0.5714	1.0000	0.7273	0.6452	0.6901
4	0.7273	0.9048	0.2857	0.6667	0.4000	0.2584	0.2973
5	0.9545	0.9714	0.8571	1.0000	0.9231	0.8911	0.8964
6	0.9545	1.0000	1.0000	0.8750	0.9333	0.8991	0.9037
7	0.7273	0.8381	0.4286	0.6000	0.5000	0.3196	0.3281
8	0.8571	0.9111	0.5000	1.0000	0.6667	0.5882	0.6455
9	0.8571	0.7778	0.5000	1.0000	0.6667	0.5882	0.6455
Mean	0.8487	0.8990	0.5952	0.8665	0.6874	0.5983	0.6264
SD	0.0862	0.0635	0.2333	0.1550	0.1898	0.2336	0.2253

# Performance Analysis

evaluate\_model(tuned\_m)



#### RandomForestClassifier Confusion Matrix



Double-click (or enter) to edit

# Predict using Tunned Model

predict\_model(tuned\_m);

	Model	Accuracy	AUC	Recall	Prec.	F1	Карра	MCC
0	Random Forest Classifier	0.8493	0.9226	0.5652	0.9286	0.7027	0.6096	0.6433

# Finalize the Model

final\_m = finalize\_model(tuned\_m)

### Save the Model

```
save model(final m, 'Final-Model')
     Transformation Pipeline and Model Successfully Saved
     (Pipeline(memory=None,
               steps=[('dtypes',
                       DataTypes_Auto_infer(categorical_features=[],
                                             display_types=False, features_todrop=[],
                                             id_columns=[],
                                             ml usecase='classification',
                                             numerical_features=[],
                                             target='DEATH_EVENT', time_features=[])),
                      ('imputer',
                       Simple_Imputer(categorical_strategy='not_available',
                                      fill_value_categorical=None,
                                      fill value numerical=None,
                                       numeric...
                       RandomForestClassifier(bootstrap=True, ccp_alpha=0.0,
                                               class_weight={}, criterion='gini',
                                               max_depth=7, max_features='log2',
                                               max leaf nodes=None, max samples=None,
                                               min_impurity_decrease=0.0005,
                                               min impurity split=None,
                                               min_samples_leaf=4, min_samples_split=5,
                                               min_weight_fraction_leaf=0.0,
                                               n_estimators=170, n_jobs=-1,
                                               oob_score=False, random_state=8441,
                                               verbose=0, warm_start=False)]],
               verbose=False), 'Final-Model.pkl')
```

#### Download Pickle File

```
from google.colab import files
files.download('Final-Model.pkl')
```

## Upload Pickle File

#### Load the Model

```
saved_final_m = load_model('Final-Model')
```

Transformation Pipeline and Model Successfully Loaded

# Predict using unseen data

files.upload()

new\_prediction.head()

	age	anaemia	<pre>creatinine_phosphokinase</pre>	diabetes	ejection_fraction	high_blood
0	70	1	125	0	25	
1	48	1	582	1	55	
2	65	1	52	0	25	
3	65	1	128	1	30	
4	68	1	220	0	35	

Basic Level: https://www.pycaret.org/tutorials/html/CLF101.html

Intermediate Level: <a href="https://www.pycaret.org/tutorials/html/CLF102.html">https://www.pycaret.org/tutorials/html/CLF102.html</a>

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