

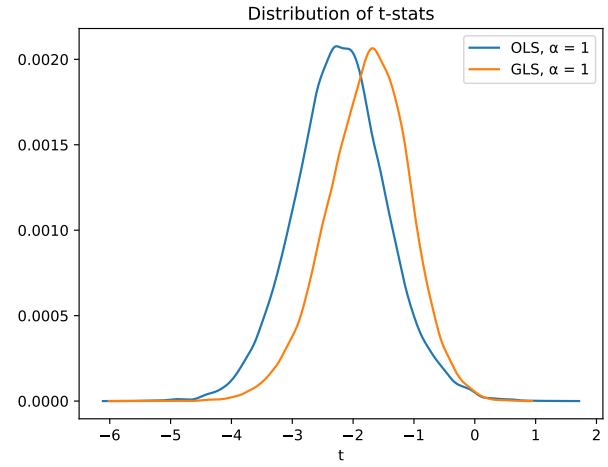
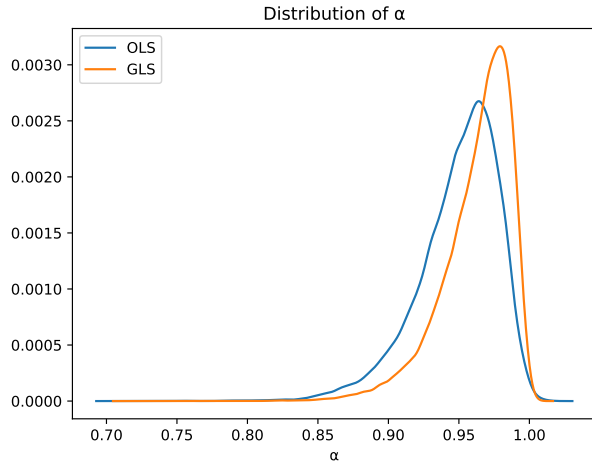
Q1. In the first panel, I plot the distribution of $\hat{\alpha}$ (not sure what the normalization $T(\hat{\alpha} - 1)$ means). In the second panel, I plot the t-stats constructed as follows.

- OLS: Estimate $y_t = a + \beta t + \alpha y_{t-1} + \text{error}_t$, then

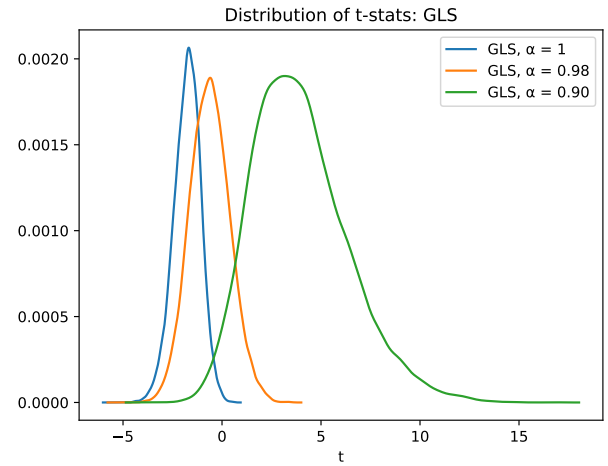
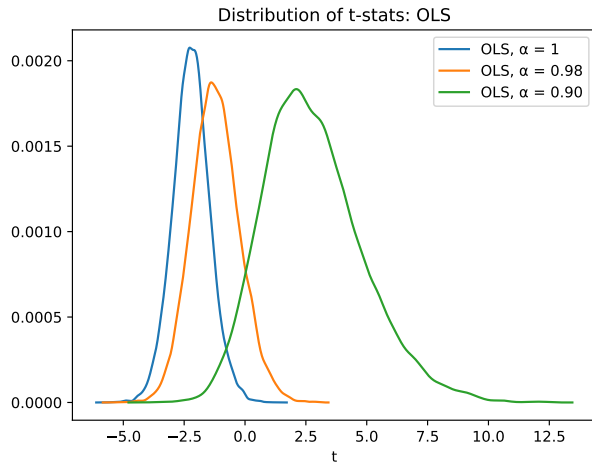
$$t_{\text{OLS}} \equiv \frac{\hat{\alpha} - 1}{\sqrt{\frac{\sum_{t=1}^T y_t^2}{T}}}$$

- GLS: Estimate $y_t^d = \alpha_G y_{t-1}^d + \text{error}_t$, then

$$t_{\text{GLS}} \equiv \frac{\hat{\alpha}_G - 1}{\sqrt{\frac{\sum_{t=1}^T (y_t^d)^2}{T}}}$$



Q2. Here I plot the same t-stats as mentioned in the previous exercise, except I substitute 1 for 0.98 and 0.90.



The critical values under the three nulls are as follows:

α	OLS	GLS
1.00	0.9458	0.707
0.98	0.1488	0.07692
0.90	0.3004	0.6324

Q3. Using the same data as in the previous exercises, the equation I estimate with OLS is

$$y_t = a + b \cdot \mathbb{1}\{t > 75\} + c \cdot (t \times \mathbb{1}\{t > 75\}) + d \cdot t + y_{t-1} + \text{error}_t \quad (1)$$

The critical values under the three nulls are as follows:

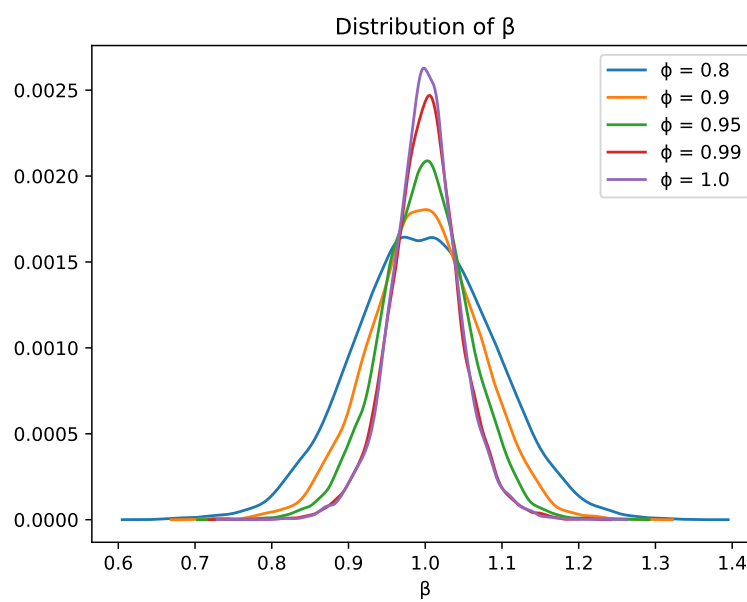
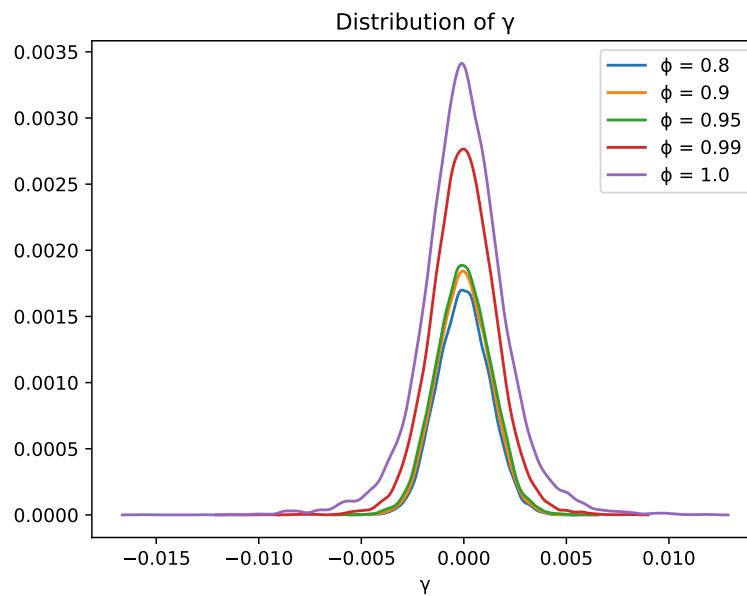
α	OLS
1.00	1.553
0.98	0.6185
0.90	0.1014

Q4. I perform the same estimation as in equation (1), except now the data are generated from a process with a trend shift. The critical values under the three nulls are as follows:

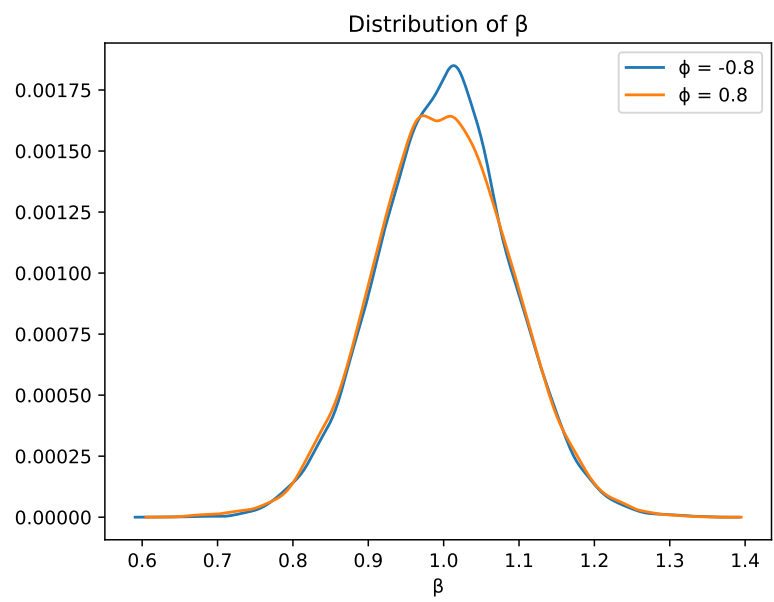
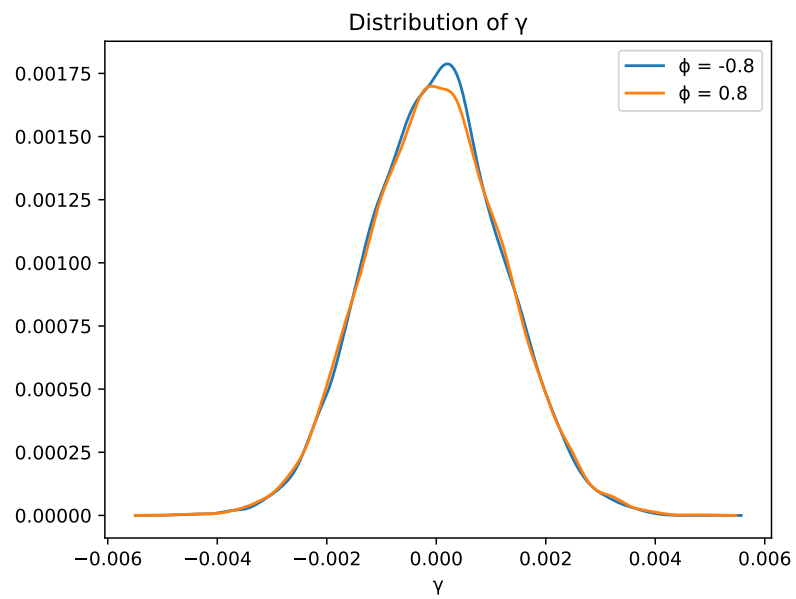
α	OLS
1.00	3.386
0.98	2.4
0.90	0.06858

Q5. I set $T = 200$ and performed 10,000 simulations for each value of ϕ .

i.



ii.



Q6.

$$i. \quad y_{1t} - y_{1t-1} = r y_{2t} - r y_{2t-1} + u_{1t} - u_{1t-1}$$

$$\Rightarrow \Delta y_{1t} = r \Delta y_{2t} + \Delta u_{1t}$$

$$= r u_{2t} + (1-L) u_{1t}$$

$$\begin{bmatrix} \Delta y_{1t} \\ \Delta y_{2t} \end{bmatrix} = \begin{bmatrix} (1-L) & r \\ 0 & 1 \end{bmatrix} \begin{bmatrix} u_{1t} \\ u_{2t} \end{bmatrix}$$

So, the MA(∞) representation is $\Delta y_t = \Psi(L) u_t$
with $\Psi(L) \equiv \begin{bmatrix} 1-L & r \\ 0 & 1 \end{bmatrix}$

$$ii. \quad y_{1t} = r y_{2t} + u_{1t}$$

$$y_{2t} = y_{2t-1} + u_{2t}$$

$$y_t = \begin{bmatrix} 0 & r \\ 0 & 1 \end{bmatrix} y_t + u_t$$

$$\Rightarrow u_t = \begin{bmatrix} 1 & -r \\ 0 & 1-L \end{bmatrix} y_t$$

So, the AR(∞) representation is $\Phi(L) y_t = u_t$
with $\Phi(L) \equiv \begin{bmatrix} 1 & -r \\ 0 & 1-L \end{bmatrix}$

$$iii \quad \Delta y_{1t} = r \Delta y_{2t} + \Delta u_{1t} = r u_{2t} + u_{1t} - (y_{1t-1} - r y_{2t-1})$$

$$\Delta y_{2t} = u_{2t}$$

$$\begin{aligned} \Rightarrow \Delta y_t &= \begin{bmatrix} -1 & r \\ 0 & 0 \end{bmatrix} y_t + \begin{bmatrix} 1 & r \\ 0 & 1 \end{bmatrix} u_t \\ &= \underbrace{\begin{bmatrix} -1 \\ 0 \end{bmatrix}}_{\beta} \underbrace{\begin{bmatrix} 1 & -r \end{bmatrix}}_{\alpha'} y_t + \underbrace{\begin{bmatrix} 1 & r \\ 0 & 1 \end{bmatrix}}_{e_t} u_t \end{aligned}$$

So, the VECM representation is $\Delta y_t = \beta \alpha' y_t + e_t$ with
 $\beta \equiv \begin{bmatrix} -1 \\ 0 \end{bmatrix}$, $\alpha \equiv \begin{bmatrix} 1 \\ -r \end{bmatrix}$, $e_t \equiv \begin{bmatrix} 1 & r \\ 0 & 1 \end{bmatrix} u_t$