ML Raport

AutoPrep

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Abstract

This raport has been generated with AutoPrep.

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1 Overview

1.1 System

Darwin
arm64
arm
64bit
3.11.10
8
8
4056
16.00
4.37
460.43
247.59

Table 1: System overview.

1.2 Dataset

Number of samples	1047
Number of features	13
Number of numerical features	6
Number of categorical features	7

Table 2: Dataset Summary.

class	number of observations	Percentage	
0	665	0.64	
1	382	0.36	

Table 3: Target class distribution.

classgit	number of observations	Percentage
pclass	0	0.00
name	0	0.00
sex	0	0.00
age	207	0.20
sibsp	0	0.00
parch	0	0.00
ticket	0	0.00
fare	1	0.00
cabin	813	0.78
embarked	1	0.00
boat	672	0.64
body	948	0.91
homedest	453	0.43

Table 4: Missing values distribution.

class	type	dtype	space usage
pclass	numerical	uint8	9.4 kB
name	categorical	object	$96.4~\mathrm{kB}$
sex	categorical	category	$9.7~\mathrm{kB}$
age	numerical	float64	$16.8~\mathrm{kB}$
sibsp	numerical	uint8	$9.4~\mathrm{kB}$
parch	numerical	uint8	$9.4~\mathrm{kB}$
ticket	categorical	object	$75.1~\mathrm{kB}$
fare	numerical	float64	$16.8~\mathrm{kB}$
cabin	categorical	object	$42.1~\mathrm{kB}$
$\operatorname{embarked}$	categorical	category	$9.7~\mathrm{kB}$
boat	categorical	object	$46.4~\mathrm{kB}$
body	numerical	float64	$16.8~\mathrm{kB}$
homedest	categorical	object	64.5 kB

Table 5: Features dtypes description.

index	count	mean	std	min	25%	50%	75%	max
pclass	1047.00	2.30	0.84	1.00	2.00	3.00	3.00	3.00
age	840.00	29.53	14.27	0.17	21.00	28.00	38.62	80.00
sibsp	1047.00	0.52	1.05	0.00	0.00	0.00	1.00	8.00
parch	1047.00	0.40	0.89	0.00	0.00	0.00	0.00	9.00
fare	1046.00	33.55	51.81	0.00	7.92	14.50	31.27	512.33
body	99.00	160.90	98.35	1.00	73.50	156.00	255.50	328.00

Table 6: Numerical features description.

index	count	unique	top	freq
name	1047	1046	Connolly, Miss. Kate	2
ticket	1047	773	CA. 2343	9
cabin	234	161	B57 B59 B63 B66	5
boat	375	25	13	34
$\underline{ \text{home}\underline{ }} \text{dest}$	594	317	New York, NY	50

Table 7: Categorical features description.

2 Eda

This part of the report provides basic insides to the data and the informations it holds..

2.1 Eda

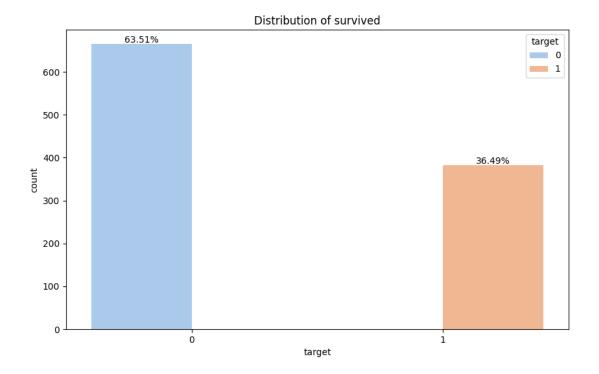


Figure 1: Target distribution.

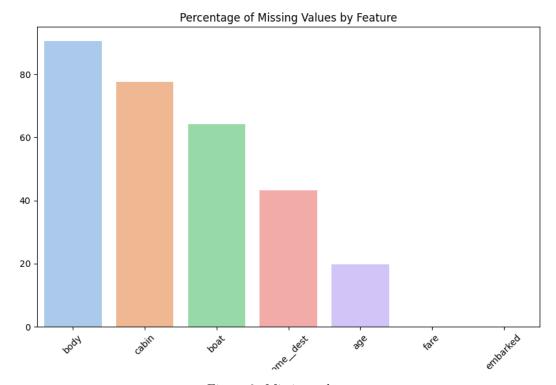


Figure 2: Missing values.

2.2 Categorical

Categorical Features Distribution - Page 1

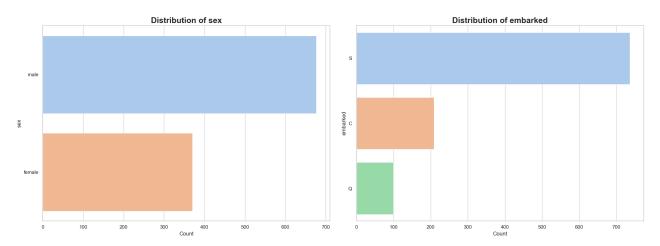


Figure 3: Categorical Features Distribution - Page 1 $\,$

2.3 Numerical

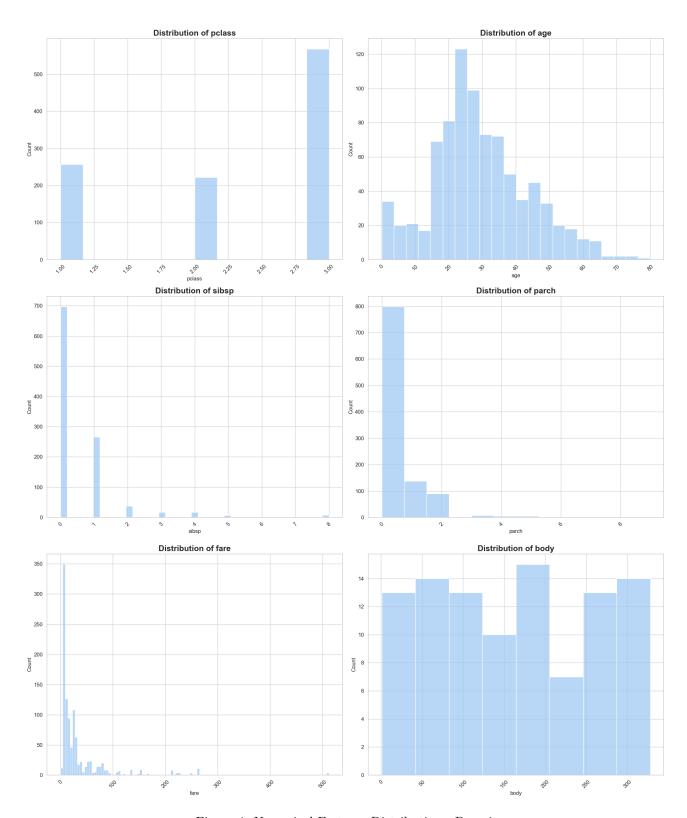


Figure 4: Numerical Features Distribution - Page 1

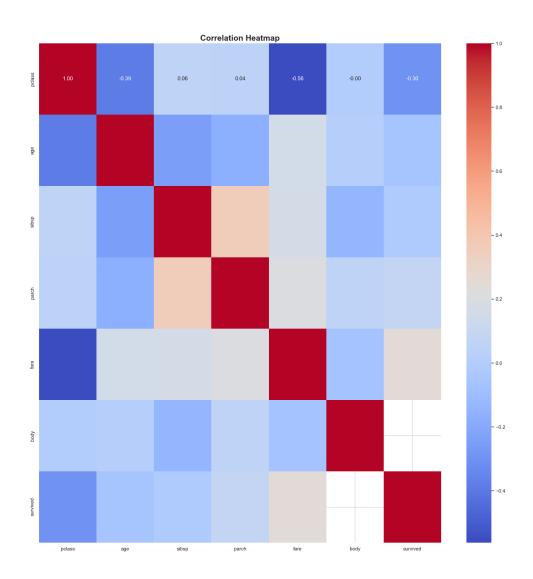


Figure 5: Correlation heatmap.

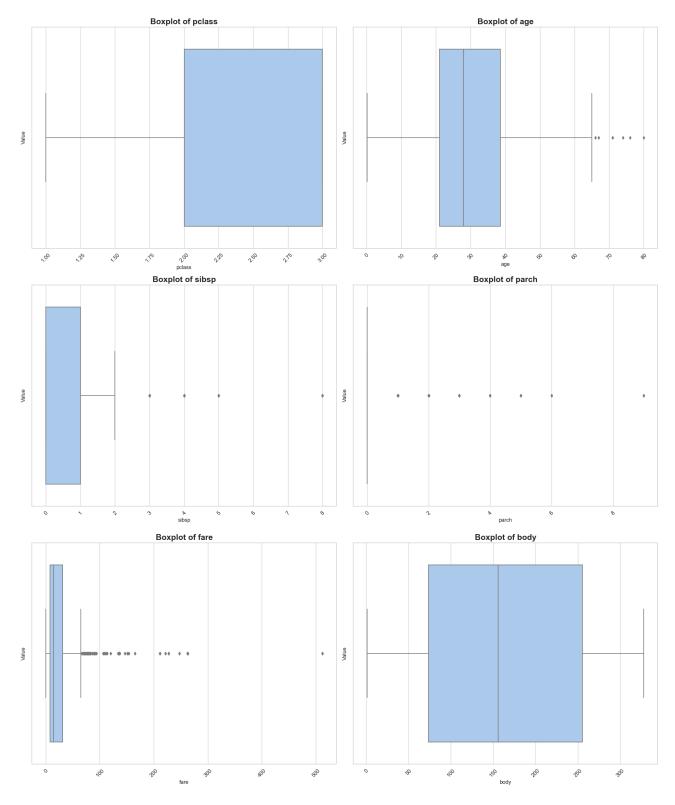


Figure 6: Boxplot page 1

3 Preprocessing

This part of the report presents the results of the preprocessing process. It was configured to create up to 3 unique preprocessing pipelines.

Category	Value
Unique created pipelines	2
All created pipelines (after exploading each step params)	6
All pipelines fit time	2 seconds
All pipelines score time	2 seconds
scores_count	6.00
scores_mean	0.76
$scores_std$	0.01
scores_min	0.75
$scores_25\%$	0.75
$scores_50\%$	0.76
$scores_75\%$	0.77
scores_max	0.78
Scoring function	<class 'str' $>$
Scoring model	${\bf Random Forest Classifier}$

Table 8: Preprocessing pipelines runtime statistics.

index	steps
0	NAImputer, UniqueFilter, ColumnEncoder, ColumnScaler, CorrelationFilter
1	NAImputer, UniqueFilter, ColumnEncoder, ColumnScaler, VarianceFilter

Table 9: Pipelines steps overview.

score index	file name	score	fit duration	score duration
0	preprocessing_pipeline_0.joblib	0.78	a moment	a moment
1	preprocessing_pipeline_1.joblib	0.77	a moment	a moment
2	preprocessing_pipeline_2.joblib	0.77	a moment	a moment

Table 10: Best preprocessing pipelines.

step	name	description	params
0	NAImputer	Imputes missing data.	{"numeric_imputer": "median", "categorical_imputer": "most_frequent"}
1	UniqueFilter	Removes categorical columns with 100% unique values. Dropped columns:	{}
2	ColumnEncoder	Encodes categorical columns using OneHotEncoder (for columns with <5 unique values) or TolerantLabelEncoder (for columns with >=5 unique values). Encodes target variable using LabelEncoder if provided.	{}
3	ColumnScaler	Scales numerical columns using one of 3 scaling methods.	$\{"method": "robust"\}$
4	VarianceFilter	Removes columns with zero variance. Dropped columns: []	{}

Table 11: 0th best pipeline overwiev on training set.

index	count	mean	std	min	25%	50%	75%	max
pclass	1047.00	0.00	1.00	-1.55	-0.36	0.84	0.84	0.84
name	1047.00	0.00	1.00	-1.73	-0.87	-0.00	0.87	1.73
age	1047.00	-0.00	1.00	-2.27	-0.57	-0.10	0.45	3.97
sibsp	1047.00	-0.00	1.00	-0.50	-0.50	-0.50	0.46	7.13
parch	1047.00	0.00	1.00	-0.44	-0.44	-0.44	-0.44	9.63
ticket	1047.00	-0.00	1.00	-1.68	-0.90	0.00	0.93	1.67
fare	1047.00	0.00	1.00	-0.65	-0.49	-0.37	-0.04	9.25
$home\dest$	1047.00	-0.00	1.00	-2.72	-0.18	0.23	0.30	2.01
sex_female	1047.00	0.00	1.00	-0.74	-0.74	-0.74	1.35	1.35
$embarked_C$	1047.00	-0.00	1.00	-0.50	-0.50	-0.50	-0.50	2.00
$embarked_Q$	1047.00	0.00	1.00	-0.32	-0.32	-0.32	-0.32	3.08
$- \underline{\text{embarked} \underline{S}}$	1047.00	-0.00	1.00	-1.55	-1.55	0.65	0.65	0.65

Table 12: 0th best pipeline output overview.

$\overline{\text{step}}$	name	description	params
0	NAImputer	Imputes missing data.	{"numeric_imputer": "median", "cate-
			gorical_imputer": "most_frequent"}
1	UniqueFilter	Removes categorical columns with	{}
		100% unique values. Dropped columns:	
2	ColumnEncoder	Encodes categorical columns using	{}
		OneHotEncoder (for columns with <5	
		unique values) or TolerantLabelEn-	
		coder (for columns with $>=5$ unique	
		values). Encodes target variable using	
		LabelEncoder if provided.	
3	ColumnScaler	Scales numerical columns using one of	${\text{"method": "standard"}}$
		3 scaling methods.	
4	VarianceFilter	Removes columns with zero variance.	{}
		Dropped columns: []	

Table 13: 1th best pipeline overwiev on training set.

index	count	mean	std	min	25%	50%	75%	max
pclass	1047.00	0.65	0.42	0.00	0.50	1.00	1.00	1.00
name	1047.00	0.50	0.29	0.00	0.25	0.50	0.75	1.00
age	1047.00	0.36	0.16	0.00	0.27	0.35	0.44	1.00
sibsp	1047.00	0.07	0.13	0.00	0.00	0.00	0.12	1.00
parch	1047.00	0.04	0.10	0.00	0.00	0.00	0.00	1.00
ticket	1047.00	0.50	0.30	0.00	0.23	0.50	0.78	1.00
fare	1047.00	0.07	0.10	0.00	0.02	0.03	0.06	1.00
$home\dest$	1047.00	0.58	0.21	0.00	0.54	0.62	0.64	1.00
sex_female	1047.00	0.35	0.48	0.00	0.00	0.00	1.00	1.00
$embarked_C$	1047.00	0.20	0.40	0.00	0.00	0.00	0.00	1.00
${\rm embarked} _{\rm Q}$	1047.00	0.10	0.29	0.00	0.00	0.00	0.00	1.00
$_{\rm embarked}_{\rm S}$	1047.00	0.70	0.46	0.00	0.00	1.00	1.00	1.00

Table 14: 1th best pipeline output overview.

step	name	description	params
0	NAImputer	Imputes missing data.	{"numeric_imputer": "median", "cate-
			gorical_imputer": "most_frequent"}
1	UniqueFilter	Removes categorical columns with	{}
		100% unique values. Dropped columns:	
2	ColumnEncoder	Encodes categorical columns using	{}
		OneHotEncoder (for columns with <5	
		unique values) or TolerantLabelEn-	
		coder (for columns with $>=5$ unique	
		values). Encodes target variable using	
		LabelEncoder if provided.	
3	ColumnScaler	Scales numerical columns using one of	${\text{"method": "minmax"}}$
		3 scaling methods.	
4	VarianceFilter	Removes columns with zero variance.	{}
		Dropped columns: []	

Table 15: 2th best pipeline overwiev on training set.

index	count	mean	std	min	25%	50%	75%	max
pclass	1047.00	-0.70	0.84	-2.00	-1.00	0.00	0.00	0.00
name	1047.00	0.00	0.58	-1.00	-0.50	0.00	0.50	1.00
age	1047.00	0.09	0.98	-2.14	-0.46	0.00	0.54	4.00
sibsp	1047.00	0.52	1.05	0.00	0.00	0.00	1.00	8.00
parch	1047.00	0.40	0.89	0.00	0.00	0.00	0.00	9.00
ticket	1047.00	-0.00	0.55	-0.92	-0.49	0.00	0.51	0.91
fare	1047.00	0.81	2.22	-0.62	-0.28	0.00	0.72	21.32
$home\dest$	1047.00	-0.48	2.06	-6.09	-0.86	0.00	0.14	3.66
sex_female	1047.00	0.35	0.48	0.00	0.00	0.00	1.00	1.00
$embarked_C$	1047.00	0.20	0.40	0.00	0.00	0.00	0.00	1.00
$\rm embarked_Q$	1047.00	0.10	0.29	0.00	0.00	0.00	0.00	1.00
$embarked_S$	1047.00	-0.30	0.46	-1.00	-1.00	0.00	0.00	0.00

Table 16: 2th best pipeline output overview.

4 Modeling

This part of the report presents the results of the modeling process. It was configured to create up to 3 models.

4.1 Overview

Category	Value
task	classification
unique models param sets	43
checked (for each dataset)	
unique models	5
scoring function	roc_auc_score
search parameters	{"cv": 3, "verbose": 0, "n_jobs": -1, "random_state": 42, "n_iter": 10}
train	1047 samples, 13 features
valid	131 samples, 13 features
test	131 samples, 13 features

Table 17: General input data overview.

name	unique params distributions checked
ModelKNeighboursClassifier	10
ModelLogisticRegression	10
${\bf Model Gaussian Naive Classifier}$	3
ModelSVC	10
${\bf Model Decision Tree Classifier}$	10

Table 18: Used models.

4.2 Scores for 0th best model

- final pipeline name: final_pipeline_0.joblib
- name: ModelKNeighboursClassifier
- params: {"weights": "distance", "p": 1, "n_neighbors": 15, "leaf_size": 30, "algorithm": "brute"}
- combined Score (after re-training): 1.0
- \bullet mean_test_score: nan
- std test score: nan
- test Score (after re-training): 0.71
- mean_fit_time: a moment
- re-training time: a moment
- std fit time: a moment

4.3 Scores for 1th best model

- final pipeline name: final_pipeline_1.joblib
- name: ModelKNeighboursClassifier
- params: {"weights": "distance", "p": 2, "n_neighbors": 10, "leaf_size": 40, "algorithm": "auto"}
- combined Score (after re-training): 1.0
- \bullet mean_test_score: nan
- std test score: nan
- test Score (after re-training): 0.69
- \bullet mean_fit_time: a moment
- re-training time: a moment
- std_fit_time: a moment

4.4 Scores for 2th best model

- final pipeline name: final_pipeline_2.joblib
- name: ModelKNeighboursClassifier
- params: {"weights": "uniform", "p": 2, "n_neighbors": 15, "leaf_size": 30, "algorithm": "kd_tree"}
- combined Score (after re-training): 0.74
- \bullet mean_test_score: nan
- std_test_score: nan
- test Score (after re-training): 0.68

 $\bullet \ \ \mathrm{mean_fit_time:} \ \mathrm{a} \ \mathrm{moment}$

 $\bullet\,$ re-training time: a moment

 $\bullet \ \, {\rm std_fit_time:} \ \, a \ \, {\rm moment}$