

SECOND-ORDER APPROXIMATION

JANUARY 17, 2018

SECOND-ORDER PERTURBATION

□ For $f(x, y, z) = 0$, second-order Taylor expansion around $(\bar{x}, \bar{y}, \bar{z})$

$$f(x, y, z) \approx f(\bar{x}, \bar{y}, \bar{z}) + f_x(\bar{x}, \bar{y}, \bar{z})(x - \bar{x}) + f_y(\bar{x}, \bar{y}, \bar{z})(y - \bar{y}) + f_z(\bar{x}, \bar{y}, \bar{z})(z - \bar{z})$$

(Illustrative
example in scalars)

SECOND-ORDER PERTURBATION

- For $f(x, y, z) = 0$, second-order Taylor expansion around $(\bar{x}, \bar{y}, \bar{z})$

$$\begin{aligned} f(x, y, z) &\approx f(\bar{x}, \bar{y}, \bar{z}) + f_x(\bar{x}, \bar{y}, \bar{z})(x - \bar{x}) + f_y(\bar{x}, \bar{y}, \bar{z})(y - \bar{y}) + f_z(\bar{x}, \bar{y}, \bar{z})(z - \bar{z}) \\ &+ \frac{1}{2} f_{xx}(\bar{x}, \bar{y}, \bar{z})(x - \bar{x})^2 + \frac{1}{2} f_{yy}(\bar{x}, \bar{y}, \bar{z})(y - \bar{y})^2 + \frac{1}{2} f_{zz}(\bar{x}, \bar{y}, \bar{z})(z - \bar{z})^2 \\ &+ \frac{1}{2} f_{xy}(\bar{x}, \bar{y}, \bar{z})(x - \bar{x})(y - \bar{y}) + \frac{1}{2} f_{yx}(\bar{x}, \bar{y}, \bar{z})(y - \bar{y})(x - \bar{x}) + \dots \end{aligned}$$

(Illustrative example in scalars)

- Why a second-order approximation?

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 f(x, y, z) &\approx f(\bar{x}, \bar{y}, \bar{z}) + f_x(\bar{x}, \bar{y}, \bar{z})(x - \bar{x}) + f_y(\bar{x}, \bar{y}, \bar{z})(y - \bar{y}) + f_z(\bar{x}, \bar{y}, \bar{z})(z - \bar{z}) \\
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 \end{aligned}$$

(Illustrative example in scalars)

- Why a second-order approximation?
 - To capture (some...) effect of standard deviations of exogenous shocks on decision functions

- e.g., σ_z

$$\ln z_{t+1} = (1 - \rho_z) \ln \bar{z} + \rho_z \ln z_t + \varepsilon_{t+1}$$

$$\varepsilon_{t+1} \sim \text{iid } N(0, \sigma_z^2)$$

- Recall first-order approximation displays certainty equivalence

LINEARIZATION OF THE RBC MODEL

Define **co-state** vector and **state** vector

$$y_t = \begin{bmatrix} c_t \\ n_t \end{bmatrix} \quad x_t = \begin{bmatrix} k_t \\ z_t \end{bmatrix}$$

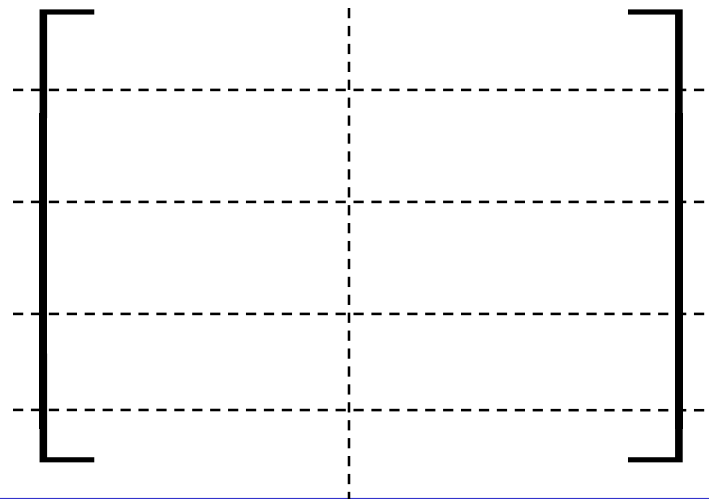
Order model's dynamic equations in a **vector** $\equiv f(y_{t+1}, y_t, x_{t+1}, x_t) = 0$

Consumption-leisure efficiency condition

Consumption-investment efficiency condition

Aggregate resource constraint

Law of motion for TFP



LINEARIZATION OF THE RBC MODEL

- A **quadratic** equation in the elements of g_x and h_x evaluated at the steady state

$$F_x(\bar{x}, 0) = f_{y_{t+1}}(\bar{x}, 0) \cdot g_x(\bar{x}, 0) \cdot h_x(\bar{x}, 0) + f_{y_t}(\bar{x}, 0) \cdot g_x(\bar{x}, 0) + f_{x_{t+1}}(\bar{x}, 0) \cdot h_x(\bar{x}, 0) + f_{x_t}(\bar{x}, 0) = 0$$

- Solve numerically for the elements of g_x and h_x (use fsolve OR conduct an eigenvalue decomposition in Matlab)
- Recall conjectured equilibrium decision rules

$$y_t = g(x_t, \sigma)$$

$$x_{t+1} = h(x_t, \sigma) + \eta\sigma\varepsilon_{t+1}$$

- First-order approximation is

$$y_t = g(x_t, \sigma) \approx g(\bar{x}, 0) + g_x(\bar{x}, 0)(x_t - \bar{x}) + g_\sigma(\bar{x}, 0)\sigma \quad \text{SGU Theorem 1:}$$

$$x_{t+1} = h(x_t, \sigma) \approx h(\bar{x}, 0) + h_x(\bar{x}, 0)(x_t - \bar{x}) + h_\sigma(\bar{x}, 0)\sigma \quad g_\sigma = 0 \text{ and } h_\sigma = 0$$

SECOND-ORDER

- How?...

SECOND-ORDER

- How?...
- The model's dynamic **expectational** equations

$$\begin{aligned}
 E_t [f(y_{t+1}, y_t, x_{t+1}, x_t)] &= 0 \\
 &= E_t [f(g(x_{t+1}, \sigma), g(x_t, \sigma), h(x_t, \sigma) + \eta\sigma\varepsilon_{t+1}, x_t)] \\
 &= E_t [f(g(h(x_t, \sigma) + \eta\sigma\varepsilon_{t+1}, \sigma), g(x_t, \sigma), h(x_t, \sigma) + \eta\sigma\varepsilon_{t+1}, x_t)] \\
 &\equiv F(x_t, \sigma)
 \end{aligned}$$



$$F_x(x_t, \sigma) = 0 \qquad F_\sigma(x_t, \sigma) = 0$$

SECOND-ORDER

- How?...
- The model's dynamic **expectational** equations

$$\begin{aligned}
 E_t [f(y_{t+1}, y_t, x_{t+1}, x_t)] &= 0 \\
 &= E_t [f(g(x_{t+1}, \sigma), g(x_t, \sigma), h(x_t, \sigma) + \eta\sigma\varepsilon_{t+1}, x_t)] \\
 &= E_t [f(g(h(x_t, \sigma) + \eta\sigma\varepsilon_{t+1}, \sigma), g(x_t, \sigma), h(x_t, \sigma) + \eta\sigma\varepsilon_{t+1}, x_t)] \\
 &\equiv F(x_t, \sigma)
 \end{aligned}$$



$$F_x(x_t, \sigma) = 0 \quad F_\sigma(x_t, \sigma) = 0$$

$$F_{xx}(x_t, \sigma) = 0 \quad F_{\sigma\sigma}(x_t, \sigma) = 0 \quad F_{x\sigma}(x_t, \sigma) = 0 \quad F_{\sigma x}(x_t, \sigma) = 0$$

SECOND-ORDER

- ❑ How?...need to compute the second derivatives of the equilibrium conditions
- ❑ SGU (2004 *JEDC*, p. 762)

$$\begin{aligned}
 [F_{xx}(\bar{x}, 0)]_{jk}^i &= ([f_{y'y'}]_{\alpha\gamma}^i [g_x]_{\delta}^{\gamma} [h_x]_k^{\delta} + [f_{y'y'}]_{\alpha\gamma}^i [g_x]_k^{\gamma} \\
 &\quad + [f_{y'x'}]_{\alpha\delta}^i [h_x]_k^{\delta} + [f_{y'x'}]_{\alpha k}^i) [g_x]_{\beta}^{\alpha} [h_x]_j^{\beta} \\
 &\quad + [f_{y'}]_{\alpha}^i [g_{xx}]_{\beta\delta}^{\alpha} [h_x]_k^{\delta} [h_x]_j^{\beta} \\
 &\quad + [f_{y'}]_{\alpha}^i [g_x]_{\beta}^{\alpha} [h_{xx}]_{jk}^{\beta} \\
 &\quad + ([f_{yy'}]_{\alpha\gamma}^i [g_x]_{\delta}^{\gamma} [h_x]_k^{\delta} + [f_{yy'}]_{\alpha\gamma}^i [g_x]_k^{\gamma} + [f_{yx'}]_{\alpha\delta}^i [h_x]_k^{\delta} + [f_{yx'}]_{\alpha k}^i) [g_x]_j^{\alpha} \\
 &\quad + [f_y]_{\alpha}^i [g_{xx}]_{jk}^{\alpha} \\
 &\quad + ([f_{x'y'}]_{\beta\gamma}^i [g_x]_{\delta}^{\gamma} [h_x]_k^{\delta} + [f_{x'y'}]_{\beta\gamma}^i [g_x]_k^{\gamma} + [f_{x't'}]_{\beta\delta}^i [h_x]_k^{\delta} + [f_{x't'}]_{\beta k}^i) [h_x]_j^{\beta} \\
 &\quad + [f_{x'}]_{\beta}^i [h_{xx}]_{jk}^{\beta} \\
 &\quad + [f_{xy'}]_{j\gamma}^i [g_x]_{\delta}^{\gamma} [h_x]_k^{\delta} + [f_{xy'}]_{j\gamma}^i [g_x]_k^{\gamma} + [f_{xx'}]_{j\delta}^i [h_x]_k^{\delta} + [f_{xx'}]_{jk}^i \\
 &= 0; \quad i = 1, \dots, n, \quad j, k, \beta, \delta = 1, \dots, n_x; \quad \alpha, \gamma = 1, \dots, n_y.
 \end{aligned}$$

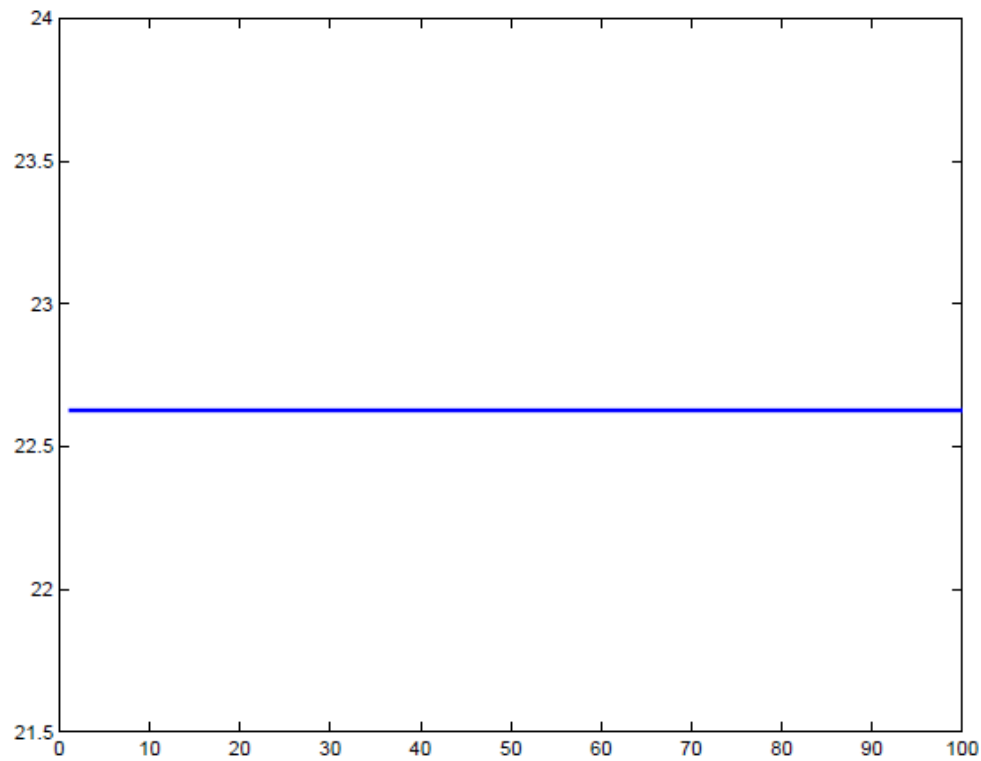
SECOND-ORDER

- How?...need to compute the second derivatives of the equilibrium conditions
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$$\begin{aligned}
 [F_{\sigma\sigma}(\bar{x}, 0)]^i &= [f_{y'}]_{\alpha}^i [g_x]_{\beta}^{\alpha} [h_{\sigma\sigma}]^{\beta} \\
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 &\quad \phi, \xi = 1, \dots, n_e.
 \end{aligned}$$

SECOND-ORDER APPROXIMATION OF RBC MODEL

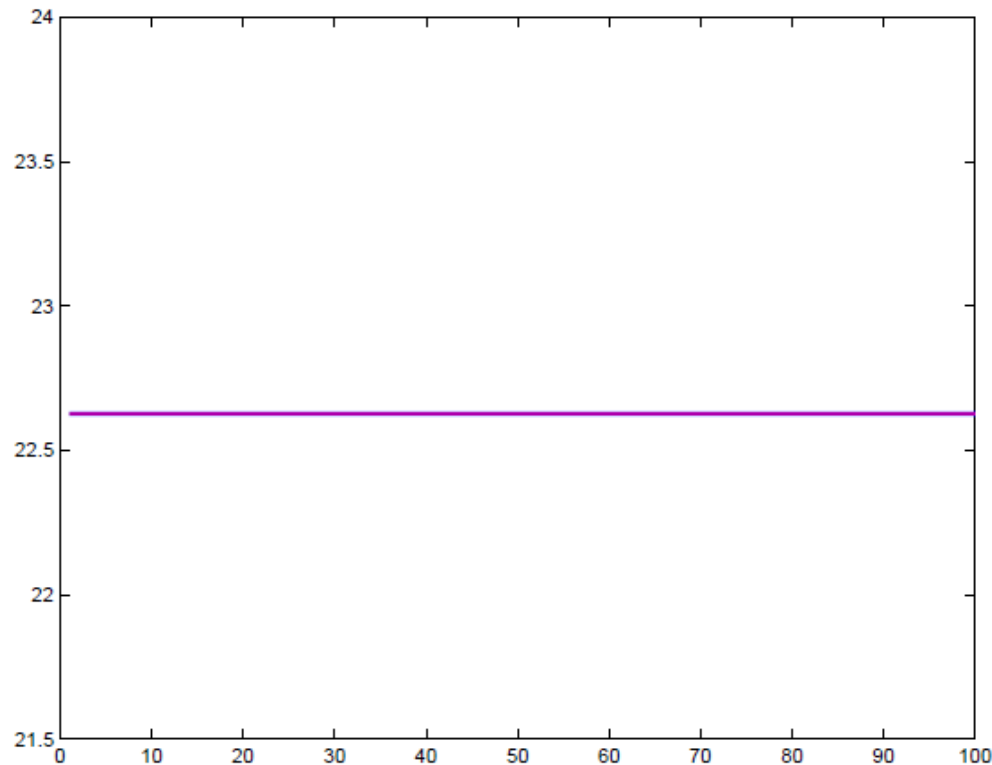
- ❑ Suppose sequence of **realized** shocks to TFP is **always 0**
- ❑ Compare **linear-** vs. **second-order** simulations
- ❑ For $\sigma_z = 0$



SECOND-ORDER APPROXIMATION OF RBC MODEL

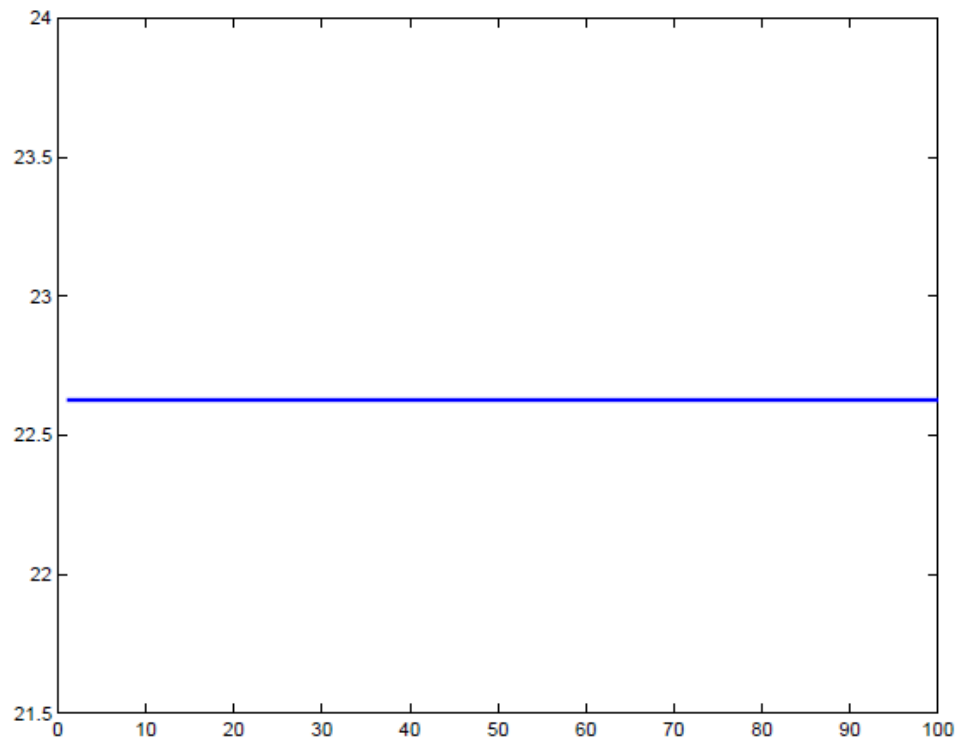
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**First-order and
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simulations
identical**



SECOND-ORDER APPROXIMATION OF RBC MODEL

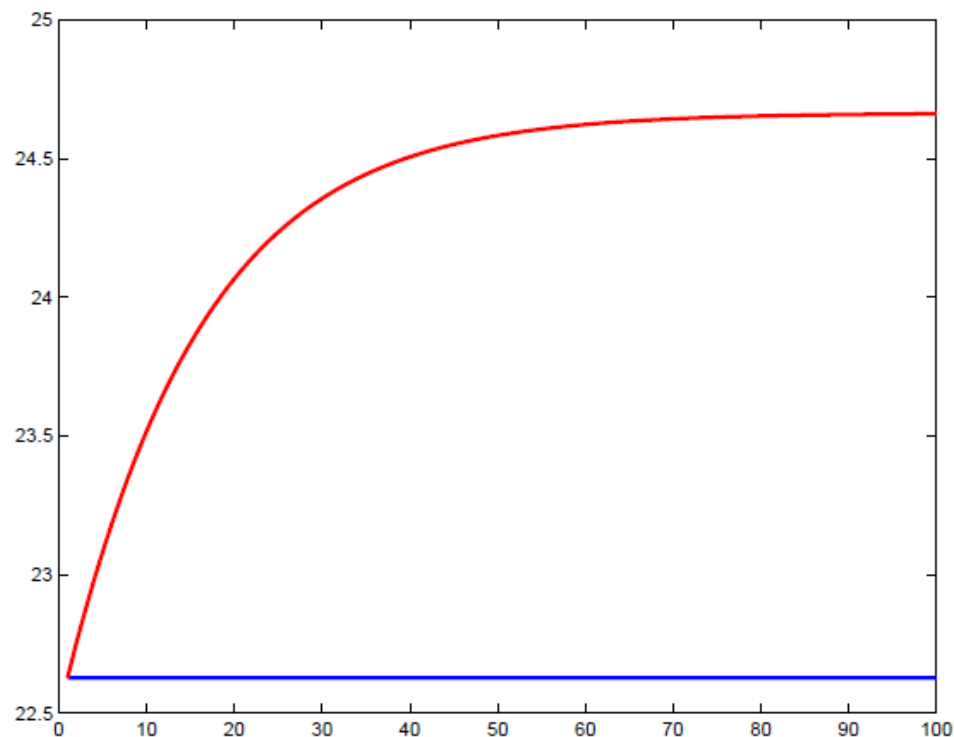
- ❑ Suppose sequence of **realized** shocks to TFP is **always 0**
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- ❑ For $\sigma_z = 0.007$



SECOND-ORDER APPROXIMATION OF RBC MODEL

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- ❑ Compare **linear-** vs. **second-order** simulations
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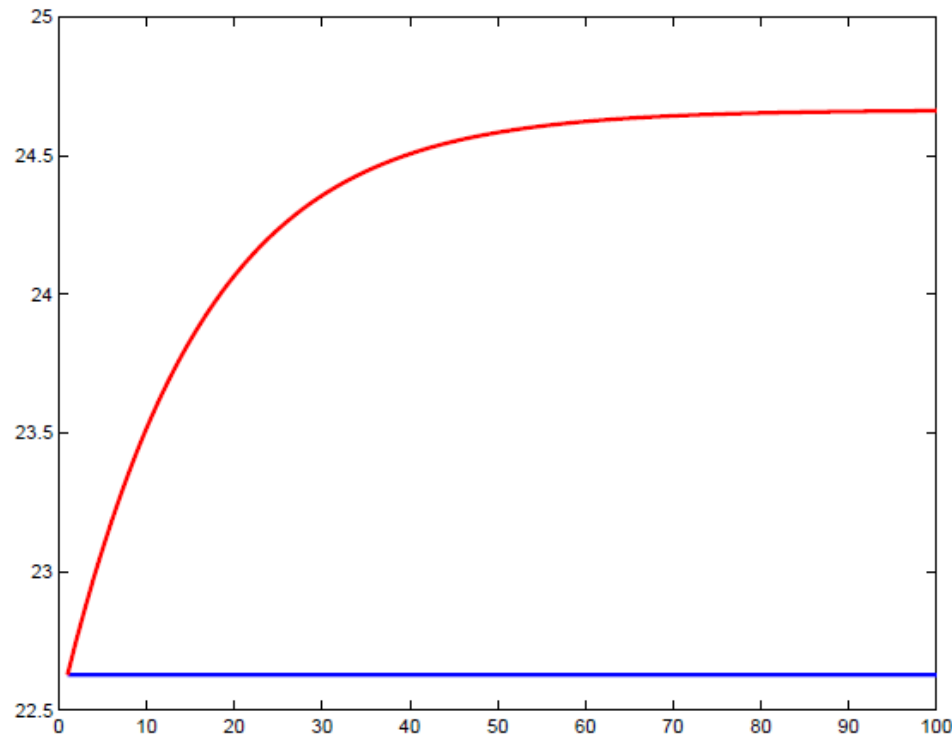
First-order and second-order simulations NOT identical



SECOND-ORDER APPROXIMATION OF RBC MODEL

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- ❑ Compare **linear-** vs. **second-order** simulations
- ❑ For $\sigma_z = 0.007$

First-order and second-order simulations NOT identical



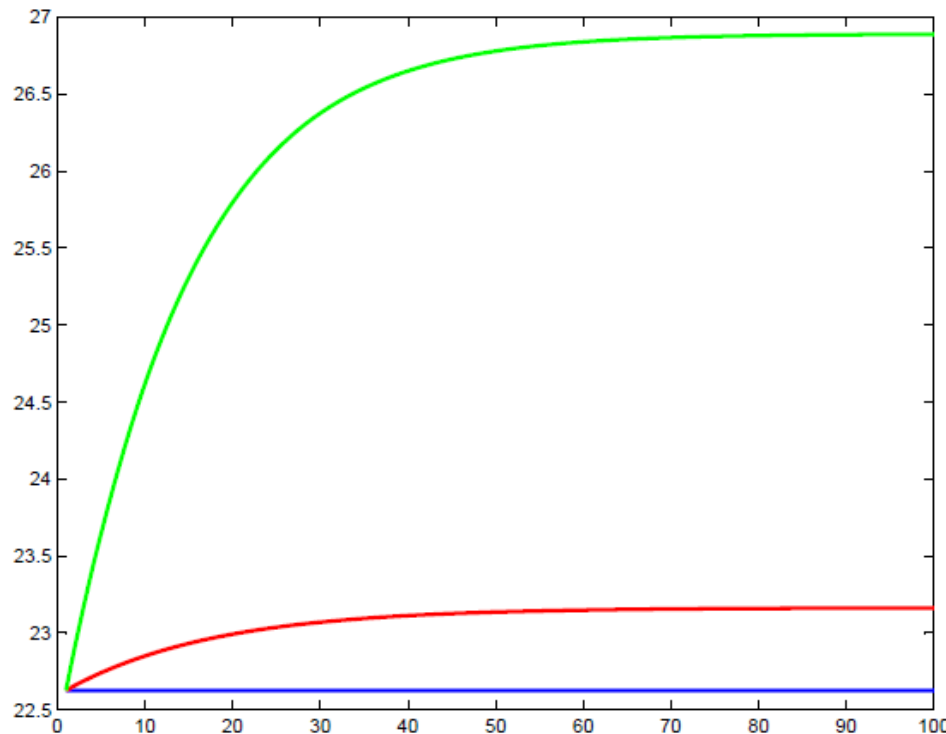
"Stochastic" steady state
(mean of the ergodic distribution)

Deterministic steady state

SECOND-ORDER APPROXIMATION OF RBC MODEL

- ❑ Suppose sequence of **realized** shocks to TFP is **always 0**
- ❑ Compare **linear-** vs. **second-order** simulations
- ❑ Compare $\sigma_z = 0.007$ vs. $\sigma_z = 0.021$

First-order and second-order simulations NOT identical



"Stochastic" steady state
(mean of the ergodic distribution)

Deterministic steady state

STOCHASTIC STEADY STATE

- ❑ Second-order approximation terms for the decision functions $g(x_t, \sigma)$ and $h(x_t, \sigma)$ **start** capturing the “precautionary savings” motive
- ❑ Precaution against what?

STOCHASTIC STEADY STATE

- Second-order approximation terms for the decision functions $g(x_t, \sigma)$ and $h(x_t, \sigma)$ **start** capturing the “precautionary savings” motive
- Precaution against what?
- Against how large shocks to the exogenous processes (e.g., TFP) **could** be...
- **...NOT** the **actual realizations** of the TFP shocks
- **$g_{\sigma\sigma}(x_t, \sigma)$ and $h_{\sigma\sigma}(x_t, \sigma)$ are NOT zero**
- (Recall $g_{\sigma}(x_t, \sigma) = 0$ and $h_{\sigma}(x_t, \sigma) = 0$)

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Can we compute by hand (“fsolve”) the stochastic steady state?

STOCHASTIC STEADY STATE

- Shut down all shocks and set exogenous variables at their means
- The idea: let economy run for many (infinite) periods
 - Time eventually “doesn’t matter” any more
 - Drop all time indices

$$-\frac{u_n(\bar{c}, \bar{n})}{u_c(\bar{c}, \bar{n})} = \bar{z}F_n(\bar{k}, \bar{n})$$

$$u_c(\bar{c}, \bar{n}) = \beta u_c(\bar{c}, \bar{n}) \left[\bar{z}F_k(\bar{k}, \bar{n}) + 1 - \delta \right]$$

$$\bar{c} + \delta \bar{k} = \bar{z}F(\bar{k}, \bar{n})$$

$$\ln \bar{z} = (1 - \rho_z) \ln \bar{z} + \rho_z \ln \bar{z} \Rightarrow \bar{z} = \bar{z} \quad (\text{a parameter of the model})$$

- Given functional forms and parameter values, solve for $(\bar{c}, \bar{n}, \bar{k})$
 - The steady state of the model
 - Taylor expansion around this point

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- ❑ $g_{\sigma\sigma}(x_t, \sigma)$ and $h_{\sigma\sigma}(x_t, \sigma)$ are **NOT** zero
- ❑ (Recall $g_{\sigma}(x_t, \sigma) = 0$ and $h_{\sigma}(x_t, \sigma) = 0$)

Can we compute by hand (“fsolve”) the stochastic steady state?

NO! – HAS TO BE NUMERICALLY COMPUTED