

$$W(\theta_t) = \max_{x_t} F(\theta_t, x_t) + \beta W(\theta_{t+1})$$

Envelope $\frac{\partial W_t}{\partial \theta_t} = \frac{\partial F}{\partial \theta_t} + \beta A_t \frac{\partial W_{t+1}}{\partial \theta_{t+1}}$

$\begin{matrix} m \times 1 \\ m \times 1 \\ m \times m \\ m \times 1 \end{matrix}$

x_t is a vector of Choice Vars
 θ_t is a vector of State Vars

$\begin{matrix} k \times 1 \\ m \times 1 \end{matrix}$

$$\frac{1}{\beta} A_t^{-1} \left(\frac{\partial W_t}{\partial \theta_t} - \frac{\partial F}{\partial \theta_t} \right) = \frac{\partial W_{t+1}}{\partial \theta_{t+1}}$$

Optimality

$$0 = \frac{\partial F_t}{\partial x_t} + \beta B_t \frac{\partial W_{t+1}}{\partial \theta_{t+1}}$$

$\begin{matrix} k \times 1 \\ k \times m \\ m \times 1 \end{matrix}$

thus

$$0 = \frac{\partial F_t}{\partial x_t} + \beta B_t \frac{1}{\beta} A_t^{-1} \left(\frac{\partial W_t}{\partial \theta_t} - \frac{\partial F_t}{\partial \theta_t} \right)$$

$$-\frac{\partial F_t}{\partial x_t} + B_t A_t^{-1} \frac{\partial F_t}{\partial \theta_t} = B_t A_t^{-1} \frac{\partial W_t}{\partial \theta_t}$$

If $k=m$, then

$$\frac{\partial W_t}{\partial \theta_t} = -A_t B_t^{-1} \frac{\partial F_t}{\partial x_t} + \frac{\partial F_t}{\partial \theta_t}$$

But $k < m$ (usually $k+1=m$ or $2k+1=m$, in my problems)

So we need to construct extra Rows.

- This can be done by taking the first $m-k$ Rows of the optimality condition and iterating forward one period to $t+2$, then substituting in the envelope conditions twice. Add these conditions to the system, and you have an invertible system.

See Next Page.

Declare

Opt 1

$$0 = \frac{\partial F_t}{\partial x_t} + \beta B_t \frac{\partial W_{t+1}}{\partial \theta_{t+1}}$$

Opt 2

$$0 = \frac{\partial F_{t+1}}{\partial x_{t+1}} + \beta B_{t+1} \frac{\partial W_{t+2}}{\partial \theta_{t+2}}$$

Substitute

$$0 = \frac{\partial F_t}{\partial x_t} + B_t A_t^{-1} \left(\frac{\partial W_t}{\partial \theta_t} - \frac{\partial F_t}{\partial \theta_t} \right)$$

Substitute
twice

$$0 = \frac{\partial F_{t+1}}{\partial x_{t+1}} (\theta_t, x^*(\theta_t)) + \beta B_{t+1}(\theta_t) \left(\frac{A_{t+1}^{-1}}{\beta} \right) \left(\frac{\partial W_{t+1}}{\partial \theta_{t+1}} - \frac{\partial F_{t+1}}{\partial \theta_{t+1}} \right)$$

$$0 = \frac{\partial F}{\partial x} (\theta_{t+1}(\theta_t), x_{t+1}^*(\theta_t)) + B_{t+1}(\theta_t) A_{t+1}^{-1}(\theta_t) \left(-\frac{\partial F_{t+1}}{\partial \theta_{t+1}} + \frac{A_t^{-1}}{\beta} \left(\frac{\partial W_t}{\partial \theta_t} - \frac{\partial F_t}{\partial \theta_t} \right) \right)$$

$$0 = \frac{\partial F_{t+1}}{\partial x_{t+1}} + B_{t+1} A_{t+1}^{-1} \left(\frac{A_t^{-1}}{\beta} \frac{\partial W_t}{\partial \theta_t} - \frac{A_t^{-1}}{\beta} \frac{\partial F_t}{\partial \theta_t} - \frac{\partial F_{t+1}}{\partial \theta_{t+1}} \right)$$

~~0 =~~

$$0 = \frac{\partial F_{t+1}}{\partial x_{t+1}} - B_{t+1} A_{t+1}^{-1} \frac{\partial F_{t+1}}{\partial \theta_{t+1}} + \frac{B_{t+1} A_{t+1}^{-1} A_t^{-1}}{\beta} \frac{\partial F_t}{\partial \theta_t} + \frac{B_{t+1} A_{t+1}^{-1} A_t^{-1}}{\beta} \frac{\partial W_t}{\partial \theta_t}$$

Which is another valid set of conditions.

Note that $\theta_{t+1}^*(\theta_t, x_t^*(\theta_t))$ etc, are a function of the policy function

$$0 = \frac{\partial F^*}{\partial x} (\theta_{t+1}^*(\theta_t)) + B(\theta_{t+1}^*(\theta_t)) A^{-1}(\theta_{t+1}^*(\theta_t)) \left[\frac{\partial F^*}{\partial \theta} (\theta_t^*(\theta_t)) + \frac{A^{-1}(\theta_t)}{\beta} \left(\frac{\partial W}{\partial \theta} (\theta_t) + \frac{\partial F^*}{\partial \theta} (\theta_t) \right) \right]$$

ie. they implicitly define the Policy function.

By appending the first $M-K$ Rows of the Second to the first we get a set of M expressions, which determine the M functions $\frac{\partial W}{\partial \theta}$. By Iterating the optimality condition Back a single Period to t , we get an implicit definition of $x_t^*(\theta_t)$.