Data-driven Lambda for the LASSO

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In this report, only the significant lags chosen by AR were included in the X matrix.

Since the plots for yr2, yr3, yr4 and yr5 are similar, so here I just plotted yr2.

The M.S.E of the training dataset increases with λ , so it is not feasible to choose a λ according to the performance in training dataset.

Two solutions:

- 1, Same as before, separate the data set into training set (80%) and test set (20%); then choose λ according to the smallest M.S.E in the test set directly and use it for the comparison.
- 2, Separate the data set into three sets, training (60%), validation (20%) and test (20%); then choose λ according to the smallest M.S.E in the validation set, use the M.S.E from the test set for comparison.

(Continued from 12th Dec):

- 1, When we have high-dimentional data, and set singular.ok = TRUE, then run OLS with lm function in R, the x matrix will be truncated to have an appropriate number of columns. Example: n=100, p=200, the $100^{th}-200^{th}$ columns will be removed before running the regression. The actual data used in the regression will be n=100, p=99.
- 2, When we have high-dimentional data, and set λ as a series of values including zero, then run LASSO with glmnet function in R, it gives non-zero estimations to every variables (when $\lambda = 0$). In fact, the function does not allow $\lambda = 0$ so it uses some very small value instead. (verified by data experiments as shown in the scatter plot, when the smallest lambda is smaller than 1E-6, the scatter plot almost reduce to a straight line, code check is ongoing)
- 3, When the data is not high-dimentional, the estimations of LASSO (when $\lambda = 0$) becomes closer and closer to the OLS estimations with decreasing "thresh" values.

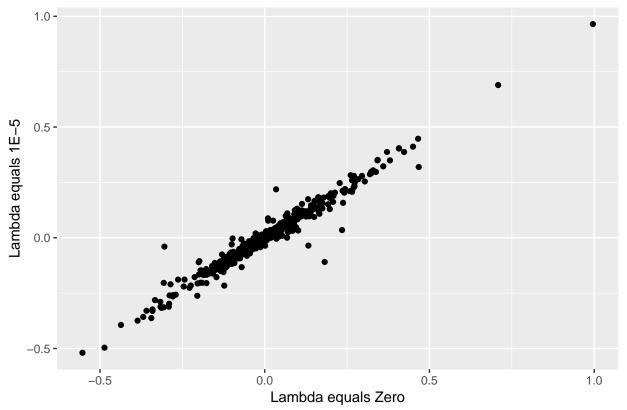
some references:

How Correlations Influence Lasso Prediction

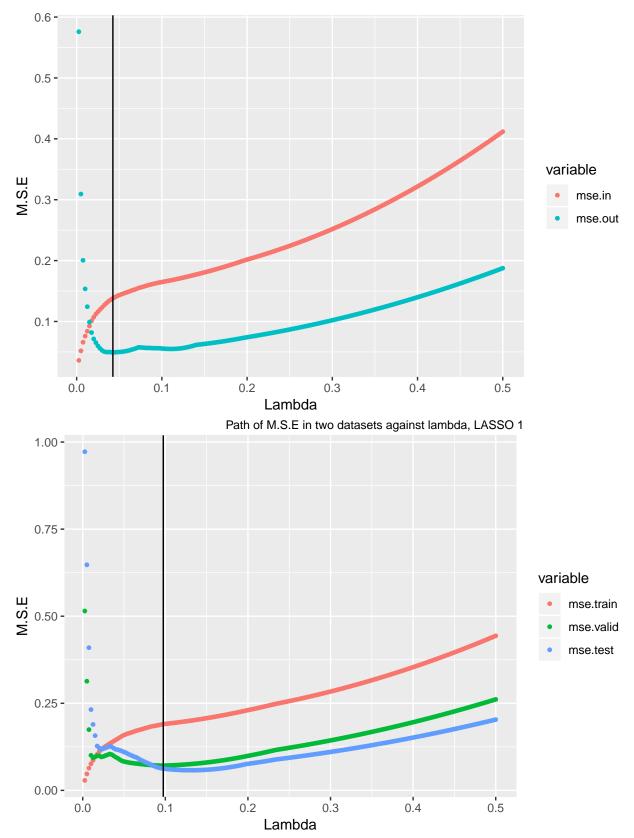
This paper argues correlation is not problematic for LASSO prediction, but it has influence on suitable λ . The higher the correlation, the smaller the tuning parameter.

• LASSO 1

The explanatory variables in this model are level I(1) variables, level I(0) variables, first-differenced I(2) variables and lags of the dependent variables.



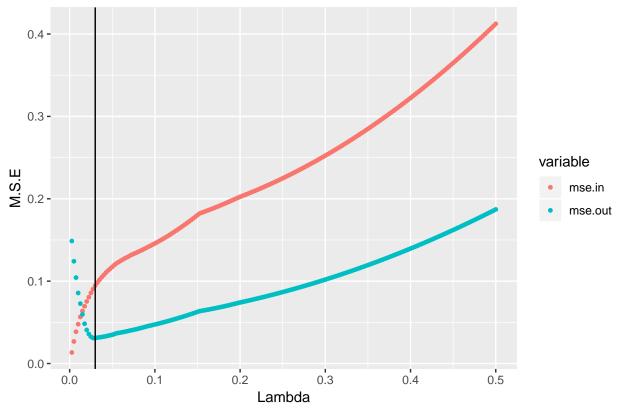
Scatter plot of two sets of coefficients of LASSO when lambda equals to zero and 1E-5



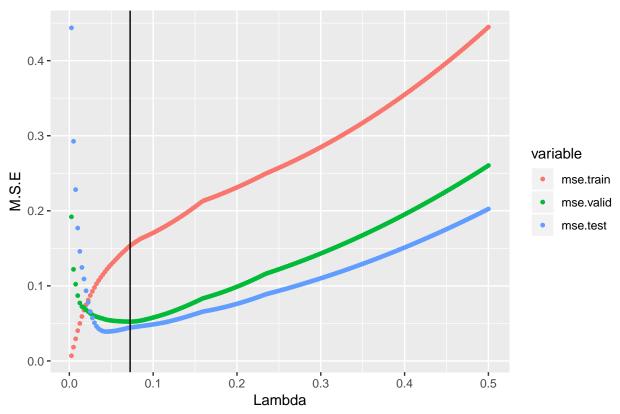
Path of M.S.E in three datasets against lambda, LASSO 1

• LASSO 2

The explanatory variables in this model are first-differenced I(1) variables, level I(0) variables, twice-differenced I(2) variables and lags of the dependent variables.



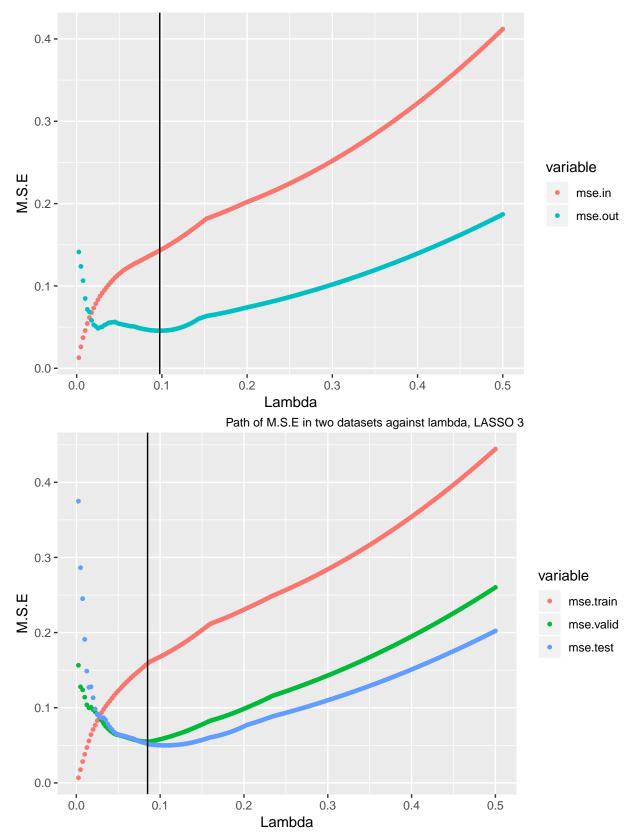
Path of M.S.E in two datasets against lambda, LASSO 2



Path of M.S.E in three datasets against lambda, LASSO 2

• LASSO 3

The explanatory variables in this model are first-differenced I(1) variables, level I(0) variables, twice-differenced I(2) variables, level I(1) variables, first-differenced I(2) variables and lags of the dependent variables.

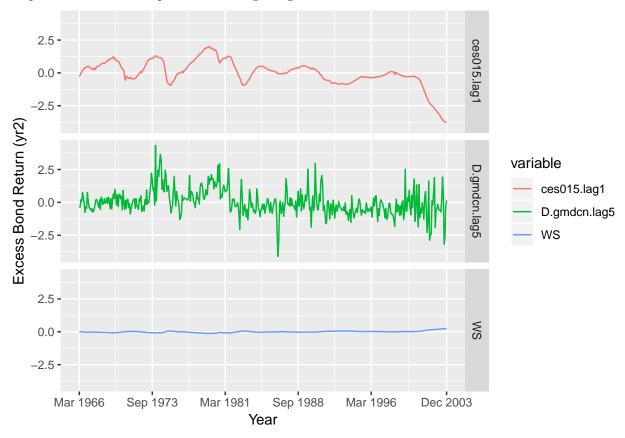


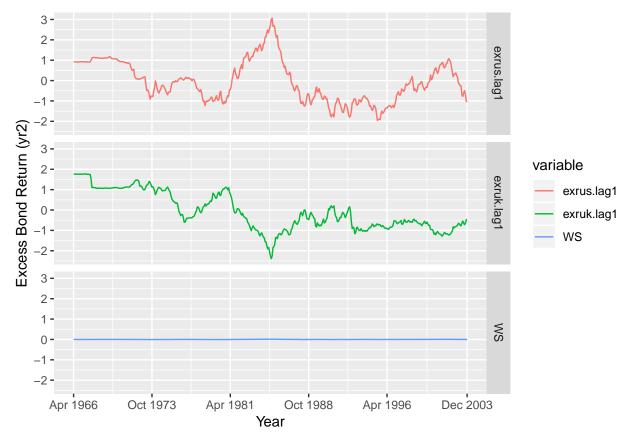
Path of M.S.E in three datasets against lambda, LASSO 3

Table 1: Variables selected by the LASSO with data driven lambda (valid) for yr2

variable	LASSO 1	LASSO 2	LASSO 3
ces 015.lag 1	-0.06302	NA	-0.081
pmcp.lag1	-0.08144	-0.07776	-0.06601
D.gmdcn.lag5	-3e-06	NA	NA
fclbmc.lag12	0.00039	NA	NA
ytem.lag1	0.7669	0.8223	0.795
ytem.lag5	NA	0.009655	NA
D.ypr.lag1	NA	-0.003463	-0.000973
D.a0m051.lag1	NA	-0.01647	-0.004059
pmcp.lag5	NA	-0.01831	-0.001411
D.fygt10.lag5	NA	-0.003416	-0.001061
D.fygt1.lag12	NA	0.001242	1.4e-05
D.fygt5.lag12	NA	0.04613	0.04731
D.fygt10.lag12	NA	0.04782	0.03405
D.fyff.lag13	NA	-0.004052	NA
exrus.lag1	NA	NA	0.002564
exruk.lag1	NA	NA	-0.002343

Graphs of the two sets of potential co-integrating vectors.





Model Confidence Set

```
##
## Model L1_valid eliminated 2019-01-25 12:28:24
## Model L1_valid_all eliminated 2019-01-25 12:28:37
## Superior Set Model created
                          :
##
             Rank_M
                        v_M MCS_M Rank_R
                                             v_R MCS_R
                 9 1.1290306 0.7820
                                        3.2915306 0.0136 0.04929423
## L1_test
## L2_test
                 2 -3.5395993 1.0000
                                      1 -0.6686474 1.0000 0.03080323
                                        3.1910667 0.0174 0.04444512
## L2_valid
                 5 -0.1096763 1.0000
## L3_test
                   0.3792867 0.9980
                                        3.0336109 0.0290 0.04567235
                 7
                                      6 2.7912930 0.0594 0.05180697
## L3_valid
                10
                   1.3434476 0.6248
                12 1.9470483 0.2336
## L1_test_all
                                     12
                                        3.6343405 0.0050 0.04960958
## L2_test_all
                 1 -3.8169600 1.0000
                                      2
                                        0.6686474 0.9976 0.03209796
                 8 0.6866800 0.9654
                                     10 3.2292734 0.0156 0.04695511
## L2_valid_all
## L3_test_all
                 6 0.2819309 0.9996
                                     7
                                        2.9539224 0.0594 0.04548748
                                      5 2.7775475 0.3302 0.05350068
## L3 valid all
                11 1.4615826 0.5398
## AR
                 4 -0.2500794 1.0000
                                      3 1.9895810 0.4032 0.04356516
## AR all
                 3 -0.2521589 1.0000
                                      4 2.1099687 0.3302 0.04361781
## p-value :
  [1] 0.2336
##
  ##
##
```

Superior Set of Models

Elapsed Time : Time difference of 40.46658 secs