

# Summer Intern Lectures

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## 1 State Space Models

### 1.1 What is a state space model?

A state space model is a way to represent the dynamics of a set of variables. A models for observable variables  $y_{i,t}$  can be thus described by a model of (possibly unobserved)  $x_{i,t}$ , i.e. the state variables. Any state space model can be rewritten as a system of an **observation equations**, relating the state space to the observation space, and a *transformation equation*, describing the dynamics of the state space.

In its simplest form, a state space model can be characterized by the following system:

$$y_t = G \cdot x_t \quad \text{observation equation} \quad (1)$$

$$x_{t+1} = A \cdot x_t + C \cdot w_{t+1}, w_{t+1} \sim \mathcal{N}(0, I) \quad \text{transition equation} \quad (2)$$

Note 1: this can be a higher-dimensional model as well, just make  $y_t$  a  $k \times 1$  vector and  $x_t$  a  $n \times 1$  vector.

Then what do we have:

- $G$  is a  $k \times n$  matrix known as the **output** matrix, relating state model to observable model.
- $A$  is a  $n \times n$  matrix known as the **transition** matrix, indicating the evolution of the state variable(s).
- $C$  is a  $\times m$  matrix known as the **volatility** matrix, indicating how uncertainty enters the model.
- $w_{t+1}$  is an  $m \times 1$  vector of randomness.

Note 2: The state and observation models can have different dimensions! This will be very useful for *Nowcasting* framework.

#### Example 1

The model for the variable  $y_t$  is:

$$y_{t+1} = \phi_0 + \phi_1 y_t + \phi_2 y_{t-1} \quad (3)$$

given  $y_0, y_{-1}$ . How do you convert this into a state space model?

Look at what variables enter the model. It's a constant term as well as  $y_t$  and  $y_{t-1}$ .

$$x_t := [1 \ y_t \ y_{t-1}]'.$$

Recall that the observation equation will be of the form  $y_t = Gx_t$ . Given our state vector  $x_t$  what is  $G$ ?

$G = [0 \ 1 \ 0]$  works!

$$G \cdot x_t = (0 \ 1 \ 0) \cdot \begin{pmatrix} 1 \\ y_t \\ y_{t-1} \end{pmatrix} = y_t.$$

What are the dynamics of the state variable? We want to be able to reproduce the initial model (3) by relating  $x_{t+1}$  and  $x_t$  as in the transition equation:  $x_{t+1} = Ax_t + Cw_{t+1}$ . What are  $A$  and  $C$ ?

$$x_{t+1} = \begin{pmatrix} 1 \\ y_{t+1} \\ y_t \end{pmatrix} = A \begin{pmatrix} 1 \\ y_t \\ y_{t-1} \end{pmatrix}$$

Solve analytically here, or just eye-ball it!  $A = \begin{pmatrix} 1 & 0 & 0 \\ \phi_1 & \phi_2 & \phi_3 \\ 0 & 1 & 0 \end{pmatrix}$  works!

Finally, what is  $C$ ? There is no randomness, so it  $C = \vec{0}$ .

For you to do:

- 1)  $y_{t+1} = \phi_1 y_t + \phi_2 y_{t-1} + \phi_3 y_{t-2} + \phi_4 y_{t-3} + \sigma w_{t+1}$  with  $w_t \sim \mathcal{N}(0, 1)$ .
- 2) As above but  $y_t$  is a  $k \times 1$  vector (i.e. it represents  $k$  many variables at time  $t$ ).  $\phi_j$  is  $k \times k$  matrix and  $w_t$  is  $k \times 1$  vector. This is then a vector auto regression!