

# Restoring Existence and Uniqueness at the Effective Lower Bound with Simple Fiscal Policy

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# Introduction

- ▶ Ascari and Mavroeidis (2022) (AM) : (non-)uniqueness/existence are challenge for NK models subject to ZLB.
  - \* Existence of unique solution is non-trivial.
- ▶ Several ways to ensure solution uniqueness
  - \* learning (Ascari, Mavroeidis, and McClung, 2023)
  - \* deviation from full information rational expectations (FIRE) (Angeletos and Lian, 2023)
  - \* **This paper:** passive fiscal policy

# This Paper

- ▶ Baseline three-equation NK model
  - \* Active Taylor rule
  - \* Rational expectations

# This Paper

- ▶ Baseline three-equation NK model
  - \* Active Taylor rule
  - \* Rational expectations
- ▶ Simple Ricardian fiscal policy restores solution uniqueness
- ▶ FP needs to satisfy two properties
  1. Sufficient counter-cyclicality.
  2. High persistence.

# Today

- ▶ NK Model
  - \* Non-uniqueness/existence of solution
- ▶ Adding fiscal policy
  - \* Perpetual change in spending
  - \* Inertial spending rule

# Equilibria in Baseline NK model

- ▶ Baseline (log-linear) NK model as in Galí (2015):

$$\text{IS: } \hat{y}_t = \mathbb{E}_t \hat{y}_{t+1} - \sigma^{-1} (\hat{i}_t - \mathbb{E}_t \hat{\pi}_{t+1}) + \varepsilon_t, \quad (1)$$

$$\text{NKPC: } \hat{\pi}_t = \beta \mathbb{E}_t \hat{\pi}_{t+1} + \kappa \hat{y}_t, \quad (2)$$

$$\text{TR: } \hat{i}_t = \max \{ -\mu, \phi_\pi \hat{\pi}_t \}, \quad (3)$$

# The Baseline NK model

- ▶  $\varepsilon_t$  materialises and persists w.p.  $p$  and vanishes w.p.  $1 - p$ :

$$\varepsilon_t = \begin{cases} \frac{p}{\sigma} \hat{r}^T & \text{w.p. } p \\ \mathbf{0} & \text{w.p. } 1 - p \end{cases} \quad (4)$$

- ▶ Model in transitory state is given by:

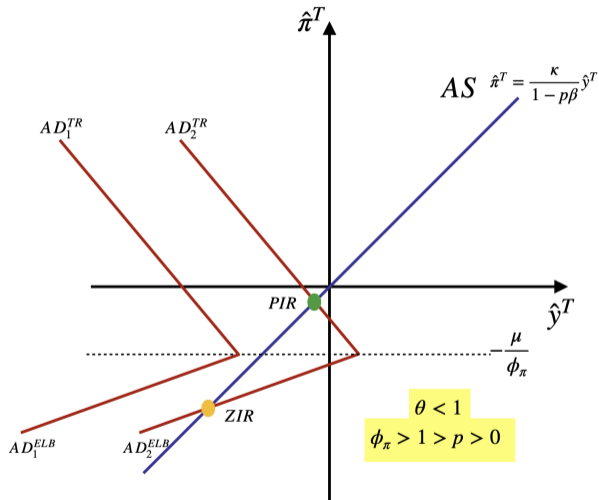
$$\hat{\pi}^T = \frac{\kappa}{1 - p\beta} \hat{y}^T \quad AS, \quad (5a)$$

$$\hat{\pi}^T = \begin{cases} \frac{\sigma(1-p)}{p-\phi_\pi} \hat{y}^T - \frac{p}{p-\phi_\pi} \hat{r}^T & AD^{TR} \text{ for } \hat{\pi}^T \geq -\frac{\mu}{\phi_\pi}, \\ \frac{\sigma(1-p)}{p} \hat{y}^T - \frac{\mu}{p} - \hat{r}^T & AD^{ELB} \text{ for } \hat{\pi}^T \leq -\frac{\mu}{\phi_\pi}. \end{cases} \quad (5b)$$

- ▶ Define slope ratio of  $AD^{ELB}$  and  $AS$ :

$$\theta = \frac{\sigma(1-p)(1-p\beta)}{p\kappa} \quad (6)$$

# Non-uniqueness/existence in NK Model ( $\theta < 1$ )



# Adding Fiscal Policy

- ▶ Now include fiscal policy; government spending is funded by output and lump-sum taxes

$$\tau_t = G_t$$

- ▶ Model now is:

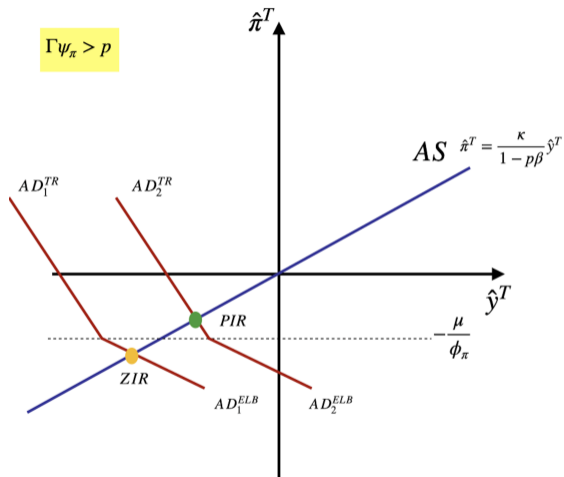
$$\hat{x}_t = \mathbb{E}_t \hat{x}_{t+1} - \frac{c}{\sigma} \left( \hat{i}_t - \mathbb{E}_t \hat{\pi}_{t+1} \right) - \Gamma \Delta \hat{g}_{t+1} + \varepsilon_t, \quad (7)$$

$$\hat{\pi}_t = \beta \mathbb{E}_t \hat{\pi}_{t+1} + \kappa_y \hat{x}_t \quad (8)$$

$$\hat{i}_t = \max\{-\mu; \phi_\pi \hat{\pi}_t\}, \quad (9)$$

$$\Delta \hat{g}_{t+1} = \psi_\pi \hat{\pi}_t + \psi_y \hat{x}_t \quad (10)$$

# Illustrating the NK-FP Model



# Key Intuition

- ▶ Fiscal policy changes the slope of  $AD^{ZLB}$
- ▶ Recall:  $AD^{ELB}/AS$  slope ratio:

$$\theta = \frac{\sigma(1-p)(1-p\beta)}{\kappa_y c(p - \Gamma\psi_\pi)} \quad (11)$$

- ▶  $\theta$  is determined by policy,  $\psi_\pi$ , and uncertainty,  $p$ .

# Endogenous State Solution

- ▶ Analytical solution with endogenous states is infeasible.
- ▶ Model dynamics depend on all past realisations of the shock and endogenous state.
- ▶ With no endogenous states, the support of  $\mathbf{Y}$  is time-invariant.

$$F(\mathbf{Y}) = \lambda(\mathbf{X}). \quad (12)$$

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- ▶ With endogenous states:

$$F(\mathbf{Y}_t) = \lambda(\mathbf{X}, \mathbf{Y}_{t-1}). \quad (13)$$

- ▶ To ensure uniqueness and existence, we use backward iterative algorithm.
- ▶ Goal: find a path from  $t = 0$  to some terminal  $T$  along which the mapping is invertible.

Sketch of methodology and algorithm

# Inertial gov. spending

- ▶ So far, we have considered “unit-root fiscal rules”.
- ▶ We extend analysis to inertial rules  $\implies$  endogenous state variables:

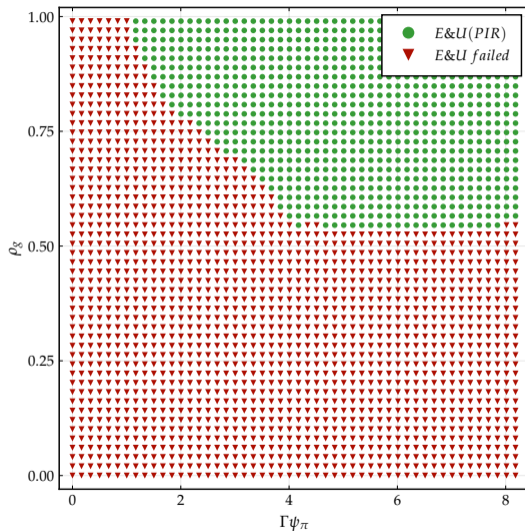
$$\hat{x}_t = \mathbb{E}_t \hat{x}_{t+1} - \frac{c}{\sigma} \left( \hat{i}_t - \mathbb{E}_t \hat{\pi}_{t+1} \right) - \Gamma \Delta \hat{g}_{t+1} + \varepsilon_t, \quad (14)$$

$$\hat{\pi}_t = \beta \mathbb{E}_t \hat{\pi}_{t+1} + \kappa_y \hat{x}_t, \quad (15)$$

$$\hat{i}_t = \max\{-\mu; \phi_\pi \hat{\pi}_t\}, \quad (16)$$

$$\hat{g}_t = \rho_g \hat{g}_{t-1} + \psi_\pi \hat{\pi}_t + \psi_y \hat{x}_t, \quad (17)$$

# Uniqueness with Inertial Rule



# Conclusion

- ▶ Ricardian fiscal policy can restore uniqueness in presence of ELB.
- ▶ FP needs to satisfy two properties: (i) sufficiently countercyclical, (ii) persistent.
- ▶ We develop operational algorithm to check uniqueness conditions with an endogenous state.

# Appendix

# Sketch of Methodology (Forward Looking) I

Ascari and Mavroeidis (2022)

Gourieroux, Laffont, and Monfort (1980)

$$\mathbf{o} = \underbrace{\mathbf{A}_{s_t} \mathbf{Y}_t + \mathbf{C}_{s_t} \mathbf{X}_t}_{\text{This paper}} + \mathbf{B}_{s_t} \mathbf{Y}_{t+1|t} + \mathbf{D}_{s_t} \mathbf{X}_{t+1|t} + \mathbf{H}_{s_t} \mathbf{Y}_{t-1}, \quad (18a)$$

$$s_t = \mathbb{1} \left\{ \mathbf{a}^\top \mathbf{Y}_t + \mathbf{c}^\top \mathbf{X}_t + \mathbf{b}^\top \mathbf{Y}_{t+1|t} + \mathbf{d}^\top \mathbf{X}_{t+1|t} + \mathbf{h}^\top \mathbf{Y}_{t-1} > 0 \right\}. \quad (18b)$$

- Our contribution: Extend (18) with endogenous states using GLM Theorem.

# Sketch of Methodology (Forward Looking) II

## Theorem (GLM Theorem 1)

A system of piecewise-linear equations satisfies coherency and completeness conditions if its matrix form defined in (18), represented as  $F(\mathbf{Y}) = \lambda(\mathbf{X})$ , where  $F(\mathbf{Y}) = \sum_j \mathcal{A}_j \mathbf{1}_{e_j} \text{vec}(\mathbf{Y})$ , is invertible; which requires that all determinants  $\mathcal{A}_j, j \in \{1, \dots, k\}$  share the same sign.

- ▶ Assume that shocks are two-state stationary Markovian

$$\mathbf{K} = \begin{pmatrix} p & 1-p \\ 1-q & q \end{pmatrix} \quad (19)$$

- ▶ Along an MSV solution, we have

$$\mathbb{E}_t[\mathbf{Y}_{t+1} | \mathbf{Y}_t = \mathbf{Y} \mathbf{e}_j] = \mathbb{E}_t[\mathbf{Y}_{t+1} | \mathbf{X}_t = \mathbf{X} \mathbf{e}_j] = \mathbf{Y} \mathbf{K}^\top \mathbf{e}_j \quad (20)$$

## Sketch of Methodology (Forward Looking) III

- ▶ This allows to rewrite the general form (with no endogenous states, i.e.  $\mathbf{h} = \mathbf{o}$ ) as

$$\begin{aligned}\mathbf{o} &= (\mathbf{A}_{s_i} \mathbf{Y} + \mathbf{B}_{s_i} \mathbf{Y} \mathbf{K}^\top + \mathbf{C}_{s_i} \mathbf{X} + \mathbf{D}_{s_i} \mathbf{X} \mathbf{K}^\top) \mathbf{e}_i, \\ s_i &= \mathbb{1}\{(\mathbf{a}^\top \mathbf{Y} + \mathbf{b}^\top \mathbf{Y} \mathbf{K}^\top + \mathbf{c}^\top \mathbf{X} + \mathbf{d}^\top \mathbf{X} \mathbf{K}^\top) \mathbf{e}_i > 0\}, \quad i = 1, \dots, k.\end{aligned}\tag{21}$$

- ▶ This can be rewritten as

$$F(\mathbf{Y}) = \sum_J \mathcal{A}_J \mathbf{1}_{\mathbf{e}_J} \text{vec}(\mathbf{Y}),\tag{22}$$

where  $\mathcal{C}_J = \{\mathbf{Y} : \mathbf{Y} \in \mathbb{R}^{n \times k}, s_i = \mathbf{1}_{\{i \in J\}}\}$

- ▶ (22) must be invertible for the system to have a unique solution.

# Sketch of Methodology (Forward Looking) IV

- ▶ This is the case if determinants of  $\mathcal{A}_J$  have the same sign:

$$\begin{aligned}\mathcal{A}_{J_1} &= \mathbf{A}_1 \mathbf{I}_2 + \mathbf{B}_1 \mathbf{K}, & J_1 &= \{1, 2\}, \\ \mathcal{A}_{J_2} &= \mathbf{e}_1 \mathbf{e}_1^\top \mathcal{A}_{J_4} + \mathbf{e}_2 \mathbf{e}_2^\top \mathcal{A}_{J_1}, & J_2 &= \{2\}, \\ \mathcal{A}_{J_3} &= \mathbf{e}_2 \mathbf{e}_2^\top \mathcal{A}_{J_4} + \mathbf{e}_1 \mathbf{e}_1^\top \mathcal{A}_{J_1}, & J_3 &= \{1\}, \\ \mathcal{A}_{J_4} &= \mathbf{A}_0 \mathbf{I}_2 + \mathbf{B}_0 \mathbf{K}, & J_4 &= \emptyset.\end{aligned}\tag{23}$$

Back

# Methodology with an Endogenous State I

- ▶ Consider the simple case of a single endogenous state. For a date  $T$  whereby  $t \geq T$ , an MSV solution  $f(y_{t-1}, \mathbf{X}_t)$  can be written as

$$\mathbf{Y}_t = \mathbf{G}y_{t-1} + \mathbf{Z},$$

- ▶ Again, assume  $k = 2$
- ▶ General form:

$$\begin{aligned} \mathbf{o} = & \left( \mathbf{A}_{S_t,i} \mathbf{G} \mathbf{e}_i + \mathbf{h}_{S_t,i} + \mathbf{B}_{S_t,i} \mathbf{G} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \mathbf{G} \mathbf{e}_i \right) y_{t-1} \\ & + \left( \mathbf{A}_{S_t,i} \mathbf{Z} + \mathbf{B}_{S_t,i} \mathbf{G} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \mathbf{Z} + \mathbf{B}_{S_t,i} \mathbf{Z} \mathbf{K}^\top + \mathbf{C}_{S_t,i} \mathbf{X} + \mathbf{D}_{S_t,i} \mathbf{X} \mathbf{K}^\top \right) \mathbf{e}_i, \end{aligned} \tag{24}$$

for all  $i = 1, \dots, k$ .

## Methodology with an Endogenous State II

- ▶ For a given regime  $J$  corresponding to the  $k$  states and their transitions, a slackness condition for the constraint  $s_{t,j}$  is determined which gives a system of  $2nk$  polynomial equations in the  $2nk$  unknowns  $\mathbf{G}$  and  $\mathbf{Z}$  by equating the coefficients on  $y_{t-1}$  and the constant terms to zero, respectively.
- ▶ As these conditions are polynomial and not piecewise linear in  $\mathbf{G}$  and  $\mathbf{Z}$ , the algorithm and theorem of Gourieroux, Laffont, and Monfort (1980) is no longer suitable to check coherency.
- ▶ Build a “brute force” algorithm which essentially goes through all possible  $2^k J$  regime configurations to check if there are any feasible solutions that satisfy the inequality constraints.

## Methodology with an Endogenous State III

- ▶ We know that at  $T$ , the solution to the model takes the following form

$$\mathbf{Y}_T = \mathbf{G}_{J_0} \mathbf{y}_{T-1} + \mathbf{Z}_{J_0},$$

where  $J_0 \in J$  defines the configuration of regimes in  $T$ .  $\mathbf{G}_{J_0}$  and  $\mathbf{Z}_{J_0}$  can be solved for from (24):

$$\mathbf{o} = \mathbf{A}_{S_{t,i}} \mathbf{G} \mathbf{e}_i + \mathbf{h}_{S_{t,i}} + \mathbf{B}_{S_{t,i}} \mathbf{G} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \mathbf{G} \mathbf{e}_i, \quad (25)$$

$$\mathbf{o} = \left( \mathbf{A}_{S_{t,1}} \mathbf{Z} + \mathbf{B}_{S_{t,i}} \mathbf{G} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \mathbf{Z} + \mathbf{B}_{S_{t,i}} \mathbf{Z} \mathbf{K}^\top + \mathbf{C}_{S_{t,i}} \mathbf{X} + \mathbf{D}_{S_{t,i}} \mathbf{X} \mathbf{K}^\top \right) \mathbf{e}_i, \quad (26)$$

$\forall i = 1, \dots, k.$

## Methodology with an Endogenous State IV

- ▶  $\mathbf{Y}_T$  is a function of  $\mathbf{G}_{J_0}$  and  $\mathbf{Z}_{J_0}$ . Thus,  $\mathbf{Y}_T$  is known and we can solve for  $\mathbf{Y}_{T-1}$  from

$$\mathbf{o} = \left( \mathbf{A}_{S_{T-1},i} + \mathbf{B}_{S_{T-1},i} \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \mathbf{Y}_{T-1} \mathbf{e}_i \right) \\ + \left( \mathbf{B}_{S_{T-1},i} \mathbf{Z}_{J_0} \mathbf{K}^\top + \mathbf{C}_{S_{T-1},i} \mathbf{X} + \mathbf{D}_{S_{T-1},i} \mathbf{X} \mathbf{K}^\top \right) \mathbf{e}_i + \mathbf{h}_{S_{T-1},i} y_{T-2}.$$

- ▶ For every  $t$  the determinants relevant for CC conditions are given by

$$|\mathcal{A}_{J_0 J_1}| = \prod_i^k \det \left( \mathbf{A}_{S_{T-1},i} + \mathbf{B}_{S_{T-1},i} \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \right).$$

# Methodology with an Endogenous State V

- ▶ If  $k = 2$ , the determinants can be rewritten as

$$|\mathcal{A}_{J_0\{1,2\}}| = \det \left( \mathbf{A}_1 + \mathbf{B}_1 \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_1 \mathbf{g}^\top \right) \det \left( \mathbf{A}_1 + \mathbf{B}_1 \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_2 \mathbf{g}^\top \right),$$

$$|\mathcal{A}_{J_0\{2\}}| = \det \left( \mathbf{A}_0 + \mathbf{B}_0 \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_1 \mathbf{g}^\top \right) \det \left( \mathbf{A}_1 + \mathbf{B}_1 \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_2 \mathbf{g}^\top \right),$$

$$|\mathcal{A}_{J_0\{1\}}| = \det \left( \mathbf{A}_0 + \mathbf{B}_0 \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_1 \mathbf{g}^\top \right) \det \left( \mathbf{A}_1 + \mathbf{B}_1 \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_2 \mathbf{g}^\top \right),$$

$$|\mathcal{A}_{J_0\{\emptyset\}}| = \det \left( \mathbf{A}_0 + \mathbf{B}_0 \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_1 \mathbf{g}^\top \right) \det \left( \mathbf{A}_0 + \mathbf{B}_0 \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_2 \mathbf{g}^\top \right).$$

# Methodology with an Endogenous State VI

- ▶ If the model satisfies CC, the solution is given by

$$\mathbf{Y}_{T-1} \mathbf{e}_i = - \left( \mathbf{A}_{S_{T-1},i} + \mathbf{B}_{S_{T-1},i} \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \right)^{-1} \left[ \left( \mathbf{B}_{S_{T-1},i} \mathbf{Z}_{J_0} \mathbf{K}^\top + \mathbf{C}_{S_{T-1},i} \mathbf{X} + \mathbf{D}_{S_{T-1},i} \mathbf{X} \mathbf{K}^\top \right) \mathbf{e}_i + \mathbf{h}_{S_{T-1},i} y_{T-2} \right],$$

$\forall i = 1, \dots, k.$

- ▶ Iterating the solution backwards implies that all the determinants  $\mathcal{A}_{J_0, \dots, J_{T-t}}$  must have the same sign. The recursive solution will be given by

$$\mathbf{Y}_t = \mathbf{G}_{J_0, \dots, J_{T-t}} y_{t-1} + \mathbf{Z}_{J_0, \dots, J_{T-t}},$$

## Methodology with an Endogenous State VII

where  $\mathbf{G}_{J_0, \dots, J_{T-t}}$  and  $\mathbf{Z}_{J_0, \dots, J_{T-t}}$  can be computed recursively using

$$\mathbf{Z}_{J_0, \dots, J_{T-t}, i} = - \left( \mathbf{A}_{S_{t,i}} + \mathbf{B}_{S_{t,i}} \mathbf{G}_{J_0, \dots, J_{T-t-1}} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \right)^{-1} \left( \mathbf{B}_{S_{t,i}} \mathbf{Z}_{J_0, \dots, J_{T-t-1}} \mathbf{K}^\top + \mathbf{C}_{S_{t,i}} \mathbf{X} + \mathbf{D}_{S_{t,i}} \mathbf{X} \mathbf{K}^\top \right) \mathbf{e}_i, \quad (27)$$

$$\mathbf{G}_{J_0, \dots, J_{T-t}, i} = - \left( \mathbf{A}_{S_{t,i}} + \mathbf{B}_{S_{t,i}} \mathbf{G}_{J_0, \dots, J_{T-t-1}} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \right)^{-1} \mathbf{h}_{S_{t,i}}. \quad (28)$$

- ▶ The recursive solution from terminal  $T$  solves the model backwards to  $t = 1$  and implies up to  $2^{(T-1)k}$  solution paths.
- ▶ Given some initial condition,  $\mathbf{y}_0$ , and conditional on satisfaction of CC conditions, the recursive solution is unique. If the CC conditions are not satisfied, there can be either no or multiple solutions. [Back](#)

# Inertial NK-FP Model I

- ▶ The MSV solution is given by

$$\mathbf{Y}_T = \mathbf{G}_{J_0} \mathbf{y}_{T-1} + \mathbf{Z}_{J_0},$$

- ▶ where  $J_0 \in J$  defines the configuration of regimes in  $T$
- ▶  $\mathbf{G}_{J_0}$  and  $\mathbf{Z}_{J_0}$  can be solved for from

$$\mathbf{o} = \mathbf{A}_{S_{t,i}} \mathbf{G} \mathbf{e}_i + \mathbf{h}_{S_{t,i}} + \mathbf{B}_{S_{t,i}} \mathbf{G} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \mathbf{G} \mathbf{e}_i, \quad (29)$$

$$\mathbf{o} = \left( \mathbf{A}_{S_{t,1}} \mathbf{Z} + \mathbf{B}_{S_{t,i}} \mathbf{G} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \mathbf{Z} + \mathbf{B}_{S_{t,i}} \mathbf{Z} \mathbf{K}^\top + \mathbf{C}_{S_{t,i}} \mathbf{X} + \mathbf{D}_{S_{t,i}} \mathbf{X} \mathbf{K}^\top \right) \mathbf{e}_i, \quad (30)$$

$$\forall i = 1, \dots, k.$$

# Inertial NK-FP Model II

- ▶ This yields  $\mathbf{Y}_T$ , we then can solve for  $\mathbf{Y}_{T-1}$

$$\begin{aligned} \mathbf{o} = & \left( \mathbf{A}_{S_{T-1},i} + \mathbf{B}_{S_{T-1},i} \mathbf{G}_{J_0} \mathbf{K}^\top \mathbf{e}_i \mathbf{g}^\top \mathbf{Y}_{T-1} \mathbf{e}_i \right) \\ & + \left( \mathbf{B}_{S_{T-1},i} \mathbf{Z}_{J_0} \mathbf{K}^\top + \mathbf{C}_{S_{T-1},i} \mathbf{X} + \mathbf{D}_{S_{T-1},i} \mathbf{X} \mathbf{K}^\top \right) \mathbf{e}_i + \mathbf{h}_{S_{T-1},i} y_{T-2}. \end{aligned} \quad (31)$$

- ▶ NK-IFP model:

$$\hat{\mathbf{x}}_t = \mathbb{E}_t \hat{\mathbf{x}}_{t+1} - \frac{c}{\sigma} \left( \hat{i}_t - \mathbb{E}_t \hat{\pi}_{t+1} - \hat{r}_t^n \right), \quad (32a)$$

$$\hat{\pi}_t = \beta \mathbb{E}_t \hat{\pi}_{t+1} + \kappa_y \hat{\mathbf{x}}_t - \kappa_g \widehat{\mathbf{GR}}_t, \quad (32b)$$

$$\hat{i}_t = \max\{-\mu; \phi_\pi \hat{\pi}_t + \phi_y \hat{\mathbf{x}}_t\}, \quad (32c)$$

$$\widehat{\mathbf{GR}}_t = \rho_\tau \widehat{\mathbf{GR}}_{t-1} + \psi_\pi \hat{\pi}_t + \psi_y \hat{\mathbf{x}}_t, \quad (32d)$$

# Inertial NK-FP Model III

with

$$\hat{r}_t^n = -\frac{\sigma}{c}(g\mathbb{E}_t\Delta\widehat{GR}_{t+1} + \varepsilon_t).$$

- ▶ Can be evaluated about two absorbing states, either PIR or ZIR.

- \* PIR:  $\{\hat{x}, \hat{\pi}, \hat{i}, \widehat{GR}\} = \{0, 0, 0, 0\}$

- \* ZIR  $\implies \hat{i} = -\mu$ :

$$\{\hat{x}, \hat{\pi}, \hat{i}, \widehat{GR}\} = \left\{ -\frac{(1-\beta)(1-\rho_\tau) + \kappa_g\psi_\pi}{\kappa_y(1-\rho_\tau) - \kappa_g\psi_y}\mu, -\mu, -\mu, \frac{\psi_y\hat{x} + \psi_\pi\hat{\pi}}{1-\rho_\tau} \right\}$$

- ▶ Under certain fiscal policy rules, the above ZIR equilibrium is not consistent with the constraint on the TR and, thus, the ZIR equilibrium is ruled out.

# Inertial NK-FP Model IV

- Specifically, we require that  $\hat{x}$  be sufficiently large such that the ZLB constraint on  $\hat{i}_t$  is not binding:

$$-\mu < -\phi_\pi \mu + \phi_y \hat{x} \implies -\frac{\mu(1 - \phi_\pi)}{\phi_y} < \hat{x}, \quad \phi_y \neq 0$$

- This implies

$$\frac{1 - \phi_\pi}{\phi_y} > \frac{(1 - \beta)(1 - \rho_\tau) + \kappa_g \psi_\pi}{\kappa_y(1 - \rho_\tau) - \kappa_g \psi_y},$$

with the LHS being negative under conventional restrictions on TR coefficients.

# Inertial NK-FP Model V

- ▶ As  $\rho_\tau$  tends to unity, we get

$$\lim_{\rho_\tau \rightarrow 1} \frac{(1 - \beta)(1 - \rho_\tau) + \kappa_g \psi_\pi}{\kappa_y(1 - \rho_\tau) - \kappa_g \psi_y} = -\frac{\psi_\pi}{\psi_y} < \frac{1 - \phi_\pi}{\phi_y},$$

which holds under countercyclical fiscal policy,  $\psi_y < 0$ ,  $\psi_\pi < 0$  with  $\psi_\pi$  sufficiently large in absolute value.

- ▶ Thus, under countercyclical fiscal policy, if  $\rho_\tau$  is sufficiently large there cannot exist a ZIR absorbing state.

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