



The Welfare of Ramsey Optimal Policy Facing Auto-Regressive Shocks

Jean-Bernard Chatelain, Kirsten Ralf

► To cite this version:

Jean-Bernard Chatelain, Kirsten Ralf. The Welfare of Ramsey Optimal Policy Facing Auto-Regressive Shocks. 2020. halshs-02564377v1

HAL Id: halshs-02564377

<https://shs.hal.science/halshs-02564377v1>

Preprint submitted on 5 May 2020 (v1), last revised 25 Jun 2020 (v2)

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.



WORKING PAPER N° 2020 – 23

The Welfare of Ramsey Optimal Policy Facing Auto-Regressive Shocks

**Jean-Bernard Chatelain
Kirsten Ralf**

JEL Codes: C61, C62, C73, E47, E52, E61, E63.

Keywords: Ramsey optimal policy, Stackelberg dynamic game, algorithm, forcing variables, augmented linear quadratic regulator, new-Keynesian Phillips curve



The Welfare of Ramsey Optimal Policy Facing Auto-Regressive Shocks

Jean-Bernard Chatelain* and Kirsten Ralf†

May 5, 2020

Abstract

With non-controllable auto-regressive shocks, the welfare of Ramsey optimal policy is the solution of a single Ricatti equation of a linear quadratic regulator. The existing theory refers to an additional Sylvester equation but miss another equation for computing the block matrix weighting the square of non-controllable variables in the welfare function. There is no need to simulate impulse response functions over a long period, to compute period loss functions and to sum their discounted value over this long period, as currently done so far. Welfare is computed for the case of the new-Keynesian Phillips curve with an auto-regressive cost-push shock.

JEL classification numbers: C61, C62, C73, E47, E52, E61, E63.

Keywords: Ramsey optimal policy, Stackelberg dynamic game, algorithm, forcing variables, augmented linear quadratic regulator, new-Keynesian Phillips curve.

*Paris School of Economics, Université Paris I Pantheon Sorbonne, PjSE, 48 Boulevard Jourdan, 75014 Paris. Email: jean-bernard.chatelain@univ-paris1.fr

†ESCE International Business School, INSEEC U. Research Center, 10 rue Sextius Michel, 75015 Paris, Email: Kirsten.Ralf@esce.fr.

1 Introduction

Dynamic stochastic general equilibrium (DSGE) models include auto-regressive shocks (Smets and Wouters (2007)). For computing the welfare of Ramsey optimal policy in DSGE models, one simulates impulse response functions over a long period, one computes period loss functions and one sums their discounted value over this long period.

Since Anderson et al. (1996), the available theory uses a Riccati equation for controllable variables and to a Sylvester equation for non-controllable variables in order to find the optimal policy rule and the optimal initial condition for non-predetermined variables. However, the matrix of the value function allowing to compute welfare is incomplete. Even worse, computing the loss function based on the two matrices solutions of the Riccati equation and Sylvester equation is often strictly negative, which is impossible. A third equation is missing in order to find the matrix related to the squares of the non-controllable variable in the value function.

We include in the Lagrangian the Lagrange multiplier times the dynamic equation of the non-controllable variables. This Lagrange multiplier is omitted in Anderson et al. (1996), p.202. Once this Lagrangian multiplier is included, the symmetry of the Hamiltonian matrix for the full system of controllable and non-controllable variables is restored. The value function is the solution of a Riccati equation for matrices related to controllable *and* non-controllable variables.

In Anderson et al. (1996), the Riccati equation only on controllable variables and the Sylvester equation only on non-controllable variables corresponds to two block matrix of the solution of our Riccati equation. The missing block matrix for computing welfare related to the square of non-controllable variables is found solving this Riccati equation.

This Riccati equation is coded in `lqr` instruction in SCILAB. We compute the welfare of Ramsey optimal policy for the new-Keynesian Phillips curve with an auto-regressive cost-push shock (Gali (2015)).

2 The Welfare of Ramsey optimal policy

To derive Ramsey optimal policy a Stackelberg leader-follower model is analyzed where the government is the leader and the private sector is the follower. Let \mathbf{k}_t be an $n_k \times 1$ vector of controllable predetermined state variables with initial conditions \mathbf{k}_0 given, \mathbf{x}_t an $n_x \times 1$ vector of non-predetermined endogenous variables free to jump at t without a given initial condition for \mathbf{x}_0 , put together in the $(n_k + n_x) \times 1$ vector $\mathbf{y}_t = (\mathbf{k}_t^T, \mathbf{x}_t^T)^T$. The $n_u \times 1$ vector \mathbf{u}_t denotes government policy instruments. We include an $n_z \times 1$ vector of non-controllable autoregressive shocks \mathbf{z}_t . All variables are expressed as absolute or proportional deviations from a steady state.

The policy maker maximizes the following quadratic function (minimizes the quadratic loss) subject to an initial condition for \mathbf{k}_0 and \mathbf{z}_0 , but not for \mathbf{x}_0 :

$$-\frac{1}{2} \sum_{t=0}^{+\infty} \beta^t (\mathbf{y}_t^T \mathbf{Q}_{yy} \mathbf{y}_t + 2\mathbf{y}_t^T \mathbf{Q}_{yz} \mathbf{z}_t + \mathbf{z}_t^T \mathbf{Q}_{zz} \mathbf{z}_t + \mathbf{u}_t^T \mathbf{R}_{uu} \mathbf{u}_t) \quad (1)$$

where β is the policy maker's discount factor. The policymaker's preferences are the relative weights included in the matrices \mathbf{Q} and \mathbf{R} . $\mathbf{Q}_{yy} \geq \mathbf{0}$ is a $(n_k + n_x) \times (n_k + n_x)$ positive symmetric semi-definite matrix, $\mathbf{R}_{uu} > \mathbf{0}$ is a $p \times p$ *strictly* positive symmetric definite matrix, so that the policy maker has at least a very small concern for the volatility

of policy instruments. The policy transmission mechanism of the private sector's behavior is summarized by this system of equations:

$$\begin{pmatrix} E_t \mathbf{y}_{t+1} \\ \mathbf{z}_{t+1} \end{pmatrix} = \begin{pmatrix} \mathbf{A}_{yy} & \mathbf{A}_{yz} \\ \mathbf{0} & \mathbf{A}_{zz} \end{pmatrix} \begin{pmatrix} \mathbf{y}_t \\ \mathbf{z}_t \end{pmatrix} + \begin{pmatrix} \mathbf{B}_{yu} \\ \mathbf{0} \end{pmatrix} \mathbf{u}_t, \quad (2)$$

where \mathbf{A} is an $(n_k + n_x + n_z) \times (n_k + n_x + n_z)$ matrix and \mathbf{B} is the $(n_k + n_x + n_z) \times p$ matrix of marginal effects of policy instruments \mathbf{u}_t on next period policy targets \mathbf{y}_{t+1} . The certainty equivalence principle of the linear quadratic regulator allows us to work with a non-stochastic model (Anderson et al. (1996)). Anderson et al. (1996) is word by word Hansen and Sargent (2007) chapter 5, so we refer only to Anderson et al. (1996) in what follows.

The government chooses sequences $\{\mathbf{u}_t, \mathbf{x}_t, \mathbf{k}_{t+1}\}_{t=0}^{+\infty}$ taking into account the policy transmission mechanism (2) and boundary conditions detailed below.

Essential boundary conditions are the initial conditions of predetermined variables \mathbf{k}_0 and \mathbf{z}_0 which are given. Natural boundary conditions are chosen by the policy maker to anchor the unique optimal initial values of the private sector's forward-looking variables. The policy maker's Lagrange multipliers of the private sector's forward (Lagrange multipliers) variables are *predetermined at the value zero*: $\frac{\partial L}{\partial \mathbf{x}_0} = \mu_{\mathbf{x}, t=0} = 0$ in order to determine the unique optimal initial value $\mathbf{x}_0 = \mathbf{x}_0^*$ of the private sector's forward variables.

Anderson et al. (1996) assume a bounded discounted quadratic loss function:

$$E \left(\sum_{t=0}^{+\infty} \beta^t (\mathbf{y}_t^T \mathbf{y}_t + \mathbf{z}_t^T \mathbf{z}_t + \mathbf{u}_t^T \mathbf{u}_t) \right) < +\infty \quad (3)$$

which implies

$$\begin{aligned} \lim_{t \rightarrow +\infty} \beta^t \mathbf{z}_t &= \mathbf{z}^* = \mathbf{0}, \mathbf{z}_t \text{ bounded,} \\ \lim_{t \rightarrow +\infty} \beta^t \mathbf{y}_t &= \mathbf{y}^* = \mathbf{0} \Leftrightarrow \lim_{t \rightarrow +\infty} \frac{\partial L}{\partial \mathbf{y}_t} = \mathbf{0} = \lim_{t \rightarrow +\infty} \beta^t \mu_t, \mu_t \text{ bounded.} \end{aligned}$$

This implies a stability criterion for eigenvalues of the dynamic system such that $|(\beta \lambda_i^2)^t| < |\beta \lambda_i^2| < 1$, so that stable eigenvalues are such that $|\lambda_i| < 1/\sqrt{\beta} < 1/\beta$. A preliminary step is to multiply matrices by $\sqrt{\beta}$ as follows $\sqrt{\beta} \mathbf{A}_{yy}$ $\sqrt{\beta} \mathbf{B}_y$ in order to apply formulas of Riccati equations for the non-discounted augmented linear quadratic regulator.

Assumption 1: The matrix pair $(\sqrt{\beta} \mathbf{A}_{yy} \sqrt{\beta} \mathbf{B}_{yu})$ is Kalman controllable if the controllability matrix has full rank:

$$\text{rank} \begin{pmatrix} \sqrt{\beta} \mathbf{B}_{yu} & \beta \mathbf{A}_{yy} \mathbf{B}_{yu} & \beta^{\frac{3}{2}} \mathbf{A}_{yy}^2 \mathbf{B}_{yu} & \dots & \beta^{\frac{n_k+n_x}{2}} \mathbf{A}_{yy}^{n_k+n_x-1} \mathbf{B}_{yu} \end{pmatrix} = n_k + n_x. \quad (4)$$

Assumption 2: The system is can be stabilized when the transition matrix \mathbf{A}_{zz} for the non-controllable auto-regressive variables has stable eigenvalues, such that $|\lambda_i| < 1/\sqrt{\beta}$.

The policy maker's choice can be solved with Lagrange multipliers. The Lagrangian includes not only the constraints of the private sector's policy transmission mechanisms multiplied by their respective Lagrange multipliers $2\beta^{t+1}\mu_{t+1}$, **BUT ALSO** the con-

straints of the non-controllable variables dynamics with their respective Lagrange multiplier $2\beta^{t+1}\nu_{t+1}$, which were omitted in Anderson et al. (1996), p.202.

$$-\frac{1}{2}\sum_{t=0}^{+\infty} \beta^t \left[\mathbf{y}_t^T \mathbf{Q}_{yy} \mathbf{y}_t + 2\mathbf{y}_t^T \mathbf{Q}_{yz} \mathbf{z}_t + \mathbf{z}_t^T \mathbf{Q}_{zz} \mathbf{z}_t + \mathbf{u}_t^T \mathbf{R}_{uu} \mathbf{u}_t \right] + \frac{2\beta^{t+1}\mu_{t+1} [\mathbf{A}_{yy}\mathbf{y}_t + \mathbf{A}_{yz}\mathbf{z}_t + \mathbf{B}_{yu}u_t - \mathbf{y}_{t+1}] + 2\beta^{t+1}\nu_{t+1} [\mathbf{A}_{zz}\mathbf{z}_t + \mathbf{0}_z u_t - \mathbf{z}_{t+1}]}{2\beta^{t+1}\nu_{t+1}} \quad (5)$$

The first order conditions are:

$$\begin{aligned} \frac{\partial L}{\partial x_t} &= \mathbf{R}\mathbf{x}_t + \beta\mathbf{B}\gamma_{t+1} = 0 \Rightarrow \mathbf{x}_t = -\beta\mathbf{R}^{-1}\mathbf{B}\gamma_{t+1} \\ \frac{\partial L}{\partial \pi_t} &= \mathbf{Q}\pi_t + \beta\mathbf{A}\gamma_{t+1} - \gamma_t = 0 \\ \frac{\partial L}{\partial z_t} &= \beta\gamma_{t+1}\mathbf{A}_{yz} + \beta\delta_{t+1}\mathbf{A}_{zz} - \delta_t = 0 \end{aligned}$$

The policy instrument are substituted by $\mathbf{x}_t = -\beta\mathbf{R}^{-1}\mathbf{B}\gamma_{t+1}$ in the transmission mechanism equation. The Hamiltonian of the linear quadratic regulator has the usual block matrices on left hand side and right hand side:

$$L = \begin{pmatrix} \mathbf{I} & -\beta\mathbf{B}_{(y,z)u}\mathbf{R}_{uu}^{-1}\mathbf{B}_{(y,z)u}^T \\ \mathbf{0} & \beta\mathbf{A}^T \end{pmatrix} \text{ and } N = \begin{pmatrix} \mathbf{A} & \mathbf{0} \\ -\mathbf{Q} & \mathbf{I} \end{pmatrix}$$

with this particular block decomposition between controllable variables \mathbf{y}_t and non-controllable variables \mathbf{z}_t :

$$\begin{pmatrix} \mathbf{I} & \mathbf{0} & -\beta\mathbf{B}_{yu}\mathbf{R}_{uu}^{-1}\mathbf{B}_{yu}^T & \mathbf{0} \\ \mathbf{0} & \mathbf{I} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \beta\mathbf{A}_{yy} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \beta\mathbf{A}_{yz} & \beta\mathbf{A}_{zz} \end{pmatrix} \begin{pmatrix} y_{t+1} \\ z_{t+1} \\ \mu_{t+1} \\ \nu_{t+1} \end{pmatrix} = \begin{pmatrix} \mathbf{A}_{yy} & \mathbf{A}_{yz} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{A}_{zz} & \mathbf{0} & \mathbf{0} \\ -\mathbf{Q}_{yy} & -\mathbf{Q}_{yz} & \mathbf{I} & \mathbf{0} \\ -\mathbf{Q}_{yz} & -\mathbf{Q}_{zz} & \mathbf{0} & \mathbf{I} \end{pmatrix} \begin{pmatrix} y_t \\ z_t \\ \mu_{t+1} \\ \nu_{t+1} \end{pmatrix}$$

The specificity of non-controllable variables is that the following matrix includes three blocks with zeros, which is not the case for controllable variables:

$$-\beta \begin{pmatrix} \mathbf{B}_{yu} \\ \mathbf{0} \end{pmatrix} (\mathbf{R}_{uu}^{-1}) \begin{pmatrix} \mathbf{B}_{yu} \\ \mathbf{0} \end{pmatrix}^T = \begin{pmatrix} -\beta\mathbf{B}_{yu}\mathbf{R}_{uu}^{-1}\mathbf{B}_{yu}^T & \mathbf{0} \\ \mathbf{0} & \mathbf{0} \end{pmatrix}$$

If \mathbf{L} is non-singular, the Hamiltonian matrix $\mathbf{H} = \mathbf{L}^{-1}\mathbf{N}$ is a symplectic matrix. With the equations of the Lagrange multipliers ν_{t+1} , all the roots ρ_i of A_{zz} have their mirror roots $(1/\beta\rho_i)$ which were all missing in Anderson et al. (1996).

The value function for welfare involve the matrix \mathbf{P} such that:

$$L_{t=0} = \begin{pmatrix} y_0 \\ z_0 \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{yy} & \mathbf{P}_{yz} \\ \mathbf{P}_{yz} & \mathbf{P}_{zz} \end{pmatrix} \begin{pmatrix} y_0 \\ z_0 \end{pmatrix}$$

A stabilizing solution of the Hamiltonian system satisfies (Anderson et al. (1996)):

$$\frac{\partial L}{\partial \mathbf{y}_{t=0}} = \mu_0 = \mathbf{P}_y \mathbf{y}_0 + \mathbf{P}_z \mathbf{z}_0. \quad (6)$$

The optimal rule of the augmented linear quadratic regulator is:

$$\mathbf{u}_t = \mathbf{F}_y \mathbf{y}_t + \mathbf{F}_z \mathbf{z}_t. \quad (7)$$

The matrix \mathbf{P} is solution of this Riccati equation:

$$\begin{aligned} \begin{pmatrix} \mathbf{P}_{yy} & \mathbf{P}_{yz} \\ \mathbf{P}_{yz} & \mathbf{P}_{zz} \end{pmatrix} &= \begin{pmatrix} \mathbf{Q}_{yy} & \mathbf{Q}_{yz} \\ \mathbf{Q}_{yz} & \mathbf{Q}_{zz} \end{pmatrix} + \beta \begin{pmatrix} \mathbf{A}_{yy} & \mathbf{A}_{yz} \\ \mathbf{0} & \mathbf{A}_{zz} \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{yy} & \mathbf{P}_{yz} \\ \mathbf{P}_{yz} & \mathbf{P}_{zz} \end{pmatrix} \begin{pmatrix} \mathbf{A}_{yy} & \mathbf{A}_{yz} \\ \mathbf{0} & \mathbf{A}_{zz} \end{pmatrix} \\ &\quad - \beta \begin{pmatrix} \mathbf{A}_{yy} & \mathbf{A}_{yz} \\ \mathbf{0} & \mathbf{A}_{zz} \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{yy} & \mathbf{P}_{yz} \\ \mathbf{P}_{yz} & \mathbf{P}_{zz} \end{pmatrix} \begin{pmatrix} \mathbf{B}_{yu} \\ \mathbf{0} \end{pmatrix} \\ &\quad \left(\mathbf{R}_{uu} + \beta \mathbf{B}'_{yu} \mathbf{P}_{yy} \mathbf{B}_{yu} \right)^{-1} \beta \begin{pmatrix} \mathbf{B}_{yu} \\ \mathbf{0} \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{yy} & \mathbf{P}_{yz} \\ \mathbf{P}_{yz} & \mathbf{P}_{zz} \end{pmatrix} \begin{pmatrix} \mathbf{A}_{yy} & \mathbf{A}_{yz} \\ \mathbf{0} & \mathbf{A}_{zz} \end{pmatrix} \end{aligned}$$

The matrix to be inverted in the Riccati equation is modified due to non-controllable variables:

$$\begin{pmatrix} \mathbf{B}_{yu} \\ \mathbf{0} \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{yy} & \mathbf{P}_{yz} \\ \mathbf{P}_{yz} & \mathbf{P}_{zz} \end{pmatrix} \begin{pmatrix} \mathbf{B}_{yu} \\ \mathbf{0} \end{pmatrix} = \mathbf{B}_{yu}^T \mathbf{P}_{yy} \mathbf{B}_{yu}$$

This Riccati equation is written as:

$$\begin{aligned} \begin{pmatrix} \mathbf{P}_{yy} & \mathbf{P}_{yz} \\ \mathbf{P}_{yz} & \mathbf{P}_{zz} \end{pmatrix} &= \begin{pmatrix} \mathbf{Q}_{yy} & \mathbf{Q}_{yz} \\ \mathbf{Q}_{yz} & \mathbf{Q}_{zz} \end{pmatrix} \\ &+ \beta \begin{pmatrix} \mathbf{A}_{yy}^T \mathbf{P}_{yy} \mathbf{A}_{yy} & \mathbf{A}_{yy}^T (\mathbf{P}_{yy} \mathbf{A}_{yz} + \mathbf{P}_{yz} \mathbf{A}_{zz}) \\ (\mathbf{A}_{yy}^T \mathbf{P}_{yy} \mathbf{A}_{yz} + \mathbf{A}_{yy}^T \mathbf{P}_{yz} \mathbf{A}_{zz})^T & \mathbf{A}_{yz}^T (\mathbf{P}_{yy} \mathbf{A}_{yz} + \mathbf{P}_{yz} \mathbf{A}_{zz}) + \mathbf{A}_{zz}^T (\mathbf{P}_{yz} \mathbf{A}_{yz} + \mathbf{P}_{zz} \mathbf{A}_{zz}) \end{pmatrix} \\ &- \beta^2 \begin{pmatrix} \mathbf{A}_{yy}^T \mathbf{P}_{yy} \mathbf{B}_{yu} & \mathbf{A}_{yz}^T \mathbf{P}_{yz} \mathbf{B}_{yu} \\ \mathbf{A}_{yz}^T \mathbf{P}_{yz} \mathbf{B}_{yu} + \mathbf{A}_{zz}^T \mathbf{P}_{zz} \mathbf{B}_{yu} \end{pmatrix} \left(\mathbf{R}_{uu} + \beta \mathbf{B}'_{yu} \mathbf{P}_{yy} \mathbf{B}_{yu} \right)^{-1} \\ &\left(\mathbf{B}_{yu}^T \mathbf{P}_{yy} \mathbf{A}_{yy} \quad \mathbf{B}_{yu}^T (\mathbf{P}_{yy} \mathbf{A}_{yz} + \mathbf{P}_{yz} \mathbf{A}_{zz}) \right) \end{aligned}$$

where \mathbf{P}_{yy} solves the matrix Riccati equation (Anderson et al. (1996)):

$$\mathbf{P}_{yy} = \mathbf{Q}_{yy} + \beta \mathbf{A}_{yy}^T \mathbf{P}_y \mathbf{A}_{yy} - \beta \mathbf{A}_{yy}^T \mathbf{P}_y \mathbf{B}_y \left(\mathbf{R}_{uu} + \beta \mathbf{B}_{yu}^T \mathbf{P}_{yy} \mathbf{B}_{yu} \right)^{-1} \beta \mathbf{B}_y^T \mathbf{P}_y \mathbf{A}_{yy},$$

where \mathbf{F}_y is computed knowing \mathbf{P}_y :

$$\mathbf{F}_y = - \left(\mathbf{R}_{uu} + \beta \mathbf{B}_{yu}^T \mathbf{P}_{yy} \mathbf{B}_{yu} \right)^{-1} \beta \mathbf{B}_y^T \mathbf{P}_y \mathbf{A}_{yy}, \quad (8)$$

where \mathbf{P}_{yz} solves the matrix Sylvester equation knowing \mathbf{P}_y and \mathbf{F}_y (Anderson et al. (1996)):

$$\mathbf{P}_{yz} = \mathbf{Q}_{yz} + \beta (\mathbf{A}_{yy} + \mathbf{B}_y \mathbf{F}_y)^T \mathbf{P}_y \mathbf{A}_{yz} + \beta (\mathbf{A}_{yy} + \mathbf{B}_y \mathbf{F}_y)^T \mathbf{P}_z \mathbf{A}_{zz}$$

where \mathbf{P}_{zz} , which is missing in Anderson et al. (1996), solves the matrix Sylvester equation knowing \mathbf{P}_y , \mathbf{F}_y and \mathbf{P}_{yz} :

$$\begin{aligned} \mathbf{P}_{zz} &= \mathbf{Q}_{zz} + \mathbf{A}_{yz}^T (\mathbf{P}_{yy} \mathbf{A}_{yz} + \mathbf{P}_{yz} \mathbf{A}_{zz}) + \mathbf{A}_{zz}^T (\mathbf{P}_{yz} \mathbf{A}_{yz} + \mathbf{P}_{zz} \mathbf{A}_{zz}) \\ &\quad - \beta^2 (\mathbf{A}_{yz}^T \mathbf{P}_{yy} \mathbf{B}_{yu} + \mathbf{A}_{zz}^T \mathbf{P}_{yz} \mathbf{B}_{yu}) \left(\mathbf{R}_{uu} + \beta \mathbf{B}_{yu}' \mathbf{P}_{yy} \mathbf{B}_{yu} \right)^{-1} \\ &\quad \mathbf{B}_{yu}^T (\mathbf{P}_{yy} \mathbf{A}_{yz} + \mathbf{P}_{yz} \mathbf{A}_{zz}) \end{aligned}$$

Now, at last, we know \mathbf{P}_{zz} so that we can compute the welfare of Ramsey optimal policy:

Proposition 1 *The welfare of Ramsey optimal policy is:*

$$- \begin{pmatrix} \mathbf{k}_0 \\ \mathbf{z}_0 \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{kk} - \mathbf{P}_{kk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} & \mathbf{P}_{kz} - \mathbf{P}_{kx} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \\ \mathbf{P}_{zx} - \mathbf{P}_{zk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} & \mathbf{P}_{zz} - \mathbf{P}_{zk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \end{pmatrix} \begin{pmatrix} \mathbf{k}_0 \\ \mathbf{z}_0 \end{pmatrix}$$

Proof. Welfare is a function of controllable non-predetermined variables \mathbf{x}_0 , controllable predetermined variables \mathbf{k}_0 and non controllable predetermined auto-regressive shocks \mathbf{z}_0 :

$$\begin{pmatrix} \mathbf{x}_0 \\ \mathbf{k}_0 \\ \mathbf{z}_0 \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{xx} & \mathbf{P}_{xk} & \mathbf{P}_{xz} \\ \mathbf{P}_{kx} & \mathbf{P}_{kk} & \mathbf{P}_{kz} \\ \mathbf{P}_{zx} & \mathbf{P}_{zk} & \mathbf{P}_{zz} \end{pmatrix} \begin{pmatrix} \mathbf{x}_0 \\ \mathbf{k}_0 \\ \mathbf{z}_0 \end{pmatrix}$$

Ramsey optimal initial anchor of non-predetermined variables \mathbf{x}_0 is (Ljungqvist L. and Sargent T.J. (2012), chapter 19):

$$\frac{\partial L}{\partial \mathbf{x}_0} = \mathbf{P}_{xk} \mathbf{k}_0 + \mathbf{P}_{xx} \mathbf{x}_0 + \mathbf{P}_{xz} \mathbf{z}_0 = \mathbf{0} \Rightarrow \mathbf{x}_0 = \mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} \mathbf{k}_0 + \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \mathbf{z}_0$$

Hence, the welfare matrix of Ramsey optimal policy is:

$$\begin{aligned} & \begin{pmatrix} \mathbf{0} & -\mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} & -\mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \\ \mathbf{0} & \mathbf{I} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{xx} & \mathbf{P}_{xk} & \mathbf{P}_{xz} \\ \mathbf{P}_{kx} & \mathbf{P}_{kk} & \mathbf{P}_{kz} \\ \mathbf{P}_{zx} & \mathbf{P}_{zk} & \mathbf{P}_{zz} \end{pmatrix} \begin{pmatrix} \mathbf{0} & -\mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} & -\mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \\ \mathbf{0} & \mathbf{I} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{I} \end{pmatrix} \\ &= \begin{pmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} \\ -(\mathbf{P}_{xx}^{-1} \mathbf{P}_{xk})^T & \mathbf{1} & \mathbf{0} \\ -(\mathbf{P}_{xx}^{-1} \mathbf{P}_{xz})^T & \mathbf{0} & \mathbf{1} \end{pmatrix} \begin{pmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{P}_{kk} - \mathbf{P}_{kk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} & \mathbf{P}_{kz} - \mathbf{P}_{kx} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \\ \mathbf{0} & \mathbf{P}_{zx} - \mathbf{P}_{zk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} & \mathbf{P}_{zz} - \mathbf{P}_{zk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \end{pmatrix} \\ &= \begin{pmatrix} \mathbf{0} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{P}_{kk} - \mathbf{P}_{kk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} & \mathbf{P}_{kz} - \mathbf{P}_{kx} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \\ \mathbf{0} & \mathbf{P}_{zx} - \mathbf{P}_{zk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xk} & \mathbf{P}_{zz} - \mathbf{P}_{zk} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \end{pmatrix} \end{aligned}$$

■

3 New Keynesian Phillips Curve Example

The new-Keynesian Phillips curve constitutes the monetary policy transmission mechanism:

$$\pi_t = \beta E_t [\pi_{t+1}] + \kappa x_t + z_t \text{ where } \kappa > 0, 0 < \beta < 1,$$

where x_t represents the output gap, i.e. the deviation between (log) output and its efficient level. π_t denotes the rate of inflation between periods $t-1$ and t and plays the

role of the vector of forward-looking variables \mathbf{x}_t in the above general case. β denotes the discount factor. E_t denotes the expectation operator. The cost push shock z_t includes an exogenous auto-regressive component:

$$z_t = \rho z_{t-1} + \varepsilon_t \text{ where } 0 < \rho < 1 \text{ and } \varepsilon_t \text{ i.i.d. normal } N(0, \sigma_\varepsilon^2),$$

where ρ denotes the auto-correlation parameter and ε_t is identically and independently distributed (i.i.d.) following a normal distribution with constant variance σ_ε^2 . The welfare loss function is such that the policy target is inflation and the policy instrument is the output gap (Gali (2015), chapter 5):

$$\begin{aligned} \max -\frac{1}{2} E_0 \sum_{t=0}^{t=+\infty} \beta^t \left(\pi_t^2 + \frac{\kappa}{\varepsilon} x_t^2 \right) \\ \begin{pmatrix} E_t \pi_{t+1} \\ z_{t+1} \end{pmatrix} = \mathbf{A} \begin{pmatrix} \pi_t \\ z_t \end{pmatrix} + \mathbf{B} x_t + \begin{pmatrix} 0_y \\ 1 \end{pmatrix} \varepsilon_t \end{aligned}$$

There is one controllable non-predetermined variable: $\mathbf{x}_t = \pi_t$. There is no controllable predetermined variable ($\mathbf{k}_t = \mathbf{0}$). Gali's (2015) calibration is:

$$\begin{aligned} \sqrt{\beta} \mathbf{A} = \sqrt{0.99} \begin{pmatrix} \mathbf{A}_{xx} = -\frac{1}{\beta} = \frac{1}{0.99} & \mathbf{A}_{xz} = -\frac{1}{\beta} = -\frac{1}{0.99} \\ 0 & \mathbf{A}_{zz} = \rho = 0.8 \end{pmatrix}, \\ \sqrt{\beta} \mathbf{B} = \sqrt{0.99} \begin{pmatrix} \mathbf{B}_x = -\frac{\kappa}{\beta} = -\frac{0.1275}{0.99} \\ \mathbf{B}_z = 0 \end{pmatrix}, \mathbf{Q} = \begin{pmatrix} \mathbf{Q}_{xx} = 1 & \mathbf{Q}_{xz} = 0 \\ \mathbf{Q}_{xz} = 0 & \mathbf{Q}_{zz} = 0 \end{pmatrix}, \mathbf{R} = \frac{\kappa}{\varepsilon} = \frac{0.1275}{6} \end{aligned}$$

One multiplies matrices by $\sqrt{\beta}$ in order to take the discount factor in the Riccati equation. The welfare matrix is:

$$\mathbf{P} = \begin{pmatrix} \mathbf{P}_{xx} & \mathbf{P}_{xz} \\ \mathbf{P}_{xz} & \mathbf{P}_{zz} \end{pmatrix} = \begin{pmatrix} 1.7518055 & -1.1389181 \\ -1.1389181 & 3.4285107 \end{pmatrix}$$

Taking into account the optimal initial anchor of inflation ($\pi_0 = 0.65$ for $z_0 = 1$), the welfare matrix is:

$$\begin{pmatrix} 0 & -\mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \\ 0 & 1 \end{pmatrix}^T \begin{pmatrix} \mathbf{P}_{xx} & \mathbf{P}_{xz} \\ \mathbf{P}_{xz} & \mathbf{P}_{zz} \end{pmatrix} \begin{pmatrix} 0 & -\mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} = 0.6504 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & \mathbf{P}_{zz} - \mathbf{P}_{xz} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz} \end{pmatrix}$$

The welfare losses of Gali (2015) impulse response functions with Ramsey optimal initial condition: $\pi_0 = 0.65$ for $z_0 = 1$ is:

$$W = -(\mathbf{P}_{zz} - \mathbf{P}_{xz} \mathbf{P}_{xx}^{-1} \mathbf{P}_{xz}) z_0^2 = -2.688 \cdot z_0^2$$

We found the same value simulating impulse response functions over two hundred periods, computing period loss function and a discounted sum of these period loss functions over two hundred periods. Additional results on this example can be found in Chatelain and Ralf (2019).

Using only the information available in Anderson et al (1996), e.g. assuming $P_{zz} = 0$ for the missing block matrix in the value function, this welfare losses would be strictly positive $P_{xz} P_{xx}^{-1} P_{xz} = 0.74 > -2.688$, which is impossible.

References

- [1] Chatelain J.B. and Ralf K. (2019). "A simple algorithm for solving Ramsey optimal policy with exogenous forcing variables", *Economics Bulletin*, 39(4), 2429-2440.
- [2] Gali J. (2015). *Monetary Policy, Inflation and the Business Cycle*. 2nd edition, Princeton University Press, Princeton.
- [3] Hansen, L.P., and T.J. Sargent (2007). *Robustness*, Princeton University Press, New Jersey
- [4] Ljungqvist L. and Sargent T.J. (2012). *Recursive Macroeconomic Theory*. 3rd edition. The MIT Press. Cambridge, Massachussets.
- [5] Smets, F., and Wouters, R. (2007). Shocks and frictions in US business cycles: A Bayesian DSGE approach. *American Economic Review*, 97(3), 586-606.

4 Appendix

The numerical solution of the welfare matrix is obtained using Scilab code:

```
beta1=0.99; eps=6; kappa=0.1275; rho=0.8;
Qpi=1; Qz=0 ; Qzpi=0; R=kappa/eps;
A1=[1/beta1 -1/beta1 ; 0 rho] ;
A=sqrt(beta1)*A1;
B1=[-kappa/beta1 ; 0];
B=sqrt(beta1)*B1;
Q=[Qpi Qzpi ;Qzpi Qz ];
Big=sysdiag(Q,R);
[w,wp]=fullrf(Big);
C1=wp(:,1:2);
D12=wp(:,3:$);
M=syslin('d',A,B,C1,D12);
[Fy,Py]=lqr(M);
Py
Py(2,2)-Py(1,2)*inv(Py(1,1))*Py(1,2)
```