

# Reports of MSE

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## Stock and Watson Dataset (GDP growth)

### Data

Data:  $145+1-5=141$  series in total.

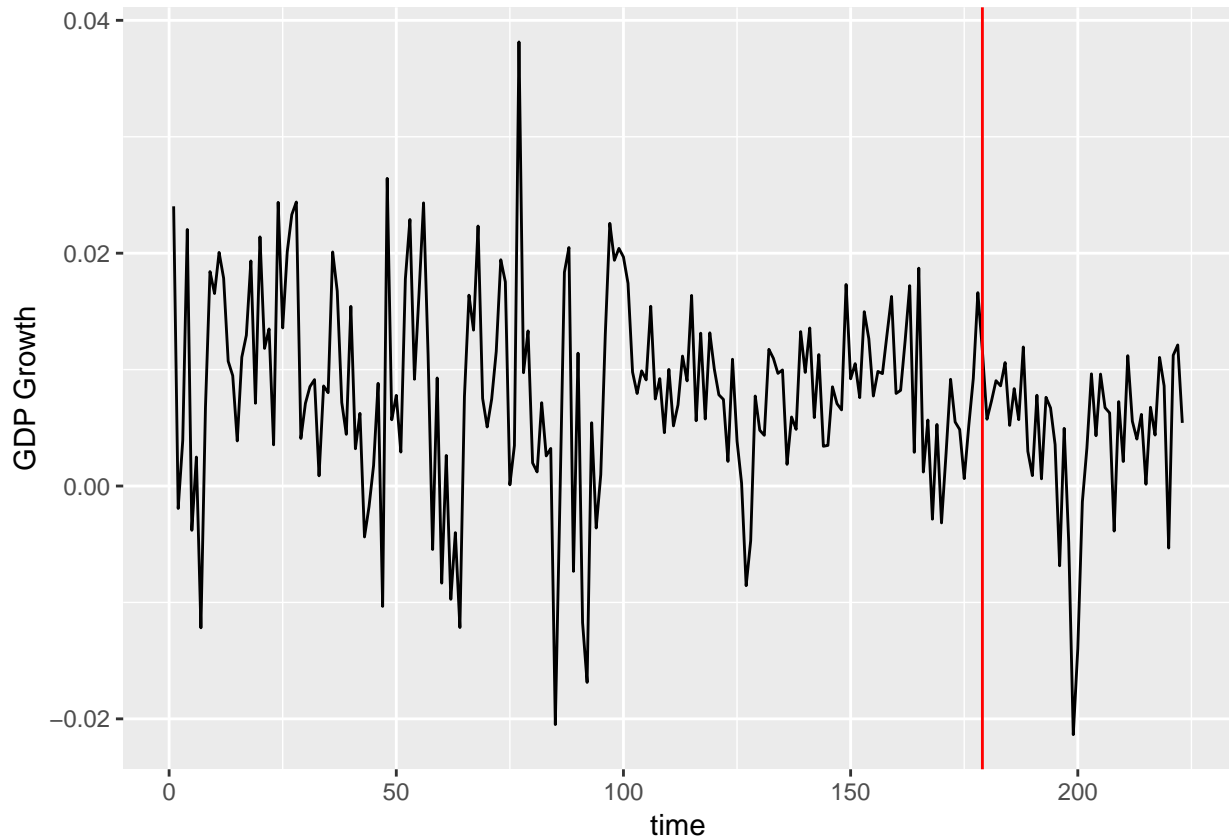
The “spread” series (difference between two I(1) series) are removed.

One of the I(1), “CP3FM” (short name: “Com Paper”) was omitted in the original data set, now is added.

$\log()$  is done.

$y$  is GDP.

The earliest 80% of data were used to estimate the coefficients, the rest 20% were used to calculate the out-of-sample MSE.



### ADF test

Step 1, ADF test to the 146 original series.

Step 2, mark “I(0)” variables as “I(0)”.

Step 3, ADF test to the first-differenced 146 series.

Step 4, check for contradictions, found “PCED\_RecServices” in AIC.

Step 5, mark “I(1)” variable as “I(2)” (including “PCED\_RecServices”).

Step 6, mark the rest as “I(1)”.

Step 7, repeat the above 6 steps for both “AIC” and “BIC”.

## AR(p) model

- The estimation method is OLS; lags are chosen by AIC.

### Lasso 1

$$\begin{aligned}\Delta y_t = & y_{t-1} \\ & + \Delta y_{t-1} + \Delta y_{t-2} + \Delta y_{t-3} + \Delta y_{t-4} \\ & + I(0)_{t-1} + I(0)_{t-2} + I(0)_{t-3} + I(0)_{t-4} \\ & + I(1)_{t-1} + I(1)_{t-2} + I(1)_{t-3} + I(1)_{t-4} \\ & + \Delta I(2)_{t-1} + \Delta I(2)_{t-2} + \Delta I(2)_{t-3} + \Delta I(2)_{t-4}\end{aligned}$$

### Lasso 2

$$\begin{aligned}\Delta y_t = & y_{t-1} \\ & + \Delta y_{t-1} + \Delta y_{t-2} + \Delta y_{t-3} + \Delta y_{t-4} \\ & + I(0)_{t-1} + I(0)_{t-2} + I(0)_{t-3} + I(0)_{t-4} \\ & + \Delta I(1)_{t-1} + \Delta I(1)_{t-2} + \Delta I(1)_{t-3} + \Delta I(1)_{t-4} \\ & + \Delta^2 I(2)_{t-1} + \Delta^2 I(2)_{t-2} + \Delta^2 I(2)_{t-3} + \Delta^2 I(2)_{t-4}\end{aligned}$$

### Lasso 3

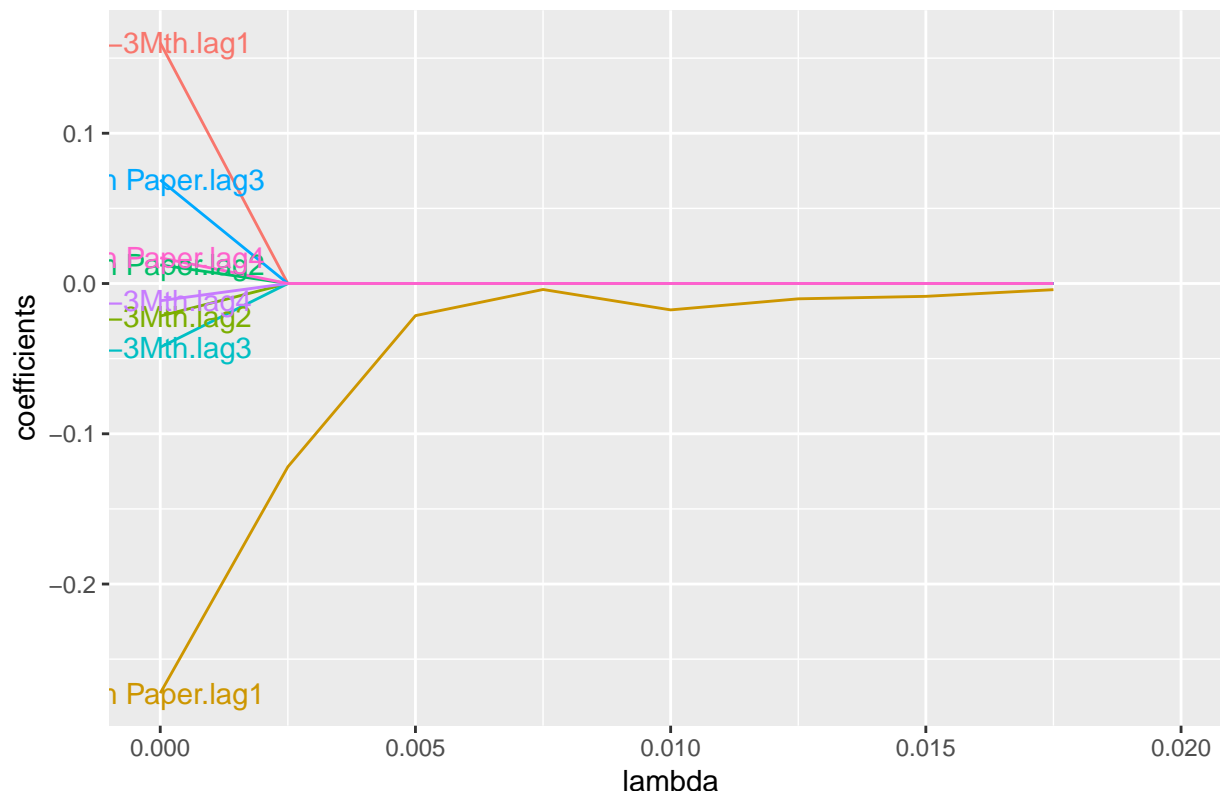
$$\begin{aligned}\Delta y_t = & y_{t-1} \\ & + \Delta y_{t-1} + \Delta y_{t-2} + \Delta y_{t-3} + \Delta y_{t-4} \\ & + I(0)_{t-1} + I(0)_{t-2} + I(0)_{t-3} + I(0)_{t-4} \\ & + \Delta I(1)_{t-1} + \Delta I(1)_{t-2} + \Delta I(1)_{t-3} + \Delta I(1)_{t-4} \\ & + \Delta^2 I(2)_{t-1} + \Delta^2 I(2)_{t-2} + \Delta^2 I(2)_{t-3} + \Delta^2 I(2)_{t-4} \\ & + I(1)_{t-1} + \Delta I(2)_{t-1}\end{aligned}$$

Table 1: out-of-sample MSE, when the dependent variable is inflation

	SW	AIC	BIC
<b>LASSO 1</b>	0.7607	0.726	0.7059
<b>LASSO 2</b>	0.4971	0.4971	0.4993
<b>LASSO 3</b>	0.4981	0.4981	0.5002

- AIC chooses 2 lags for the GDP growth.
- The out-of-sample MSE of AR(2) is 0.89.
- The estimated AR(1) coefficient is 0.3078.

### Trace plot of two interest rates



The trace plot shows the paths of the four lags of the two interest rates that are  $I(1)$  variables: “3-month treasury bill (TB-3Mth)” and “3-Month AA Financial Commercial Paper Rate post 1997 ... linked to XLI CP90 before 1997 (Com Paper)”.

Previously, when we had the difference of these two variables “CP\_Tbill Spread” in the data, it was selected by the LASSO and the absolute value of its coefficient was the biggest among other non-zero coefficients. Since we have removed all the spreads in the data, I try to figure out what is happening to the two “parents”  $I(1)$  variables.

From the graph, we see:

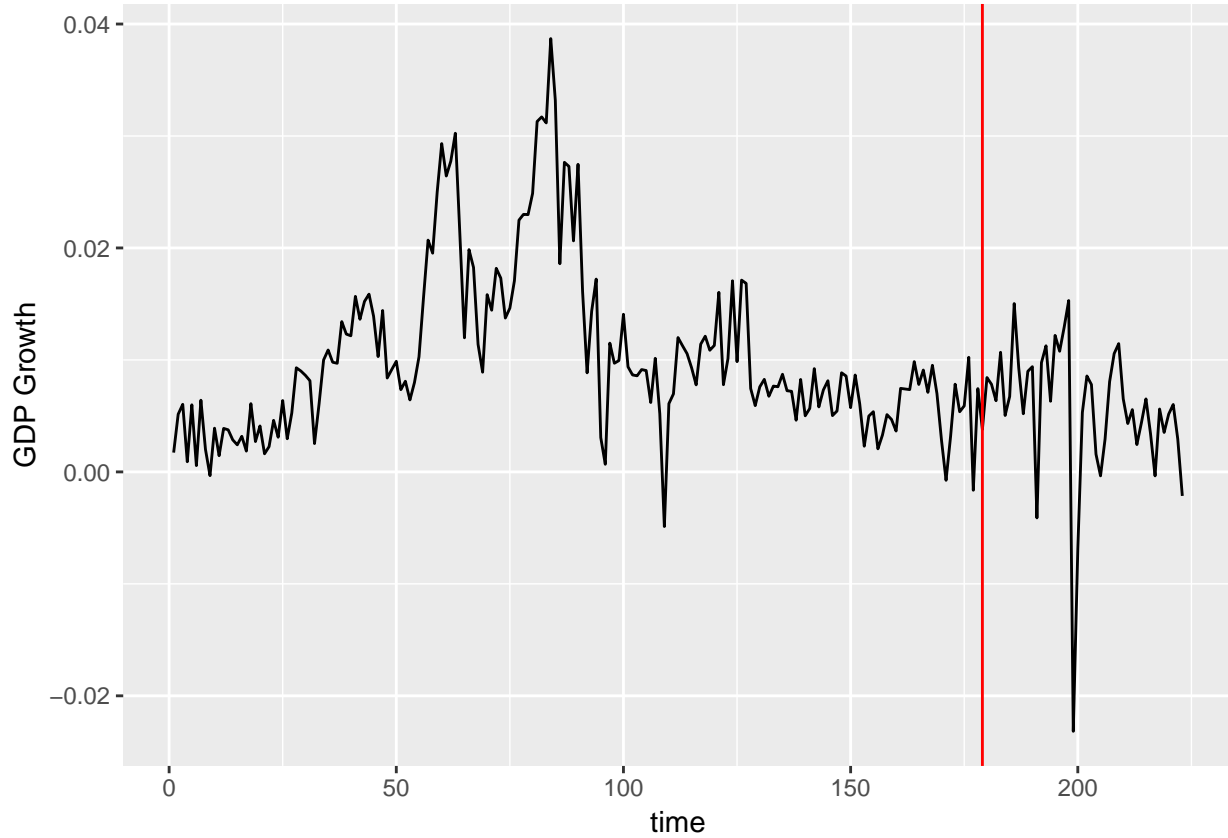
1, when  $\lambda = 0$ , which is the OLS, the magnitude and signs of the estimated coefficients all make sense, for example, lag-1 of TB-3Mth is positive and has a big coefficient, while lag-1 of Com Paper is negative and also has a big absolute coefficient. I think this may indicate that OLS works OK in terms of estimating co-integrating vectors.

2, Once  $\lambda$  becomes non-zero, (in our case, the smallest non-zero  $\lambda = 0$  is 0.0025) seven out of eight of these variables are dropped immediately, which is a bit disappointing.

Note: Although this graph only shows the trace of eight variables (two series with four lags), the regression was conducted with all 146 variables.

## Stock and Watson Dataset (Inflation)

- Stock and Watson twice differenced the CPI to make it stationary.
- ADF test with AIC/BIC conclude Inflation as  $I(0)$  with full set of observations.
- ADF test with AIC/BIC conclude Inflation as  $I(1)$  with the first 80% of observations.



### Inflation as $I(0)$

Table 2: out-of-sample MSE, when the dependent variable is inflation

	SW	AIC	BIC
<b>LASSO 1</b>	0.7649	0.7189	0.732
<b>LASSO 2</b>	1.399	0.8646	0.8219
<b>LASSO 3</b>	0.7413	0.7642	0.7626

- AIC chooses 3 lags for Inflation.
- The out-of-sample MSE of  $AR(3)$  is 2.266.
- The estimated  $AR(1)$  coefficient is 0.7565.

Table 3: ADF test for Inflation

Series	Conclusion	Type	Lags
Inflation	I(0)	trend	2
Inflation	I(0)	trend	2
y.train	I(1)	drift	3
y.train	I(1)	drift	3

### Inflation as I(1)

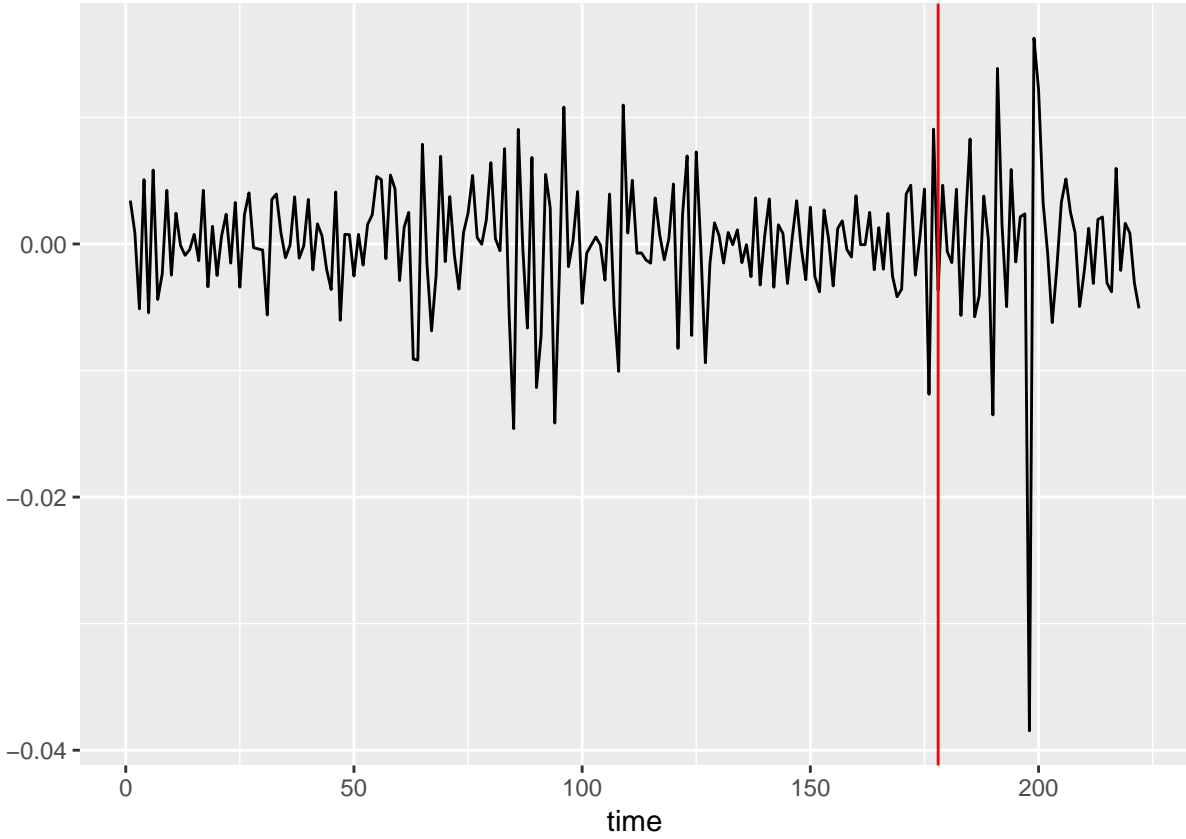
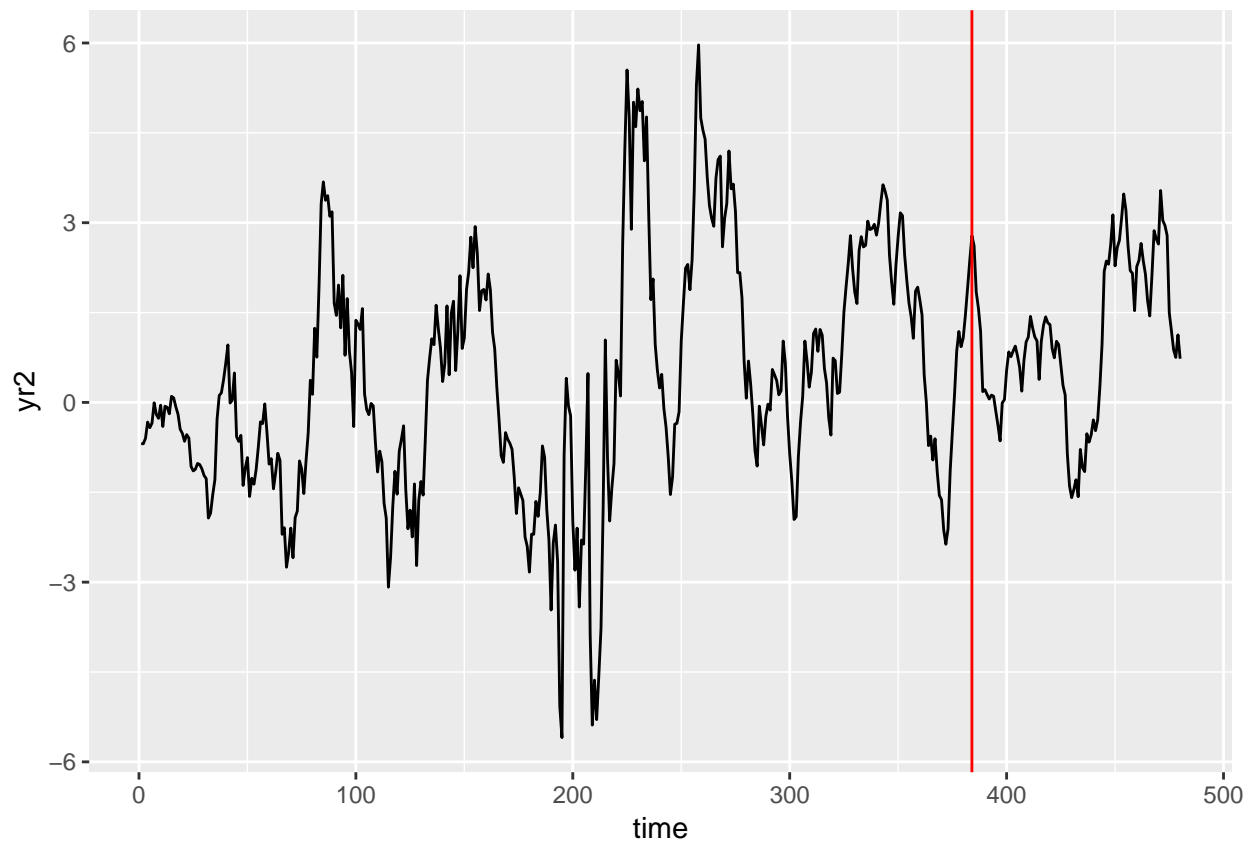


Table 4: out-of-sample MSE, when the dependent variable is first-differenced inflation

	SW	AIC	BIC
<b>LASSO 1</b>	2.091	2.298	2.298
<b>LASSO 2</b>	2.216	2.058	2.069
<b>LASSO 3</b>	2.075	2.064	2.076

- AIC chooses 4 lags for the first-differenced inflation.
- The out-of-sample MSE of AR(4) is 1.969.
- The estimated AR(1) coefficient is -0.2728.



## NG Dataset (bond returns)

Data: 131-8=123 series in total, 480 observations.

The “spread” series (difference between two I(1) series) are removed.

log() is done.

ADF test suggest I(0) for all four bond returns, whether use “trend” or “drift” specification.

### Lasso 1

I(2) is first differenced, others are original.

$$\begin{aligned}
 y_t = & I(0)_{t-1} + I(0)_{t-2} + I(0)_{t-3} + I(0)_{t-4} \\
 & + I(1)_{t-1} + I(1)_{t-2} + I(1)_{t-3} + I(1)_{t-4} \\
 & + \Delta I(2)_{t-1} + \Delta I(2)_{t-2} + \Delta I(2)_{t-3} + \Delta I(2)_{t-4}
 \end{aligned}$$

### Lasso 2

All stationary.

$$\begin{aligned}
y_t = & I(0)_{t-1} + I(0)_{t-2} + I(0)_{t-3} + I(0)_{t-4} \\
& + \Delta I(1)_{t-1} + \Delta I(1)_{t-2} + \Delta I(1)_{t-3} + \Delta I(1)_{t-4} \\
& + \Delta^2 I(2)_{t-1} + \Delta^2 I(2)_{t-2} + \Delta^2 I(2)_{t-3} + \Delta^2 I(2)_{t-4}
\end{aligned}$$

### Lasso 3

Combination of Lasso 1 and 2.

$$\begin{aligned}
y_t = & I(0)_{t-1} + I(0)_{t-2} + I(0)_{t-3} + I(0)_{t-4} \\
& + \Delta I(1)_{t-1} + \Delta I(1)_{t-2} + \Delta I(1)_{t-3} + \Delta I(1)_{t-4} \\
& + \Delta^2 I(2)_{t-1} + \Delta^2 I(2)_{t-2} + \Delta^2 I(2)_{t-3} + \Delta^2 I(2)_{t-4} \\
& + I(1)_{t-1} + \Delta I(2)_{t-1}
\end{aligned}$$

### yr2 as the dependent variable

- AIC chooses 13 lags for yr2.
- The out-of-sample MSE of AR(13) is 0.091.
- The estimated AR(1) coefficient is 0.9308.

Table 5: out-of-sample MSE of LASSO, when the dependent variable is YR2

	NG	AIC	BIC
<b>LASSO 1</b>	0.4879	0.4879	0.4879
<b>LASSO 2</b>	0.4512	0.3884	0.4523
<b>LASSO 3</b>	0.4725	0.4725	0.4725

### yr3 as the dependent variable

- AIC chooses 25 lags for yr3.
- The out-of-sample MSE of AR(25) is 0.077.
- The estimated AR(1) coefficient is 0.934.

Table 6: out-of-sample MSE of LASSO, when the dependent variable is yr3

	NG	AIC	BIC
<b>LASSO 1</b>	0.521	0.5146	0.5141
<b>LASSO 2</b>	0.5726	0.492	0.5599
<b>LASSO 3</b>	0.5068	0.509	0.5109

### yr4 as the dependent variable

- AIC chooses 25 lags for yr4.
- The out-of-sample MSE of AR(25) is 0.0735.
- The estimated AR(1) coefficient is 0.9342.

Table 7: out-of-sample MSE of LASSO, when the dependent variable is yr4

	NG	AIC	BIC
<b>LASSO 1</b>	0.5356	0.5266	0.5226
<b>LASSO 2</b>	0.5811	0.4749	0.5708
<b>LASSO 3</b>	0.5179	0.5223	0.5251

### yr5 as the dependent variable

- AIC chooses 25 lags for yr5.
- The out-of-sample MSE of AR(25) is 0.066.
- The estimated AR(1) coefficient is 0.9251.

Table 8: out-of-sample MSE of LASSO, when the dependent variable is yr5

	NG	AIC	BIC
<b>LASSO 1</b>	0.5724	0.5609	0.5548
<b>LASSO 2</b>	0.6202	0.5122	0.6116
<b>LASSO 3</b>	0.5526	0.5634	0.5663