

New Keynesian Model: Approximation

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NK Model is often analyzed in LQ—Linear-Quadratic—framework.

- Easier to interpret.
- Easier to solve.

Technical takeaways from this lecture:

- Learn how to take the first-order approximation to the equilibrium conditions.
- Learn how to take the second-order approximation to the household's welfare.

Substantive takeaways from this lecture (I):

- Output today depends on the expected sum of future real interest rates.
- Inflation today depends on the expected discounted sum of future marginal costs.

Substantive takeaways from this lecture (II):

- Welfare today depends on the expected discounted sum of future per-period utility where…
 - Per-period utility depends on inflation and output volatility.
 - Per-period utility also depends on the level of output if steady state is not efficient (if τ is not $1/(\theta - 1)$).

Plan

- (1) Log-linear Approximation of the Equilibrium Conditions
- (2) Quadratic Approximation of the Household's Welfare

Plan

(1) Log-linear Approximation of the Equilibrium Conditions

(2) Quadratic Approximation of the Household's Welfare

Math Background

Log-linearization =

First-order Taylor expansion

+

Log-approximation of deviation

First-order Taylor expansion:

Let $f(x)$ be a continuous and differentiable function of x .
Then,

$$f(x) \approx f(x_0) + \frac{\partial f}{\partial x}(x_0)(x - x_0) + (\text{higher order terms})$$

Let f_x denote $\frac{\partial f}{\partial x}$.

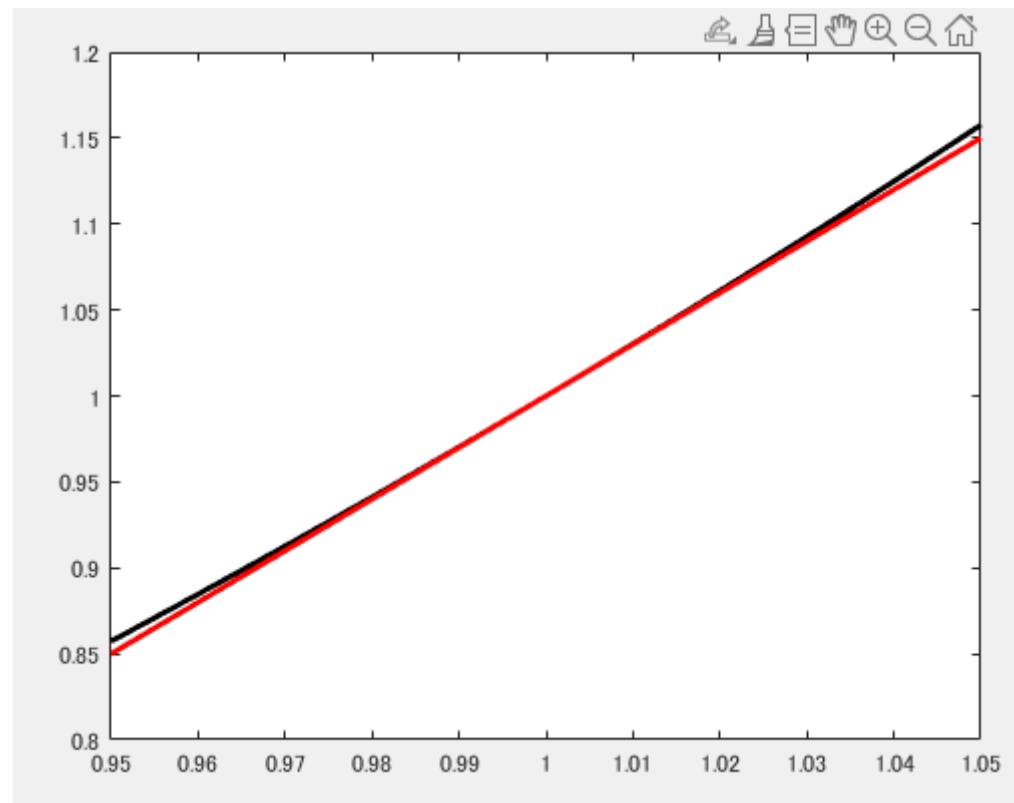
Example:

$$f(x) = x^3$$

$$f_x = 3x^2$$

Let's take the first-order Taylor expansion around $x_0 = 1$.

$$\begin{aligned} f(x) &\approx f(1) + f_x(1)(x - 1) + (\text{h.o.t.}) \\ &= 1^3 + 3 * 1^2(x - 1) + (\text{h.o.t.}) \\ &= 1 + 3(x - 1) + (\text{h.o.t.}) \\ &= -2 + 3x + (\text{h.o.t.}) \end{aligned}$$

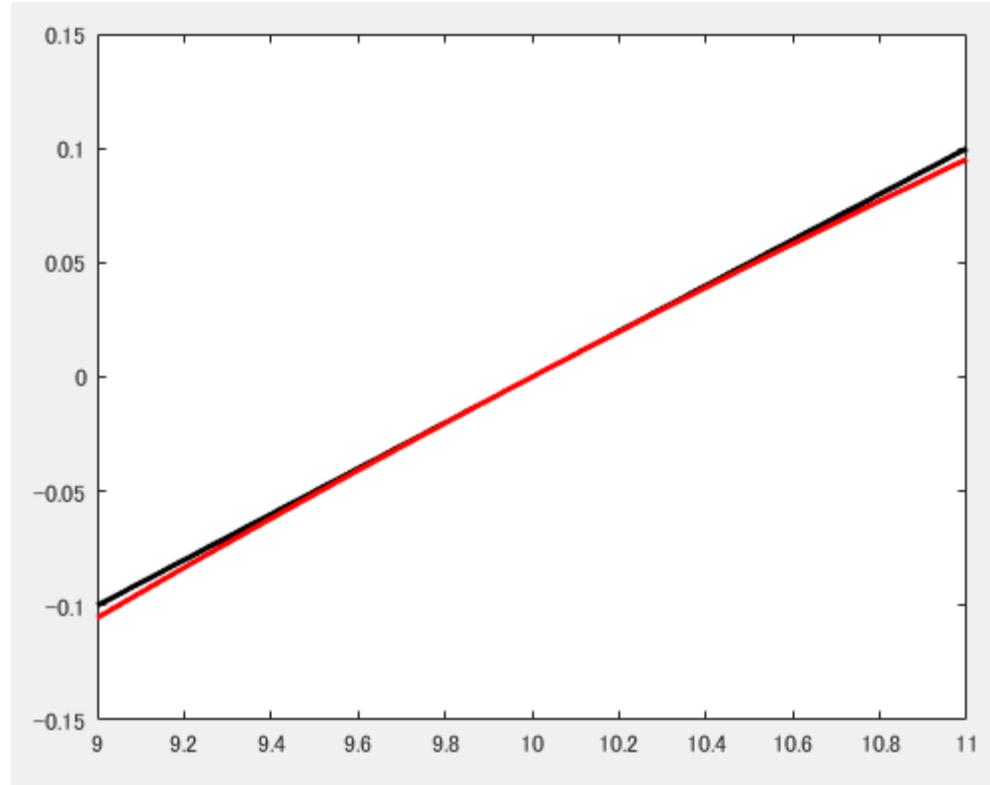


Log-approximation of percent deviation:

$$\frac{(x - x_0)}{x_0} \approx \log(x) - \log(x_0)$$

Example:

$$\frac{(x - 10)}{10} \approx \log(x) - \log(10)$$



First-order Taylor expansion + Log-approximation of percent changes:

$$f(x) \approx f(x_0) + f_x(x_0)(x - x_0) + (\text{h.o.t.})$$

$$= f(x_0) + f_x(x_0)x_0 \frac{(x - x_0)}{x_0} + (\text{h.o.t.})$$

$$\approx f(x_0) + f_x(x_0)x_0(\log(x) - \log(x_0)) + (\text{h.o.t.})$$

$$\approx f(x_0) + f_x(x_0)x_0 \hat{x} + (\text{h.o.t.})$$

where $\hat{x} := \log(x) - \log(x_0)$.

To summarize

Log-linearization:

$$f(x) \approx f(x_0) + f_x(x_0)x_0 \hat{x} + (\text{h.o.t.})$$

where $\hat{x} := \log(x) - \log(x_0)$.

Second-order Taylor expansion:

Let $f(x)$ be a continuous and differentiable function of x .
Then,

$$f(x) \approx f(x_0) + \frac{\partial f}{\partial x}(x_0)(x - x_0) + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(x_0)(x - x_0)^2 + (\text{h. o. t.})$$

Let f_x and f_{xx} denote $\frac{\partial f}{\partial x}$ and $\frac{\partial^2 f}{\partial x^2}$, respectively.

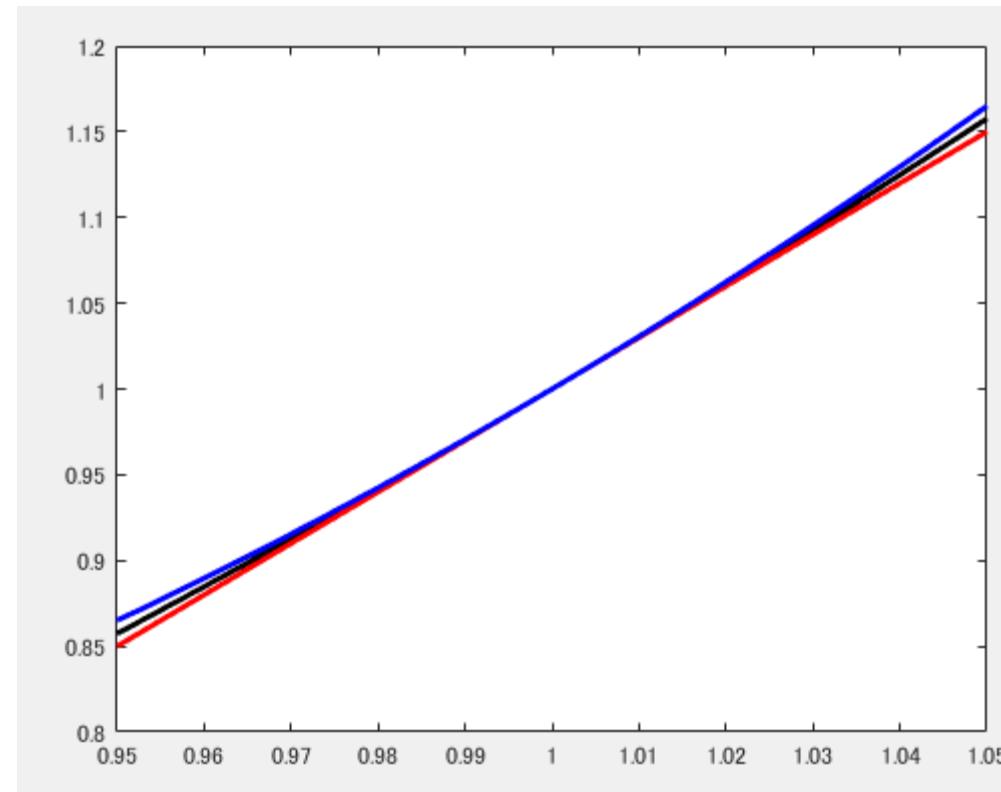
Example:

$$f(x) = x^3$$

$$\begin{aligned}f_x &= 3x^2 \\f_{xx} &= 6x\end{aligned}$$

Let's take the second-order Taylor expansion around $x_0 = 1$.

$$\begin{aligned}f(x) &\approx f(1) + f_x(1)(x - 1) + f_{xx}(1)(x - 1)^2 + (\text{h. o. t.}) \\&= 1^3 + 3 * 1^2(x - 1) + 6 * 1^3(x - 1)^2 + (\text{h. o. t.}) \\&= 1 + 3(x - 1) + 6(x - 1)^2 + (\text{h. o. t.})\end{aligned}$$



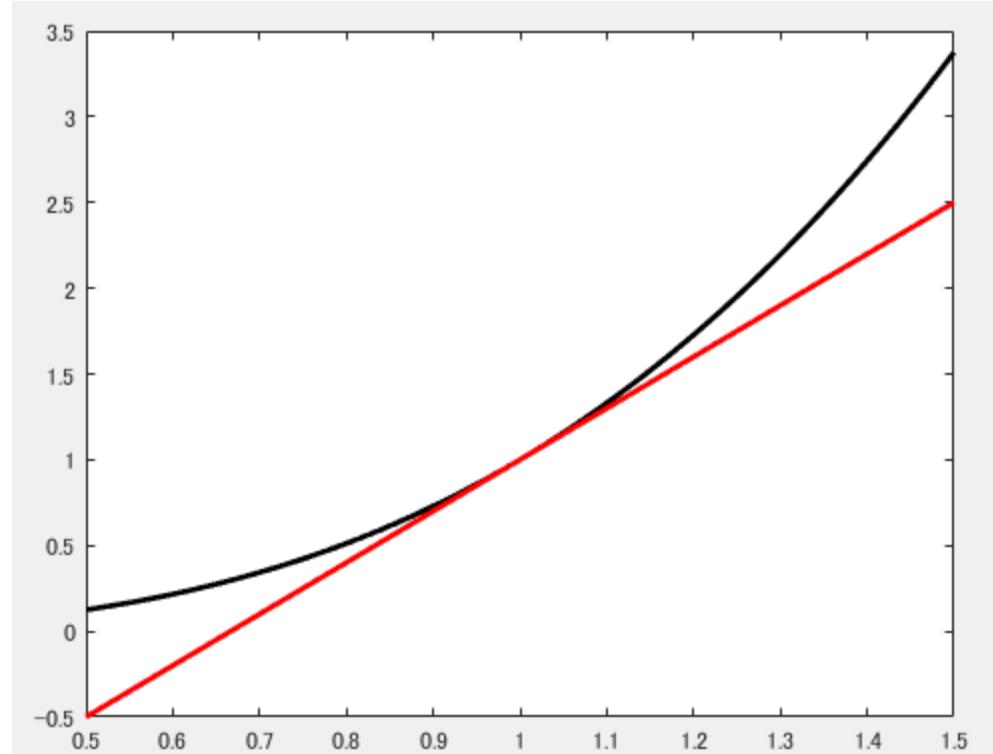
Combined with log-approximation of percent changes…

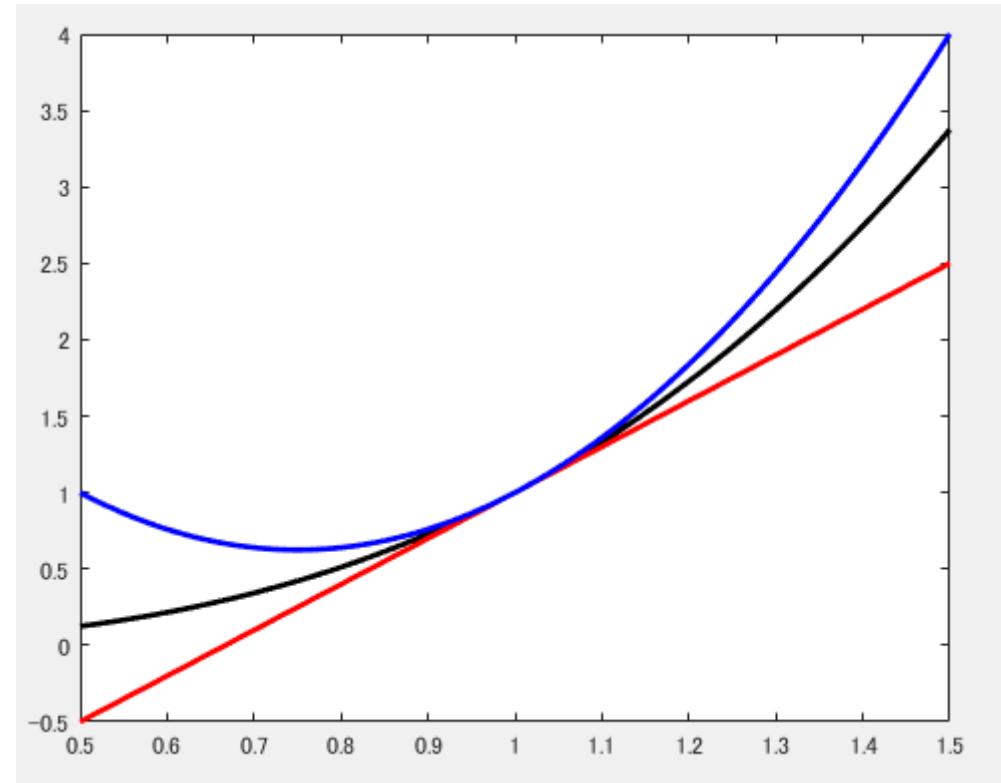
$$f(x) \approx f(x_0) + f_x(x_0)x_0 \hat{x} + f_{xx}(x_0)x_0^2 \hat{x}^2 + (\text{h. o. t.})$$

where $\hat{x} := \log(x) - \log(x_0)$.

Word of caution on Taylor expansion.

They are good “local” approximation, but may not be good globally.





End of Math Background

Recall:

$$C_t^{-\chi_c} = \beta \delta_t R_t C_{t+1}^{-\chi_c} \Pi_{t+1}^{-1}$$

$$w_t = N_t^{\chi_n} C_t^{\chi_c}$$

$$\frac{Y_t}{C_t^{\chi_c}} [\varphi(\Pi_t - 1)\Pi_t - (1 + \tau)(1 - \theta) - \theta w_t]$$

$$= \beta \delta_t \frac{Y_{t+1}}{C_{t+1}^{\chi_c}} \varphi(\Pi_{t+1} - 1)\Pi_{t+1}$$

$$Y_t = C_t + \frac{\varphi}{2} [\Pi_t - 1]^2 Y_t$$

$$Y_t = N_t$$

$$R_t = \max \left[\frac{1}{\beta} \Pi_t^{\Phi_\pi}, 1 \right]$$

$$U_t = \frac{C_t^{1-\chi_c}}{1-\chi_c} - \frac{N_t^{1+\chi_n}}{1+\chi_n}$$

Loglinearize the consumption Euler equation around the zero-inflation steady state:

$$C_t^{-\chi_c} R_t^{-1} = \beta \delta_t C_{t+1}^{-\chi_c} \Pi_{t+1}^{-1}$$

$$\text{LHS} = f(C_t, R_t)$$

$$\text{RHS} = g(\delta_t, C_{t+1}, \Pi_{t+1})$$

$$f_C = -\chi_c C_t^{-\chi_c - 1} R_t^{-1}$$
$$f_R = -C_t^{-\chi_c} R_t^{-2}$$

$$g_\delta = \beta C_{t+1}^{-\chi_c} \Pi_{t+1}^{-1}$$
$$g_C = -\chi_c \beta C_{t+1}^{-\chi_c - 1} \Pi_{t+1}^{-1}$$
$$g_\Pi = -\beta C_{t+1}^{-\chi_c} \Pi_{t+1}^{-2}$$

Notation: $h_x := \frac{\partial h}{\partial x}$ for any function h and for any variable x .

$$\text{LHS} \approx f(C_{ss}, R_{ss}) + f_{C,ss} C_{ss} \hat{C}_t + f_{R,ss} R_{ss} \hat{R}_t$$

$$= f(C_{ss}, R_{ss}) + f(C_{ss}, R_{ss})[-\chi_c \hat{C}_t] + f(C_{ss}, R_{ss})[-\hat{R}_t]$$

$$\text{RHS} \approx g(\delta_{ss}, C_{ss}, \Pi_{ss}) + g_{\delta,ss} \delta_{ss} \hat{\delta}_t + g_{C,ss} C_{ss} \hat{C}_{t+1} + g_{\Pi,ss} \Pi_{ss} \hat{\Pi}_{t+1}$$

$$= g(\delta_{ss}, C_{ss}, \Pi_{ss}) + g(\delta_{ss}, C_{ss}, \Pi_{ss}) \hat{\delta}_t \\ + g(\delta_{ss}, C_{ss}, \Pi_{ss})[-\chi_c \hat{C}_{t+1}] + g(\delta_{ss}, C_{ss}, \Pi_{ss})[-\hat{\Pi}_{t+1}]$$

Using $f(C_{ss}, R_{ss}) = g(\delta_{ss}, C_{ss}, \Pi_{ss})$ and rearranging terms,

$$\hat{C}_t = \hat{C}_{t+1} - \frac{1}{\chi_c} [\hat{R}_t - \hat{\Pi}_{t+1} + \hat{\delta}_t]$$

Loglinearize the intratemporal optimality condition around the zero-inflation steady state:

$$w_t = N_t^{\chi_n} C_t^{\chi_c}$$

$$\text{LHS} = f(w_t)$$

$$\text{RHS} = g(C_t, N_t)$$

$$f_w = 1$$

$$g_C = \chi_c N_t^{\chi_n} C_t^{\chi_c - 1}$$

$$g_N = \chi_n N_t^{\chi_n - 1} C_t^{\chi_c}$$

$$\begin{aligned}\text{LHS} &\approx f(w_{ss}) + f_{w,ss} w_{ss} \hat{w}_t \\ &= f(w_{ss}) + f(w_{ss}) \hat{w}_t\end{aligned}$$

$$\begin{aligned}\text{RHS} &\approx g(C_{ss}, N_{ss}) + g_{C,ss} C_{ss} \hat{C}_t + g_{N,ss} N_{ss} \hat{N}_t \\ &= g(C_{ss}, N_{ss}) + g(C_{ss}, N_{ss}) [\chi_c \hat{C}_t] + g(C_{ss}, N_{ss}) [\chi_N \hat{N}_t]\end{aligned}$$

Using $f(w_{ss}) = g(C_{ss}, N_{ss})$ and rearranging terms,

$$\hat{w}_t = \chi_c \hat{C}_t + \chi_N \hat{N}_t$$

Loglinearize the optimality condition of intermediate-goods producers **around the zero-inflation steady state**:

$$Y_t C_t^{-\chi_c} [\varphi(\Pi_t - 1)\Pi_t - (1 - \theta)(1 + \tau) - \theta w_t]$$

$$= \beta Y_{t+1} C_{t+1}^{-\chi_c} \varphi(\Pi_{t+1} - 1)\Pi_{t+1}$$

$$\text{LHS} = f(Y_t, C_t, \Pi_t, w_t)$$

$$\text{RHS} = g(Y_{t+1}, C_{t+1}, \Pi_{t+1})$$

$$f_Y = C_t^{-\chi_c} [\varphi(\Pi_t - 1)\Pi_t - (1 - \theta)(1 + \tau) - \theta w_t]$$

$$f_C = -\chi_c Y_t C_t^{-\chi_c - 1} [\varphi(\Pi_t - 1)\Pi_t - (1 - \theta)(1 + \tau) - \theta w_t]$$

$$f_{\Pi} = Y_t C_t^{-\chi_c} \varphi [2\Pi_t - 1]$$

$$f_w = -Y_t C_t^{-\chi_c} \theta$$

$$g_Y = \beta C_{t+1}^{-\chi_c} \varphi(\Pi_{t+1} - 1)\Pi_{t+1}$$

$$g_C = -\chi_c \beta Y_{t+1} C_{t+1}^{-\chi_c - 1} \varphi(\Pi_{t+1} - 1)\Pi_{t+1}$$

$$g_{\Pi} = \beta Y_{t+1} C_{t+1}^{-\chi_c} \varphi [2\Pi_{t+1} - 1]$$

Note that $\Pi_{ss} = 1$ and $w_{ss} = \frac{(\theta-1)(1+\tau)}{\theta}$. Thus,

$$f_{Y,ss} = 0$$

$$f_{C,ss} = 0$$

$$f_{\Pi,ss} = Y_{ss} C_{ss}^{-\chi_c} \varphi$$

$$f_{w,ss} = -Y_{ss} C_{ss}^{-\chi_c} \theta$$

$$g_{Y,ss} = 0$$

$$g_{C,ss} = 0$$

$$g_{\Pi,ss} = \beta Y_{ss} C_{ss}^{-\chi_c} \varphi$$

$$\begin{aligned}\text{LHS} &\approx f(Y_{ss}, C_{ss}, \Pi_{ss}, w_{ss}) + f_{\Pi,ss} \Pi_{ss} \widehat{\Pi}_t + f_{w,ss} w_{ss} \widehat{w}_t \\ &= f(Y_{ss}, C_{ss}, \Pi_{ss}, w_{ss}) + Y_{ss} C_{ss}^{-\chi_c} \varphi \Pi_{ss} \widehat{\Pi}_t - Y_{ss} C_{ss}^{-\chi_c} \theta w_{ss} \widehat{w}_t\end{aligned}$$

$$\begin{aligned}\text{RHS} &\approx g(Y_{ss}, C_{ss}, \Pi_{ss}) + g_{\Pi,ss} \Pi_{ss} \widehat{\Pi}_{t+1} \\ &= g(Y_{ss}, C_{ss}, \Pi_{ss}) + \beta Y_{ss} C_{ss}^{-\chi_c} \varphi \Pi_{ss} \widehat{\Pi}_{t+1}\end{aligned}$$

Using $f(Y_{ss}, C_{ss}, \Pi_{ss}, w_{ss}) = g(Y_{ss}, C_{ss}, \Pi_{ss})$, $\Pi_{ss} = 1$, $w_{ss} = \frac{(\theta-1)(1+\tau)}{\theta}$

and rearranging terms,

$$Y_{ss} C_{ss}^{-\chi_c} \varphi \Pi_{ss} \widehat{\Pi}_t - Y_{ss} C_{ss}^{-\chi_c} \theta w_{ss} \widehat{w}_t = Y_{ss} C_{ss}^{-\chi_c} \varphi \Pi_{ss} \widehat{\Pi}_{t+1}$$

$$\Leftrightarrow \varphi \widehat{\Pi}_t - \theta \frac{(\theta-1)(1+\tau)}{\theta} \widehat{w}_t = \beta \varphi \widehat{\Pi}_{t+1}$$

$$\Leftrightarrow \widehat{\Pi}_t = \frac{(\theta-1)(1+\tau)}{\varphi} \widehat{w}_t + \beta \widehat{\Pi}_{t+1}$$

Loglinearize the aggregate production function **around the zero-inflation steady state**:

$$Y_t = N_t$$

$$\text{LHS} = f(Y_t)$$

$$\text{RHS} = g(N_t)$$

$$f_Y = 1$$

$$g_N = 1$$

$$\begin{aligned}\text{LHS} &\approx f(Y_{ss}) + f_{Y,ss} Y_{ss} \hat{Y}_t \\ &= f(Y_{ss}) + f(Y_{ss}) \hat{Y}_t\end{aligned}$$

$$\begin{aligned}\text{RHS} &\approx g(N_{ss}) + g_{N,ss} N_{ss} \hat{N}_t \\ &= g(N_{ss}) + g(N_{ss}) \hat{N}_t\end{aligned}$$

Using $f(Y_{ss}) = g(N_{ss})$, we get

$$\hat{Y}_t = \hat{N}_t$$

Loglinearize the aggregate resource constraint around the zero-inflation steady state:

$$Y_t = C_t + \frac{\varphi}{2} [\Pi_t - 1]^2 Y_t$$

$$\text{LHS} = f(Y_t)$$

$$\text{RHS} = g(C_t, Y_t, \Pi_t)$$

$$f_Y = 1, \quad g_C = 1$$

$$g_Y = \frac{\varphi}{2} [\Pi_t - 1]^2$$

$$g_\Pi = \varphi [\Pi_t - 1] Y_t$$

Note that $g_{Y,ss} = g_{\Pi,ss} = 0$ because $\Pi_{ss} = 1$.

$$\begin{aligned}\text{LHS} &\approx f(Y_{ss}) + f_{Y,ss} Y_{ss} \hat{Y}_t \\ &= f(Y_{ss}) + Y_{ss} \hat{Y}_t\end{aligned}$$

$$\begin{aligned}\text{RHS} &\approx g(C_{ss}, Y_{ss}, \Pi_{ss}) + g_{C,ss} C_{ss} \hat{C}_t \\ &= g(C_{ss}, Y_{ss}, \Pi_{ss}) + C_{ss} \hat{C}_t\end{aligned}$$

Note $f(Y_{ss}) = g(C_{ss}, Y_{ss}, \Pi_{ss})$. Also, $Y_{ss} = C_{ss}$ because $\Pi_{ss} = 1$. We obtain

$$\hat{Y}_t = \hat{C}_t$$

Loglinearize the Taylor rule **around the zero-inflation steady state**:

$$R_t = \max \left[1, \frac{1}{\beta} \Pi_t^\phi \right]$$

$$\text{LHS} = f(R_t)$$

$$\text{RHS} = g(\Pi_t)$$

$$f_R = 1$$

$$g_\Pi = \phi \frac{1}{\beta} \Pi_t^{\phi-1}$$

$$\begin{aligned}
\text{LHS} &\approx f(R_{ss}) + f_{R,ss} R_{ss} \hat{R}_t \\
&= f(R_{ss}) + 1 * R_{ss} \hat{R}_t \\
&= f(R_{ss}) + f(R_{ss}) \hat{R}_t
\end{aligned}$$

$$\begin{aligned}
\text{RHS} &\approx g(\Pi_{ss}) + g_{\Pi,ss} \Pi_{ss} \hat{\Pi}_t \\
&= g(\Pi_{ss}) + \phi \frac{1}{\beta} \Pi_{ss}^{\Phi^{-1}} \Pi_{ss} \hat{\Pi}_t \\
&= g(\Pi_{ss}) + \phi \frac{1}{\beta} \Pi_{ss}^{\Phi} \hat{\Pi}_t \\
&= g(\Pi_{ss}) + g(\Pi_{ss}) \phi \hat{\Pi}_t
\end{aligned}$$

$$\hat{R}_t = \max \left[1 - \frac{1}{\beta}, \phi \hat{\Pi}_t \right]$$

So far, we have

$$\hat{C}_t = \hat{C}_{t+1} - \frac{1}{\chi_c} [\hat{R}_t - \hat{\Pi}_{t+1} + \hat{\delta}_t]$$

$$\hat{w}_t = \chi_c \hat{C}_t + \chi_n \hat{N}_t$$

$$\hat{\Pi}_t = \frac{(\theta - 1)(1 + \tau)}{\varphi} \hat{w}_t + \beta \hat{\Pi}_{t+1}$$

$$\hat{Y}_t = \hat{N}_t$$

$$\hat{Y}_t = \hat{C}_t$$

$$\hat{R}_t = \max \left[1 - \frac{1}{\beta}, \phi \hat{\Pi}_t \right]$$

Eliminating \hat{w}_t , \hat{N}_t , and \hat{C}_t , we obtain

$$\hat{Y}_t = \hat{Y}_{t+1} - \frac{1}{\chi_c} [\hat{R}_t - \hat{\Pi}_{t+1} + \hat{\delta}_t] \quad [\text{"IS Curve"}]$$

$$\hat{\Pi}_t = \frac{(\theta-1)(1+\tau)}{\varphi} (\chi_c + \chi_n) \hat{Y}_t + \beta \hat{\Pi}_{t+1} \quad [\text{"Phillips Curve"}]$$

$$\hat{R}_t = \max \left[1 - \frac{1}{\beta}, \phi \hat{\Pi}_t \right]$$

$$\hat{R}_t + r^* = \max \left[r^* + 1 - \frac{1}{\beta}, r^* + \phi \hat{\Pi}_t \right]$$

$$i_t = \max \left[\frac{1}{\beta} - 1 + 1 - \frac{1}{\beta}, r^* + \phi \hat{\Pi}_t \right]$$

$$i_t = \max [0, r^* + \phi \hat{\Pi}_t]$$

Change of notation: Using

$$y_t := \hat{Y}_t, \quad \pi_t := \hat{\Pi}_t, \quad i_t := \hat{R}_t + r^*$$
$$r^* := \frac{1}{\beta} - 1, \quad \sigma := \frac{1}{\chi_c}, \quad \kappa := \frac{(\theta-1)(1+\tau)}{\varphi} (\chi_c + \chi_n)$$

we obtain

$$y_t = y_{t+1} - \sigma[i_t - \pi_{t+1} - r^*]$$

$$\pi_t = \kappa y_t + \beta \pi_{t+1}$$

$$i_t = \max[0, r^* + \phi \pi_t]$$

Implications of the IS curve and the Phillips curve

1. Output today depends on the sum of future real rates.

$$\begin{aligned}y_t &= y_{t+1} - \sigma[i_t - \pi_{t+1} - r^*] \\&= [y_{t+2} - \sigma[i_{t+1} - \pi_{t+2} - r^*]] - \sigma[i_t - \pi_{t+1} - r^*] \\&= [[y_{t+3} - \sigma[i_{t+2} - \pi_{t+3} - r^*]] - \sigma[i_{t+1} - \pi_{t+2} - r^*]] - \\&\quad \sigma[i_t - \pi_{t+1} - r^*] \\&\dots \\&= -\sigma \sum_{k=0}^{\infty} [i_{t+k} - \pi_{t+k+1} - r^*]\end{aligned}$$

...using $y_\infty = y_{ss} = 0$

2. Inflation today depends on the sum of output (which is proportional to real wage---the marginal cost).

$$\begin{aligned}\pi_t &= \kappa y_t + \beta \pi_{t+1} \\&= \kappa y_t + \beta [\kappa y_{t+1} + \beta \pi_{t+2}] \\&= \kappa y_t + \beta \kappa y_{t+1} + \beta^2 [\kappa y_{t+2} + \beta \pi_{t+3}] \\&= \kappa y_t + \beta \kappa y_{t+1} + \beta^2 \kappa y_{t+2} + \dots \\&= \dots \\&= \kappa \sum_{k=0}^{\infty} \beta^k y_{t+k}\end{aligned}$$

...using $\pi_{\infty} = \pi_{ss} = 0$

Plan

(1) Log-linear Approximation of the Equilibrium Conditions

(2) Quadratic Approximation of the Household's Welfare

Take the second-order expansion of the household utility around the zero-inflation steady state.

$$U(C_t, N_t) \approx U_{ss} + U_{C,ss} C_{ss} \hat{C}_t + U_{N,ss} N_{ss} \hat{N}_t$$

$$+ \frac{1}{2} U_{CC,ss} C_{ss}^2 \hat{C}_t^2 + \frac{1}{2} U_{CN,ss} C_{ss} N_{ss} \hat{C}_t \hat{N}_t$$

$$+ \frac{1}{2} U_{NC,ss} N_{ss} C_{ss} \hat{N}_t \hat{C}_t + \frac{1}{2} U_{NN,ss} N_{ss}^2 \hat{N}_t^2$$

Useful stuff (I):

$$U_{C,ss} = C_t^{-\chi_c}, \quad U_{N,ss} = -N_t^{\chi_n}, \quad U_{CC,ss} = -\chi_c C_t^{-\chi_c - 1},$$

$$U_{CN,ss} = 0, \quad U_{NC,ss} = 0, \quad U_{NN,ss} = -\chi_n N_t^{\chi_n - 1}$$

$$\frac{U_{CC,ss}}{U_{C,ss}} = \frac{-\chi_c C_{ss}^{-\chi_c - 1}}{C_{ss}^{-\chi_c}} = -\chi_c C_{ss}^{-1}$$

$$\frac{U_{NN,ss}}{U_{N,ss}} = \frac{-\chi_n N_t^{\chi_n - 1}}{-N_t^{\chi_n}} = \chi_n N_{ss}^{-1}$$

Useful stuff (II): Second-order approximation of the aggregate resource constraint.

$$N_t = C_t + \frac{\varphi}{2}(\Pi_t - 1)^2 N_t$$

$$\text{LHS} = f(N_t)$$

$$\text{RHS} = g(C_t, N_t, \Pi_t)$$

$$f_N = 1, \quad f_{NN} = 0$$

$$g_C = 1, \quad g_N = \frac{\varphi}{2}(\Pi_t - 1)^2, \quad g_\Pi = \varphi(\Pi_t - 1)N_t$$

$$g_{CC} = 0, \quad g_{CN} = 0, \quad g_{CP} = 0$$

$$g_{NC} = 0, \quad g_{NN} = 0, \quad g_{NP} = \varphi(\Pi_t - 1)$$

$$g_{PC} = 0, \quad g_{PN} = \varphi(\Pi_t - 1), \quad g_{PP} = \varphi N_t$$

For g , at the steady state, the only non-zero coefficient is $g_C = 1$ and $g_{PP} = \varphi N_{ss}$ (because $\Pi_{ss} = 1$).

$$\begin{aligned}\text{LHS} &\approx f_{ss} + f_{N,ss} N_{ss} \widehat{N}_t + \frac{f_{N,ss}}{2} N_{ss}^2 \widehat{N}_t^2 \\ &= f_{ss} + N_{ss} \widehat{N}_t\end{aligned}$$

$$\begin{aligned}\text{RHS} &\approx g_{ss} + g_{C,ss} C_{ss} \widehat{C}_t + g_{N,ss} N_{ss} \widehat{N}_t + g_{\Pi,ss} \Pi_{ss} \widehat{\Pi}_t \\ &\quad + \frac{1}{2} [g_{CC,ss} C_{ss}^2 \widehat{C}_t^2 + g_{CN,ss} C_{ss} N_{ss} \widehat{C}_t \widehat{N}_t + g_{CP,ss} \Pi_{ss}^2 \widehat{\Pi}_t^2 \\ &\quad + g_{NC,ss} N_{ss} C_{ss} \widehat{N}_t \widehat{C}_t + g_{NN,ss} N_{ss}^2 \widehat{N}_t^2 + g_{NP,ss} N_{ss} \Pi_{ss} \widehat{N}_t \widehat{\Pi}_t \\ &\quad + g_{PC,ss} \Pi_{ss} C_{ss} \widehat{\Pi}_t \widehat{C}_t + g_{PN,ss} \Pi_{ss} N_{ss} \widehat{\Pi}_t \widehat{N}_t \\ &\quad + g_{PP,ss} \Pi_{ss}^2 \widehat{\Pi}_t^2]\end{aligned}$$

$$\begin{aligned}&= g_{ss} + g_{C,ss} C_{ss} \widehat{C}_t + \frac{1}{2} g_{PP,ss} \Pi_{ss}^2 \widehat{\Pi}_t^2 \\ &= g_{ss} + C_{ss} \widehat{C}_t + \frac{\varphi}{2} N_{ss} \widehat{\Pi}_t^2\end{aligned}$$

$$\Rightarrow \widehat{N}_t = \widehat{C}_t + \frac{\varphi}{2} \widehat{\Pi}_t^2$$

$$U_t - U_{ss} \approx U_{C,ss} C_{ss} \hat{C}_t + U_{N,ss} N_{ss} \hat{N}_t + \frac{1}{2} U_{CC,ss} C_{ss}^2 \hat{C}_t^2 + \frac{1}{2} U_{NN,ss} N_{ss}^2 \hat{N}_t^2$$

$$= U_{C,ss} C_{ss} \left(\hat{C}_t + \frac{1}{2} \frac{U_{CC,ss} C_{ss}^2}{U_{C,ss} C_{ss}} \hat{C}_t^2 \right) + U_{N,ss} N_{ss} \left(\hat{N}_t + \frac{1}{2} \frac{U_{NN,ss} N_{ss}^2}{U_{N,ss} N_{ss}} \hat{N}_t^2 \right)$$

$$= U_{C,ss} C_{ss} \left(\hat{C}_t - \frac{1}{2} \chi_c \hat{C}_t^2 \right) + U_{N,ss} N_{ss} \left(\hat{N}_t + \frac{1}{2} \chi_n \hat{N}_t^2 \right)$$

$$\frac{U_t - U_{ss}}{U_{C,ss} C_{ss}} \approx \left(\widehat{C}_t - \frac{\chi_c}{2} \widehat{C}_t^2 \right) + \frac{U_{N,ss} N_{ss}}{U_{C,ss} C_{ss}} \left(\widehat{N}_t + \frac{\chi_n}{2} \widehat{N}_t^2 \right)$$

$$= \left(\widehat{C}_t - \frac{\chi_c}{2} \widehat{C}_t^2 \right) - w_{ss} \left(\widehat{N}_t + \frac{\chi_n}{2} \widehat{N}_t^2 \right)$$

$$= \left(\widehat{C}_t - \frac{\chi_c}{2} \widehat{C}_t^2 \right) - \frac{(\theta-1)(1+\tau)}{\theta} \left(\widehat{N}_t + \frac{\chi_n}{2} \widehat{N}_t^2 \right)$$

$$= \left(\widehat{C}_t - \frac{\chi_c}{2} \widehat{C}_t^2 \right)$$

$$- \frac{(\theta-1)(1+\tau)}{\theta} \left(\widehat{C}_t + \frac{\varphi}{2} \widehat{\Pi}_t^2 + \frac{\chi_n}{2} \left[\widehat{C}_t + \frac{\varphi}{2} \widehat{\Pi}_t^2 \right]^2 \right)$$

$$\approx \left(\widehat{C}_t - \frac{\chi_c}{2} \widehat{C}_t^2 \right) - \frac{(\theta-1)(1+\tau)}{\theta} \left(\widehat{C}_t + \frac{\varphi}{2} \widehat{\Pi}_t^2 + \frac{\chi_n}{2} \widehat{C}_t^2 \right)$$

$$= \frac{\theta-(\theta-1)(1+\tau)}{\theta} \widehat{C}_t$$

$$- \left(\frac{\chi_c}{2} + \frac{(\theta-1)(1+\tau)}{\theta} \frac{\chi_n}{2} \right) \widehat{C}_t^2 - \frac{(\theta-1)(1+\tau)}{\theta} \frac{\varphi}{2} \widehat{\Pi}_t^2$$

$$\alpha \frac{\theta-(\theta-1)(1+\tau)}{\varphi(\theta-1)(1+\tau)} \widehat{C}_t - \frac{1}{2} \left[\widehat{\Pi}_t^2 + \frac{\theta\chi_c+(\theta-1)(1+\tau)\chi_n}{\varphi(\theta-1)(1+\tau)} \widehat{C}_t^2 \right]$$

$$= -\frac{1}{2} \left[\widehat{\Pi}_t^2 + \frac{\chi_c+\chi_n}{\varphi} \widehat{C}_t^2 \right] \quad (\text{if } \tau = \frac{1}{\theta-1})$$

Change of notation: Using

$$\begin{aligned} y_t &:= \widehat{Y}_t, & \pi_t &:= \widehat{\Pi}_t, & i_t &:= \widehat{R}_t + r^* \\ r^* &:= \frac{1}{\beta} - 1, & \sigma &:= \frac{1}{\chi_c}, & \kappa &:= \frac{\theta-1}{\varphi}(\chi_c + \chi_n) \end{aligned}$$

we obtain

$$\begin{aligned} &\frac{U_t - U_{ss}}{U_{C,ss} C_{ss}} \\ &\propto \frac{\theta - (\theta-1)(1+\tau)}{\varphi(\theta-1)(1+\tau)} y_t - \frac{1}{2} \left[\pi_t^2 + \frac{\theta \chi_c + (\theta-1)(1+\tau) \chi_n}{\varphi(\theta-1)(1+\tau)} y_t^2 \right] \\ &= -\frac{1}{2} \left[\pi_t^2 + \frac{\chi_c + \chi_n}{\varphi} y_t^2 \right] \quad (\text{if } \tau = \frac{1}{\theta-1}) \end{aligned}$$

$$u(\pi_t, y_t) := \lambda_y y_t - \frac{1}{2} [\pi_t^2 + \lambda y_t^2]$$

where

$$\lambda_y := \frac{\theta - (\theta - 1)(1 + \tau)}{\varphi(\theta - 1)(1 + \tau)}$$

$$\lambda := \frac{\theta \chi_c + (\theta - 1)(1 + \tau) \chi_n}{\varphi(\theta - 1)(1 + \tau)}$$

Note that, if $\tau = \frac{1}{\theta - 1}$,

$$\lambda_y := 0$$

$$\lambda := \frac{\chi_c + \chi_n}{\varphi}$$

$$\begin{aligned}
u(\pi_t, y_t) &\coloneqq \lambda_y y_t - \frac{1}{2} [\pi_t^2 + \lambda y_t^2] \\
&= -\frac{1}{2} [\pi_t^2 + \lambda y_t^2 - 2\lambda_y y_t] \\
&= -\frac{1}{2} \left[\pi_t^2 + \lambda y_t^2 - 2\lambda_y y_t + \lambda \left(\frac{\lambda_y}{\lambda} \right)^2 - \lambda \left(\frac{\lambda_y}{\lambda} \right)^2 \right] \\
&= -\frac{1}{2} \left[\pi_t^2 + \lambda y_t^2 - 2\lambda \frac{\lambda_y}{\lambda} y_t + \lambda \left(\frac{\lambda_y}{\lambda} \right)^2 - \lambda \left(\frac{\lambda_y}{\lambda} \right)^2 \right] \\
&= -\frac{1}{2} \left[\pi_t^2 + \lambda \left(y_t - \frac{\lambda_y}{\lambda} \right)^2 - \lambda \left(\frac{\lambda_y}{\lambda} \right)^2 \right] \\
&= -\frac{1}{2} \left[\pi_t^2 + \lambda \left(y_t - \frac{\lambda_y}{\lambda} \right)^2 \right] + \frac{\lambda}{2} \left(\frac{\lambda_y}{\lambda} \right)^2 \\
&= -\frac{1}{2} [\pi_t^2 + \lambda(y_t - y^*)^2] + (\text{t.i.p.})
\end{aligned}$$

t.i.p. stands for “terms independent of policy.”

With some abuse of notation, I will use “ u ” to denote the part of the utility flow that is independent of policy going forward. That is,

$$u(\pi_t, y_t) := -\frac{1}{2} [\pi_t^2 + \lambda(y_t - y^*)^2]$$

where

$$y^* := \frac{\lambda_y}{\lambda}$$

If $\tau < \frac{1}{\theta-1}$, we can show that

$$y^* > 0$$

- Thus, if $y_t = 0$, the output is too low relative to the first best.

If $\tau > \frac{1}{\theta-1}$, we can show that

$$y^* < 0$$

- Thus, if $y_t = 0$, the output is too high relative to the first best.

These implications are consistent with the steady-state analysis in the nonlinear model.

Taylor Rule Equilibrium

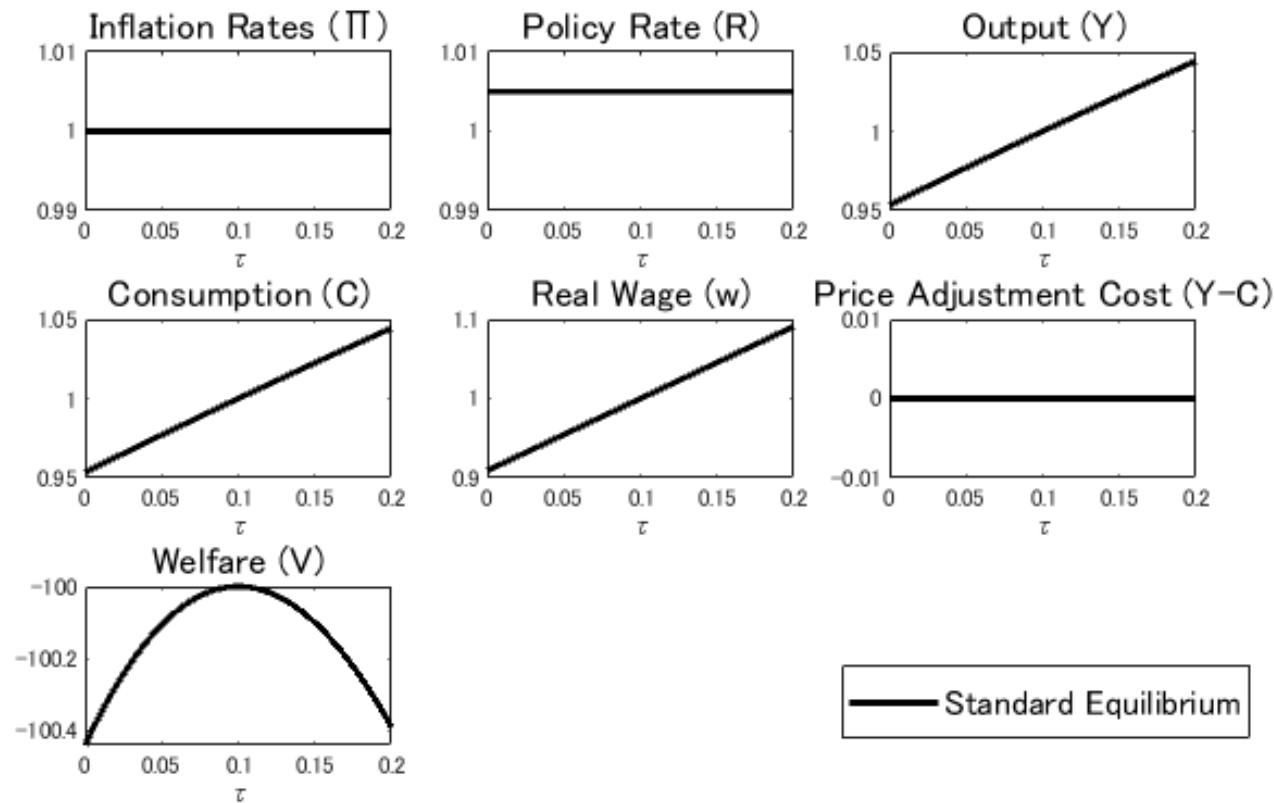


Fig. Standard Steady States with respect to τ ,
where $\{\beta, \theta, \chi_c, \chi_n\} = \{0.995, 11, 1, 1\}$

*From Slides_NK_SteadyStates.