

Machine Learning for econometrics

Flexible models for tabular data

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A lot of today's content is taken from the excellent [sklearn mooc](#) (Estève et al., 2022)

Reminder from previous session

- Statistical learning 101: bias-variance trade-off
- Regularization for linear models: Lasso, Ridge, Elastic Net
- Transformation of variables: polynomial regression

Reminder from previous session

- Statistical learning 101: bias-variance trade-off
 - Regularization for linear models: Lasso, Ridge, Elastic Net
 - Transformation of variables: polynomial regression
-  But... How to select the best model? the best hyper-parameters?

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Model evaluation and selection
with cross-validation

A closer look at model evaluation: Wage example

Example with the Wage dataset

- Raw dataset: (N=534, p=11)

EDUCATION	SOUTH	SEX	EXPERIENCE	UNION	WAGE	AGE	RACE	OCCUPATION	SECTOR	MARR
8	no	female	21	not_member	5.10	35	Hispanic	Other	Manufacturing	Married
9	no	female	42	not_member	4.95	57	White	Other	Manufacturing	Married
12	no	male	1	not_member	6.67	19	White	Other	Manufacturing	Unmarried
12	no	male	4	not_member	4.00	22	White	Other	Other	Unmarried
12	no	male	17	not_member	7.50	35	White	Other	Other	Married

A closer look at model evaluation: Wage example

Example with the Wage dataset

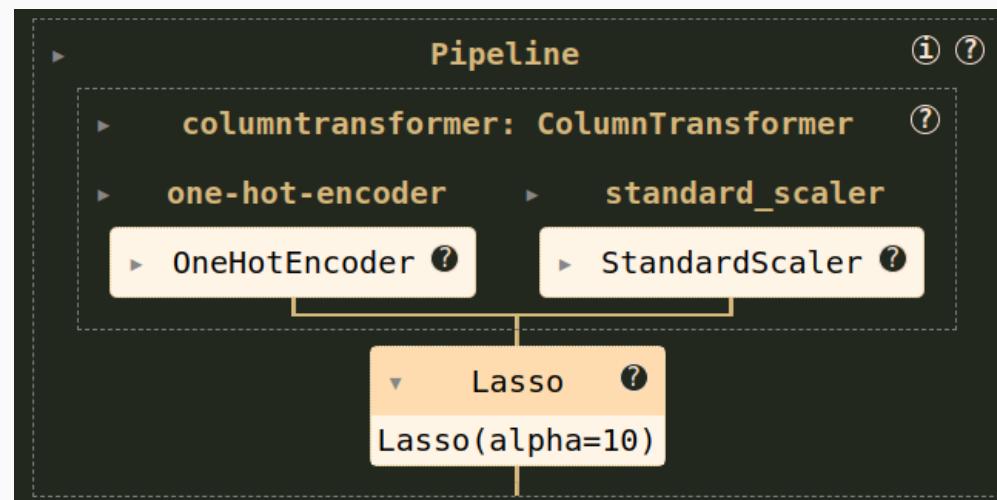
- Raw dataset: (N=534, p=11)
- Transformation: encoding categorical data, scaling numerical data: (N=534, p=23)

one-hot-encoder_SOUTH_no	one-hot-encoder_SOUTH_yes	one-hot-encoder_SEX_female	one-hot-encoder_SEX_male	one-hot-encoder_UNION_member	one-hot-encoder_UNION_not
1.0	0.0	1.0	0.0	0.0	
1.0	0.0	1.0	0.0	0.0	
1.0	0.0	0.0	1.0	0.0	
1.0	0.0	0.0	1.0	0.0	
1.0	0.0	0.0	1.0	0.0	

A closer look at model evaluation: Wage example

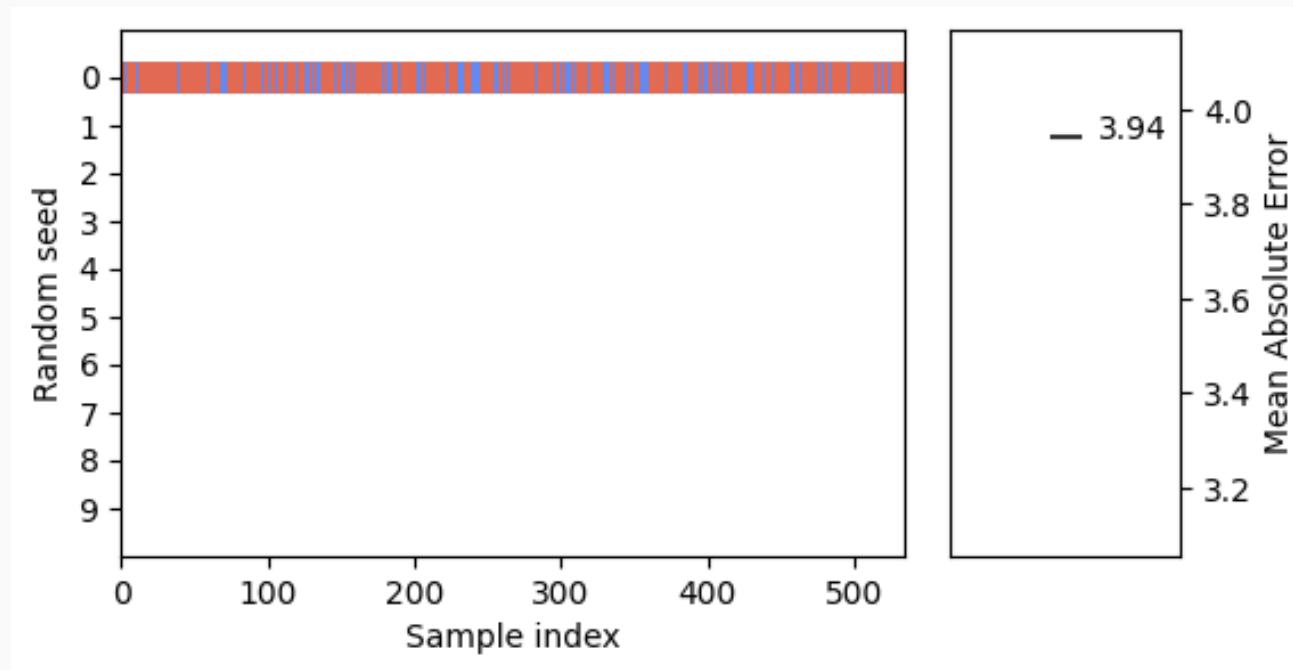
Example with the Wage dataset

- Raw dataset: ($N=534$, $p=11$)
- Transformation: encoding categorical data, scaling numerical data: ($N=534$, $p=23$)
- Regressor: Lasso with regularization parameter ($\alpha = 10$)



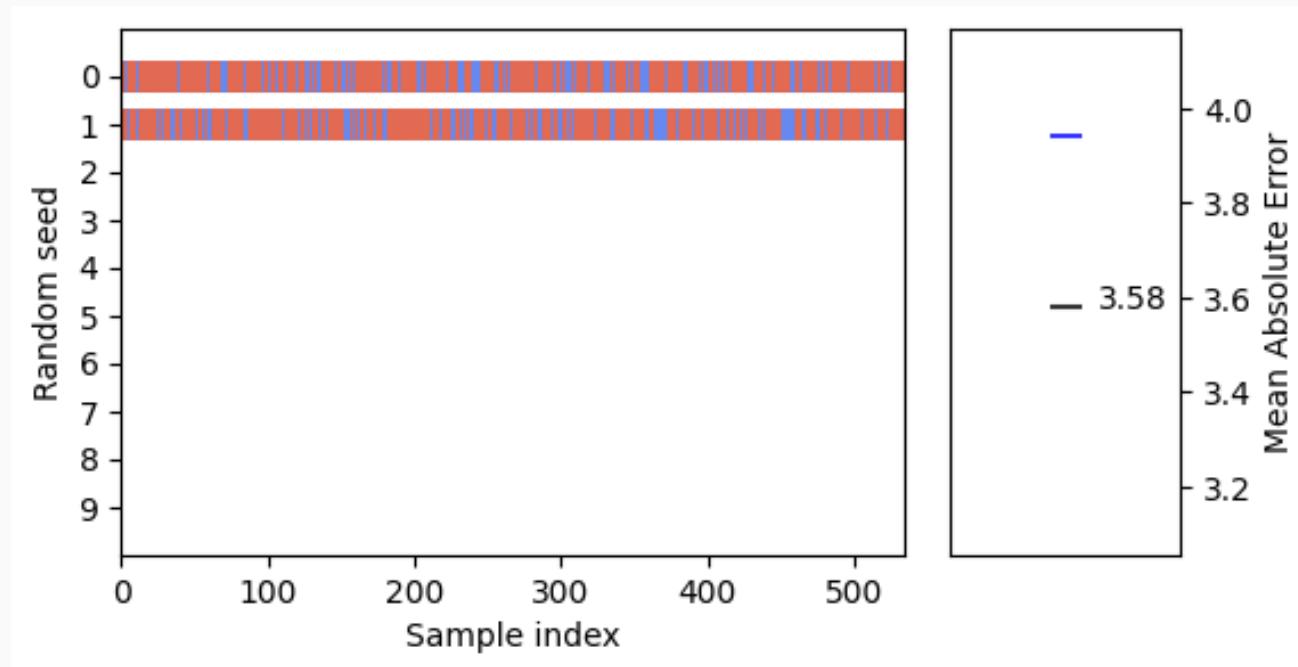
Repeated train/test splits

Splitting once: In red, the training set, in blue, the test set



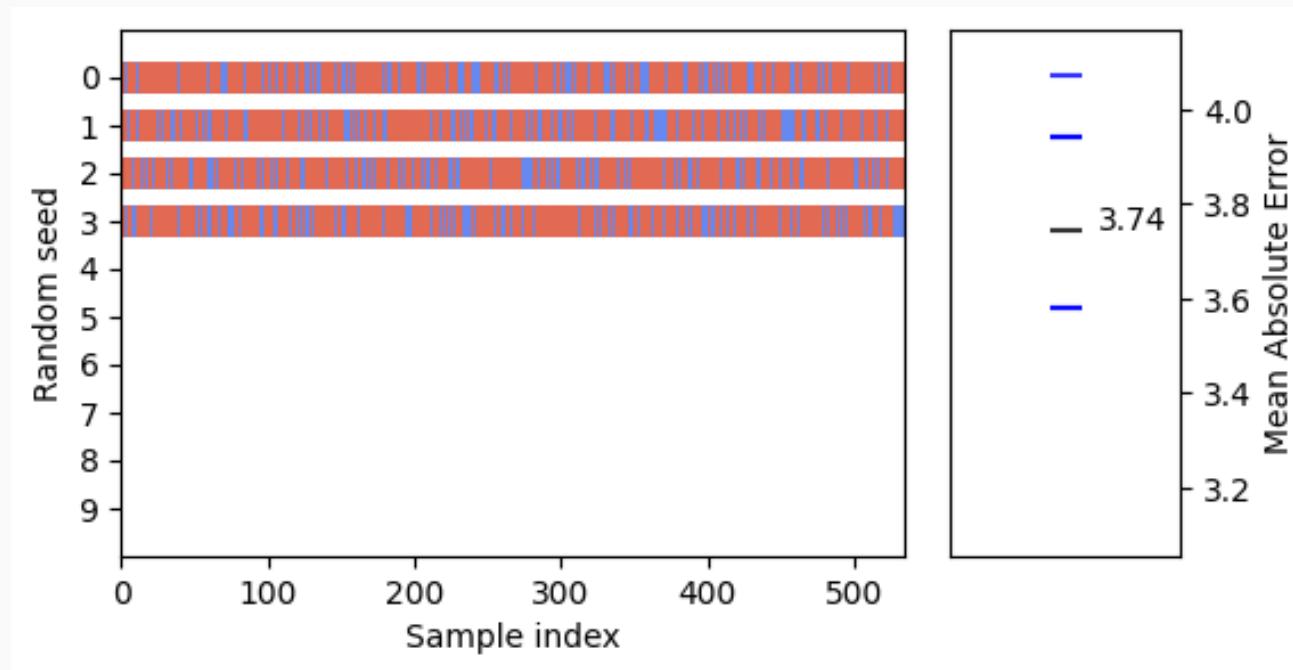
Repeated train/test splits

But we could have chosen another split ! Yielding a different MAE



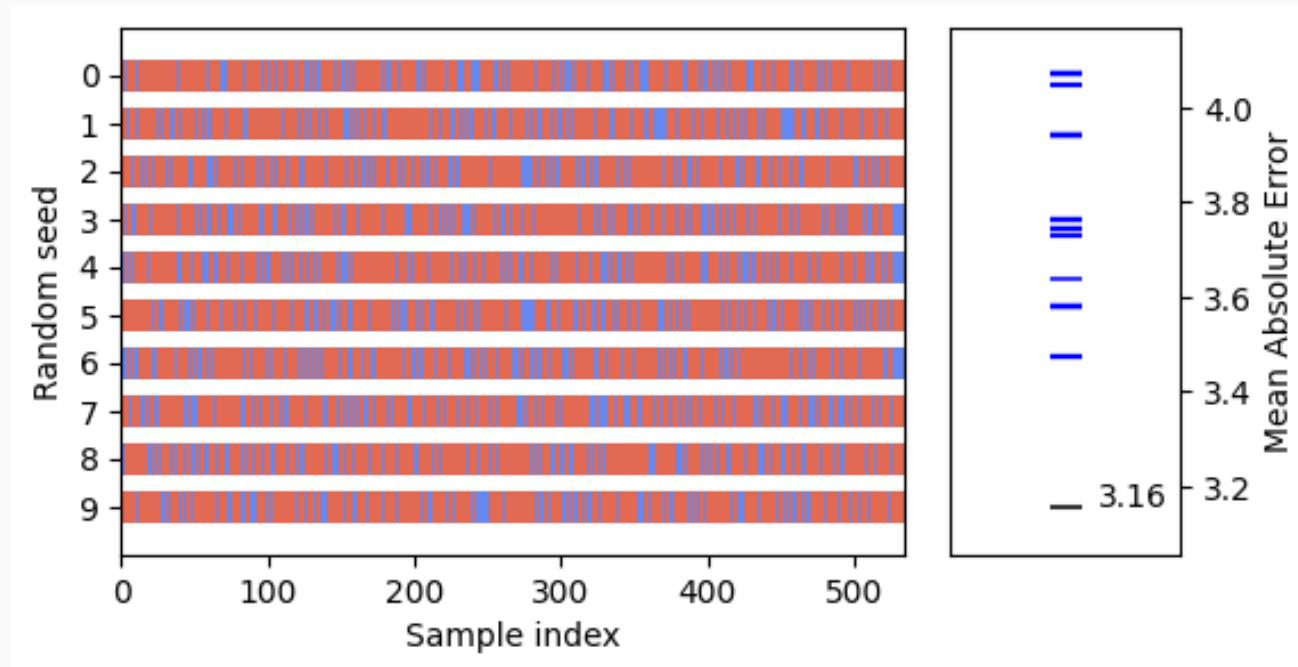
Repeated train/test splits

And another split...



Repeated train/test splits

Splitting ten times



Distribution of MAE: 3.71 ± 0.26

Repeated exclusive train/test splits = Cross-validation

Practical usage with sklearn: `cross_validate`.

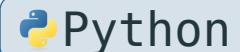
```
1 from sklearn.model_selection import cross_validate
2 cv_results = cross_validate(
3     regressor, data, target, cv=5,
4     scoring="neg_mean_absolute_error")
```



Repeated exclusive train/test splits = Cross-validation

Practical usage with sklearn: `cross_validate`.

```
1 from sklearn.model_selection import cross_validate
2 cv_results = cross_validate(
3     regressor, data, target, cv=5,
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```



- 😊 Robustly estimate generalization performance.
- 😎 Estimate data variability of the performance : bigger source of variation (Bouthillier et al., 2021).
- 🚀 Let's use it to select the best models among several candidates!

Contents

Python hands-on

To your notebooks !



- url: <https://github.com/strayMat/causal-ml-course/tree/main/notebooks>

Bibliography

Bibliography

Bouthillier, X., Delaunay, P., Bronzi, M., Trofimov, A., Nichyporuk, B., Szeto, J., Mohammadi Sepahvand, N., Raff, E., Madan, K., Voleti, V., & others. (2021). Accounting for variance in machine learning benchmarks. *Proceedings of Machine Learning and Systems*, 3, 747–769.

Estève, L., Lemaitre, G., Grisel, O., Varoquaux, G., Amor, A., Lilian, Rospars, B., Schmitt, T., Liu, L., Kinoshita, B. P., *hackmd-deploy, ph4ge, Steinbach, P., Boucaud, A., Muite, B., Boisberranger, J. du, Notter, M., Pierre, P, S., ... parmentelat*. (2022). INRIA/scikit-learn-mooc: Third MOOC session. Zenodo. <https://doi.org/10.5281/zenodo.7220307>