
Lazy Predict Documentation

Release 0.2.9

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CHAPTER 1

Lazy Predict

Lazy Predict helps build a lot of basic models without much code and helps understand which models works better without any parameter tuning.

- Free software: MIT license
- Documentation: <https://lazypredict.readthedocs.io>.

CHAPTER 2

Installation

To install Lazy Predict:

```
pip install lazypredict
```


CHAPTER 3

Usage

To use Lazy Predict in a project:

```
import lazypredict
```


Example

```

from lazypredict.Supervised import LazyClassifier
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split

data = load_breast_cancer()
X = data.data
y = data.target

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=.5, random_state=
↪=123)

clf = LazyClassifier(verbose=0, ignore_warnings=True, custom_metric=None)
models, predictions = clf.fit(X_train, X_test, y_train, y_test)

print(models)

```

Model	F1 Score	Time Taken	Accuracy	Balanced Accuracy	ROC AUC	
LinearSVC	↪989462	0.0150008	0.989474	0.987544	0.987544	0.
SGDClassifier	↪989462	0.0109992	0.989474	0.987544	0.987544	0.
MLPClassifier	↪985994	0.426	0.985965	0.986904	0.986904	0.
Perceptron	↪985965	0.0120046	0.985965	0.984797	0.984797	0.
LogisticRegression	↪985934	0.0200036	0.985965	0.98269	0.98269	0.
LogisticRegressionCV	↪985934	0.262997	0.985965	0.98269	0.98269	0.

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SVC		0.982456	0.979942	0.979942	0.
↪982437	0.0140011				
CalibratedClassifierCV		0.982456	0.975728	0.975728	0.
↪982357	0.0350015				
PassiveAggressiveClassifier		0.975439	0.974448	0.974448	0.
↪975464	0.0130005				
LabelPropagation		0.975439	0.974448	0.974448	0.
↪975464	0.0429988				
LabelSpreading		0.975439	0.974448	0.974448	0.
↪975464	0.0310006				
RandomForestClassifier		0.97193	0.969594	0.969594	0.
↪97193	0.033				
GradientBoostingClassifier		0.97193	0.967486	0.967486	0.
↪971869	0.166998				
QuadraticDiscriminantAnalysis		0.964912	0.966206	0.966206	0.
↪965052	0.0119994				
HistGradientBoostingClassifier		0.968421	0.964739	0.964739	0.
↪968387	0.682003				
RidgeClassifierCV		0.97193	0.963272	0.963272	0.
↪971736	0.0130029				
RidgeClassifier		0.968421	0.960525	0.960525	0.
↪968242	0.0119977				
AdaBoostClassifier		0.961404	0.959245	0.959245	0.
↪961444	0.204998				
ExtraTreesClassifier		0.961404	0.957138	0.957138	0.
↪961362	0.0270066				
KNeighborsClassifier		0.961404	0.95503	0.95503	0.
↪961276	0.0560005				
BaggingClassifier		0.947368	0.954577	0.954577	0.
↪947882	0.0559971				
BernoulliNB		0.950877	0.951003	0.951003	0.
↪951072	0.0169988				
LinearDiscriminantAnalysis		0.961404	0.950816	0.950816	0.
↪961089	0.0199995				
GaussianNB		0.954386	0.949536	0.949536	0.
↪954337	0.0139935				
NuSVC		0.954386	0.943215	0.943215	0.
↪954014	0.019989				
DecisionTreeClassifier		0.936842	0.933693	0.933693	0.
↪936971	0.0170023				
NearestCentroid		0.947368	0.933506	0.933506	0.
↪946801	0.0160074				
ExtraTreeClassifier		0.922807	0.912168	0.912168	0.
↪922462	0.0109999				
CheckingClassifier		0.361404	0.5	0.5	0.
↪191879	0.0170043				
DummyClassifier		0.512281	0.489598	0.489598	0.
↪518924	0.0119965				

Example

```

from lazypredict.Supervised import LazyRegressor
from sklearn import datasets
from sklearn.utils import shuffle
import numpy as np

boston = datasets.load_boston()
X, y = shuffle(boston.data, boston.target, random_state=13)
X = X.astype(np.float32)

offset = int(X.shape[0] * 0.9)

X_train, y_train = X[:offset], y[:offset]
X_test, y_test = X[offset:], y[offset:]

reg = LazyRegressor(verbose=0, ignore_warnings=False, custom_metric=None)
models, predictions = reg.fit(X_train, X_test, y_train, y_test)

print(models)

```

Model	Adjusted R-Squared	R-Squared	RMSE	Time Taken
SVR	0.83	0.88	2.62	0.01
BaggingRegressor	0.83	0.88	2.63	0.03
NuSVR	0.82	0.86	2.76	0.03
RandomForestRegressor	0.81	0.86	2.78	0.21

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XGBRegressor		0.81		0.86		2.79		0.06	↵
↵									
GradientBoostingRegressor		0.81		0.86		2.84		0.11	↵
↵									
ExtraTreesRegressor		0.79		0.84		2.98		0.12	↵
↵									
AdaBoostRegressor		0.78		0.83		3.04		0.07	↵
↵									
HistGradientBoostingRegressor		0.77		0.83		3.06		0.17	↵
↵									
PoissonRegressor		0.77		0.83		3.11		0.01	↵
↵									
LGBMRegressor		0.77		0.83		3.11		0.07	↵
↵									
KNeighborsRegressor		0.77		0.83		3.12		0.01	↵
↵									
DecisionTreeRegressor		0.65		0.74		3.79		0.01	↵
↵									
MLPRegressor		0.65		0.74		3.80		1.63	↵
↵									
HuberRegressor		0.64		0.74		3.84		0.01	↵
↵									
GammaRegressor		0.64		0.73		3.88		0.01	↵
↵									
LinearSVR		0.62		0.72		3.96		0.01	↵
↵									
RidgeCV		0.62		0.72		3.97		0.01	↵
↵									
BayesianRidge		0.62		0.72		3.97		0.01	↵
↵									
Ridge		0.62		0.72		3.97		0.01	↵
↵									
TransformedTargetRegressor		0.62		0.72		3.97		0.01	↵
↵									
LinearRegression		0.62		0.72		3.97		0.01	↵
↵									
ElasticNetCV		0.62		0.72		3.98		0.04	↵
↵									
LassoCV		0.62		0.72		3.98		0.06	↵
↵									
LassoLarsIC		0.62		0.72		3.98		0.01	↵
↵									
LassoLarsCV		0.62		0.72		3.98		0.02	↵
↵									
Lars		0.61		0.72		3.99		0.01	↵
↵									
LarsCV		0.61		0.71		4.02		0.04	↵
↵									
SGDRegressor		0.60		0.70		4.07		0.01	↵
↵									
TweedieRegressor		0.59		0.70		4.12		0.01	↵
↵									
GeneralizedLinearRegressor		0.59		0.70		4.12		0.01	↵
↵									
ElasticNet		0.58		0.69		4.16		0.01	↵
↵									
Lasso		0.54		0.66		4.35		0.02	↵
↵									

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RANSACRegressor		0.53		0.65		4.41		0.04	↵
↵									
OrthogonalMatchingPursuitCV		0.45		0.59		4.78		0.02	↵
↵									
PassiveAggressiveRegressor		0.37		0.54		5.09		0.01	↵
↵									
GaussianProcessRegressor		0.23		0.43		5.65		0.03	↵
↵									
OrthogonalMatchingPursuit		0.16		0.38		5.89		0.01	↵
↵									
ExtraTreeRegressor		0.08		0.32		6.17		0.01	↵
↵									
DummyRegressor		-0.38		-0.02		7.56		0.01	↵
↵									
LassoLars		-0.38		-0.02		7.56		0.01	↵
↵									
KernelRidge		-11.50		-8.25		22.74		0.01	↵
↵									

Warning: Regression and Classification are replaced with LazyRegressor and LazyClassifier. Regression and Classification classes will be removed in next release

6.1 Stable release

To install Lazy Predict, run this command in your terminal:

```
$ pip install lazypredict
```

This is the preferred method to install Lazy Predict, as it will always install the most recent stable release.

If you don't have `pip` installed, this [Python installation guide](#) can guide you through the process.

6.2 From sources

The sources for Lazy Predict can be downloaded from the [Github repo](#).

You can either clone the public repository:

```
$ git clone git://github.com/shankarpandala/lazypredict
```

Or download the [tarball](#):

```
$ curl -OJL https://github.com/shankarpandala/lazypredict/tarball/master
```

Once you have a copy of the source, you can install it with:

```
$ python setup.py install
```


CHAPTER 7

Usage

To use Lazy Predict in a project:

```
import lazypredict
```


Example

```

from lazypredict.Supervised import LazyClassifier
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
data = load_breast_cancer()
X = data.data
y = data.target
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=.5, random_state=
↳=123)
clf = LazyClassifier(verbose=0, ignore_warnings=True, custom_metric=None)
models, predictions = clf.fit(X_train, X_test, y_train, y_test)
models

```

Model	Accuracy	Balanced Accuracy	ROC AUC	F1 Score	Time Taken
LinearSVC	0.989474	0.987544	0.987544	0.989462	0.0150008
SGDClassifier	0.989474	0.987544	0.987544	0.989462	0.0109992
MLPClassifier	0.985965	0.986904	0.986904	0.985994	0.426
Perceptron	0.985965	0.984797	0.984797	0.985965	0.0120046
LogisticRegression	0.985965	0.98269	0.98269	0.985934	0.0200036
LogisticRegressionCV	0.985965	0.98269	0.98269	0.985934	0.262997
SVC	0.982456	0.979942	0.979942	0.982437	0.0140011
CalibratedClassifierCV	0.982456	0.975728	0.975728	0.982357	0.0350015

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PassiveAggressiveClassifier	0.975439	0.974448	0.974448	0.
↪975464 0.0130005				
LabelPropagation	0.975439	0.974448	0.974448	0.
↪975464 0.0429988				
LabelSpreading	0.975439	0.974448	0.974448	0.
↪975464 0.0310006				
RandomForestClassifier	0.97193	0.969594	0.969594	0.
↪97193 0.033				
GradientBoostingClassifier	0.97193	0.967486	0.967486	0.
↪971869 0.166998				
QuadraticDiscriminantAnalysis	0.964912	0.966206	0.966206	0.
↪965052 0.0119994				
HistGradientBoostingClassifier	0.968421	0.964739	0.964739	0.
↪968387 0.682003				
RidgeClassifierCV	0.97193	0.963272	0.963272	0.
↪971736 0.0130029				
RidgeClassifier	0.968421	0.960525	0.960525	0.
↪968242 0.0119977				
AdaBoostClassifier	0.961404	0.959245	0.959245	0.
↪961444 0.204998				
ExtraTreesClassifier	0.961404	0.957138	0.957138	0.
↪961362 0.0270066				
KNeighborsClassifier	0.961404	0.95503	0.95503	0.
↪961276 0.0560005				
BaggingClassifier	0.947368	0.954577	0.954577	0.
↪947882 0.0559971				
BernoulliNB	0.950877	0.951003	0.951003	0.
↪951072 0.0169988				
LinearDiscriminantAnalysis	0.961404	0.950816	0.950816	0.
↪961089 0.0199995				
GaussianNB	0.954386	0.949536	0.949536	0.
↪954337 0.0139935				
NuSVC	0.954386	0.943215	0.943215	0.
↪954014 0.019989				
DecisionTreeClassifier	0.936842	0.933693	0.933693	0.
↪936971 0.0170023				
NearestCentroid	0.947368	0.933506	0.933506	0.
↪946801 0.0160074				
ExtraTreeClassifier	0.922807	0.912168	0.912168	0.
↪922462 0.0109999				
CheckingClassifier	0.361404	0.5	0.5	0.
↪191879 0.0170043				
DummyClassifier	0.512281	0.489598	0.489598	0.
↪518924 0.0119965				

Example

```

from lazypredict.Supervised import LazyRegressor
from sklearn import datasets
from sklearn.utils import shuffle
import numpy as np
boston = datasets.load_boston()
X, y = shuffle(boston.data, boston.target, random_state=13)
X = X.astype(np.float32)
offset = int(X.shape[0] * 0.9)
X_train, y_train = X[:offset], y[:offset]
X_test, y_test = X[offset:], y[offset:]
reg = LazyRegressor(verbose=0, ignore_warnings=False, custom_metric=None )
models, predictions = reg.fit(X_train, X_test, y_train, y_test)

```

Model	R-Squared	RMSE	Time Taken
SVR	0.877199	2.62054	0.0330021
RandomForestRegressor	0.874429	2.64993	0.0659981
ExtraTreesRegressor	0.867566	2.72138	0.0570002
AdaBoostRegressor	0.865851	2.73895	0.144999
NuSVR	0.863712	2.7607	0.0340044
GradientBoostingRegressor	0.858693	2.81107	0.13
KNeighborsRegressor	0.826307	3.1166	0.0179954
HistGradientBoostingRegressor	0.810479	3.25551	0.820995
BaggingRegressor	0.800056	3.34383	0.0579946
MLPRegressor	0.750536	3.73503	0.725997
HuberRegressor	0.736973	3.83522	0.0370018
LinearSVR	0.71914	3.9631	0.0179989
RidgeCV	0.718402	3.9683	0.018003
BayesianRidge	0.718102	3.97041	0.0159984
Ridge	0.71765	3.9736	0.0149941
LinearRegression	0.71753	3.97444	0.0190051

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TransformedTargetRegressor	0.71753	3.97444	0.012001	
LassoCV	0.717337	3.9758	0.0960066	
ElasticNetCV	0.717104	3.97744	0.0860076	
LassoLarsCV	0.717045	3.97786	0.0490005	
LassoLarsIC	0.716636	3.98073	0.0210001	
LarsCV	0.715031	3.99199	0.0450008	
Lars	0.715031	3.99199	0.0269964	
SGDRegressor	0.714362	3.99667	0.0210009	
RANSACRegressor	0.707849	4.04198	0.111998	
ElasticNet	0.690408	4.16088	0.0190012	
Lasso	0.662141	4.34668	0.0180018	
OrthogonalMatchingPursuitCV	0.591632	4.77877	0.0180008	
ExtraTreeRegressor	0.583314	4.82719	0.0129974	
PassiveAggressiveRegressor	0.556668	4.97914	0.0150032	
GaussianProcessRegressor	0.428298	5.65425	0.0580051	
OrthogonalMatchingPursuit	0.379295	5.89159	0.0180039	
DecisionTreeRegressor	0.318767	6.17217	0.0230272	
DummyRegressor	-0.0215752	7.55832	0.0140116	
LassoLars	-0.0215752	7.55832	0.0180008	
KernelRidge	-8.24669	22.7396	0.0309792	

Warning: Regression and Classification are replaced with LazyRegressor and LazyClassifier. Regression and Classification classes will be removed in next release

Contributions are welcome, and they are greatly appreciated! Every little bit helps, and credit will always be given. You can contribute in many ways:

10.1 Types of Contributions

10.1.1 Report Bugs

Report bugs at <https://github.com/shankarpandala/lazypredict/issues>.

If you are reporting a bug, please include:

- Your operating system name and version.
- Any details about your local setup that might be helpful in troubleshooting.
- Detailed steps to reproduce the bug.

10.1.2 Fix Bugs

Look through the GitHub issues for bugs. Anything tagged with “bug” and “help wanted” is open to whoever wants to implement it.

10.1.3 Implement Features

Look through the GitHub issues for features. Anything tagged with “enhancement” and “help wanted” is open to whoever wants to implement it.

10.1.4 Write Documentation

Lazy Predict could always use more documentation, whether as part of the official Lazy Predict docs, in docstrings, or even on the web in blog posts, articles, and such.

10.1.5 Submit Feedback

The best way to send feedback is to file an issue at <https://github.com/shankarpandala/lazypredict/issues>.

If you are proposing a feature:

- Explain in detail how it would work.
- Keep the scope as narrow as possible, to make it easier to implement.
- Remember that this is a volunteer-driven project, and that contributions are welcome :)

10.2 Get Started!

Ready to contribute? Here's how to set up *lazypredict* for local development.

1. Fork the *lazypredict* repo on GitHub.
2. Clone your fork locally:

```
$ git clone git@github.com:your_name_here/lazypredict.git
```

3. Install your local copy into a virtualenv. Assuming you have virtualenvwrapper installed, this is how you set up your fork for local development:

```
$ mkvirtualenv lazypredict
$ cd lazypredict/
$ python setup.py develop
$ pip install -r requirements_dev.txt
```

4. Create a branch for local development:

```
$ git checkout -b name-of-your-bugfix-or-feature
```

Now you can make your changes locally.

5. When you're done making changes, check that your changes pass flake8 and the tests, including testing other Python versions with tox:

```
$ flake8 lazypredict tests
$ python setup.py test or pytest
$ tox
```

To get flake8 and tox, just pip install them into your virtualenv.

6. Commit your changes and push your branch to GitHub:

```
$ git add .
$ git commit -m "Your detailed description of your changes."
$ git push origin name-of-your-bugfix-or-feature
```

7. Submit a pull request through the GitHub website.

10.3 Pull Request Guidelines

Before you submit a pull request, check that it meets these guidelines:

1. The pull request should include tests.
2. If the pull request adds functionality, the docs should be updated. Put your new functionality into a function with a docstring, and add the feature to the list in README.rst.
3. The pull request should work for Python 2.7, 3.5, 3.6, 3.7 and 3.8, and for PyPy. Check https://travis-ci.org/shankarpandala/lazypredict/pull_requests and make sure that the tests pass for all supported Python versions.

10.4 Tips

To run a subset of tests:

```
$ pytest tests.test_lazypredict
```

10.5 Deploying

A reminder for the maintainers on how to deploy. Make sure all your changes are committed (including an entry in HISTORY.rst). Then run:

```
$ bump2version patch # possible: major / minor / patch
$ git push
$ git push --tags
```

Travis will then deploy to PyPI if tests pass.

11.1 Development Lead

- Shankar Rao Pandala <shankar.pandala@live.com>

11.2 Contributors

- Breno Batista da Silva <brenophp@gmail.com>

12.1 0.2.8 (2021-02-06)

- Removed StackingRegressor and CheckingClassifier.
- Added provided_models method.
- Added adjusted r-squared metric.
- Added cardinality check to split categorical columns into low and high cardinality features.
- Added different transformation pipeline for low and high cardinality features.
- Included all number dtypes as inputs.
- Fixed dependencies.
- Improved documentation.

12.2 0.2.7 (2020-07-09)

- Removed catboost regressor and classifier

12.3 0.2.6 (2020-01-22)

- Added xgboost, lightgbm, catboost regressors and classifiers

12.4 0.2.5 (2020-01-20)

- Removed troublesome regressors from list of CLASSIFIERS

12.5 0.2.4 (2020-01-19)

- Removed troublesome regressors from list of REGRESSORS
- Added feature to input custom metric for evaluation
- Added feature to return predictions as dataframe
- Added model training time for each model

12.6 0.2.3 (2019-11-22)

- Removed TheilSenRegressor from list of REGRESSORS
- Removed GaussianProcessClassifier from list of CLASSIFIERS

12.7 0.2.2 (2019-11-18)

- Fixed automatic deployment issue.

12.8 0.2.1 (2019-11-18)

- Release of Regression feature.

12.9 0.2.0 (2019-11-17)

- Release of Classification feature.

12.10 0.1.0 (2019-11-16)

- First release on PyPI.

CHAPTER 13

Indices and tables

- `genindex`
- `modindex`
- `search`