
XLearning

Release 0.0.2

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ABOUT XLEARNING

A small library of machine learning models and utility & plotting functions:

1. a set of utility functions, e.g., wrap function for cross-validation on regression and classification models
2. a set of small models, e.g., mixture of linear regression model
3. a set of plotting functions, e.g., *corr_plot*, *ROC_curve*, *PR_curve*

1.1 Installation

1.1.1 Easy install

xlearning is available through [pypi](#). To install, type the following command line, and add -U for upgrading:

```
pip install -U xlearning
```

Alternatively, you can install from this GitHub repository for latest (often development) version by following command line

```
pip install -U git+https://github.com/Rongtingting/XLearning
```

In either case, if you don't have write permission for your current Python environment, add --user, but check the previous section on create your own conda environment.

To remove the package, type the following command line,

```
pip uninstall xlearning
```

1.1.2 Issues

If you have any issue, please report it to the issue on [XLearning github](#).

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        "lassoreg = linear_model.LassoCV(cv=3)n", "randforest = RandomForestRe-  
        gressor(n_estimators=100, n_jobs=-1)n", "n", "fig = plt.figure(figsize=(10,  
        3.5), dpi=80)n", "plt.subplot(1,3,1)n", "# Cross-Validation wrap & corr_plotn",  
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```

```

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“corr_plot(Y, Y_pre, size=20)n”, “n”, “plt.xlabel("observation")n”,
“plt.ylabel("prediction")n”, “plt.title("linear regression")n”, “plt.grid(alpha=0.2)n”,
“# pl.ylim(-6,4)n”, “n”, “plt.subplot(1,3,2)n”, “CV = CrossValidation(X,Y)n”,
“Y_pre = CV.cv_regression(lassoreg)n”, “corr_plot(Y, Y_pre, size=20)n”,
“plt.xlabel("observation")n”, “plt.ylabel("prediction")n”, “plt.title("Lasso regression")n”,
“plt.grid(alpha=0.2)n”, “# pl.ylim(-6,4)n”, “n”, “plt.subplot(1,3,3)n”, “CV = CrossValidation(X,Y)n”,
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“plt.xlabel("observation")n”, “plt.ylabel("prediction")n”,
“plt.title("random forest regression")n”, “plt.grid(alpha=0.2)n”, “# pl.ylim(-6,4)n”, “n”,
“plt.tight_layout(n)”, “# fig.savefig("cv_regression.pdf”, dpi=300,
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```

```
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