These notes loosely follow Fernández-Villaverde and Levintal (2018) "Solution Methods for Models with Rare Disasters" but omits several features, such as recursive preferences and disaster risk.

### 1 Model

#### 1.1 Household

The model admits a representative agent, so I directly write households' problem as the representative agent's. The representative household solves, in the cashless limit,

$$\max_{C_t, L_t, B_t, X_t, K_t} \mathbb{E}_0 \sum_{t=0}^{\infty} \beta^t \exp(\eta_{\beta, t}) \left( \frac{C_t^{1-\gamma}}{1-\gamma} - \varphi \exp(\eta_{L, t}) \frac{L_t^{1+\nu}}{1+\nu} \right)$$
 (1)

subject to the budget constraint

$$C_t + \frac{B_t}{P_t} + X_t \le W_t L_t + R_{K,t} K_{t-1} + R_{t-1} \frac{B_{t-1}}{P_t} + F_t + T_t, \tag{2}$$

where  $C_t$  is consumption,  $B_{t-1}$  nominal bonds,  $X_t$  investment,  $W_{N,t}$  the real wage,  $L_t$  labor,  $R_{KN,t}$  the gross real rental rate on capital,  $K_{t-1}$  capital,  $R_t$  the gross nominal interest rate on bonds,  $F_t$  real profits from firms, and  $T_t$  real lump-sum transfers from the government. The price of the final consumption good is  $P_t$ . Markets are assumed complete, but securities are in zero net supply. Because there is a representative agent, I may omit the Arrow securities from the budget constraint. My notation treats  $B_{t-1}$  and  $K_{t-1}$  as the stocks of bonds and capital present at time t, while  $B_t$  and  $K_t$  are the chosen stocks of bonds and capital for the following period. I adopt this notation so that all time t choices are dated at time t rather than having to differentiate between the predetermined time-t variables from the endogenous controls.

Investment for capital follows the law of motion

$$K_{t} = (1 - \delta)K_{t-1} + \Phi\left(\frac{X_{t}}{K_{t-1}}\right)K_{t-1}.$$
(3)

The Lagrangian for the household is

$$\mathcal{L} = \mathbb{E}_{0} \sum_{t=0}^{\infty} \beta^{t} \exp(\eta_{\beta,t}) \left[ \frac{C_{t}^{1-\gamma}}{1-\gamma} - \varphi \exp(\eta_{L,t}) \frac{L_{t}^{1+\nu}}{1+\nu} \right]$$

$$+ \mathbb{E}_{0} \sum_{t=0}^{\infty} \beta^{t} \exp(\eta_{\beta,t}) \lambda_{t} \left[ W_{t} L_{t} + R_{K,t} K_{t-1} + R_{t-1} \frac{B_{t-1}}{P_{t}} + F_{t} + T_{t} - C_{t} - \frac{B_{t}}{P_{t}} - X_{t} \right]$$

$$+ \mathbb{E}_{0} \sum_{t=0}^{\infty} \beta^{t} \exp(\eta_{\beta,t}) \lambda_{t} Q_{t} \left[ (1-\delta) K_{t-1} + \Phi \left( \frac{X_{t}}{K_{t-1}} \right) K_{t-1} - K_{t} \right],$$

which implies first-order conditions

$$0 = C_t^{-\gamma} - \lambda_t$$

$$0 = -\varphi \exp(\eta_{L,t}) L_t^{\nu} + \lambda_t W_t$$

$$0 = -\exp(\eta_{\beta,t}) \lambda_t + \beta \mathbb{E}_t [\exp(\eta_{\beta,t+1}) \lambda_{t+1} R_t]$$

$$0 = -\exp(\eta_{\beta,t}) \lambda_t + \exp(\eta_{\beta,t}) \lambda_t Q_t \Phi' \left(\frac{X_t}{K_{t-1}}\right) K_{t-1} \frac{1}{K_{t-1}}$$

$$0 = -\exp(\eta_{\beta,t}) \lambda_t Q_t + \beta \mathbb{E}_t \left[\exp(\eta_{\beta,t+1}) \lambda_{t+1} R_{K,t+1}\right]$$

$$\beta \mathbb{E}_t \left[\exp(\eta_{\beta,t+1}) \lambda_{t+1} Q_{t+1} \left(1 - \delta + \Phi' \left(\frac{X_{t+1}}{K_t}\right) K_t \left(-\frac{X_{t+1}}{K_t^2}\right) + \Phi \left(\frac{X_{t+1}}{K_t}\right)\right)\right]$$

The first two equations can be combined by isolating  $\lambda_t$ , which obtains the intratemporal consumption-labor condition

$$C_t^{-\gamma}W_t = \varphi \exp(\eta_{L,t})L_t^{\nu},$$

Using  $\lambda_t = C_t^{-\gamma}$  and defining the gross inflation rate  $\Pi_t \equiv P_t/P_{t-1}$ , I can obtain the Euler equation for households

$$\exp(\eta_{\beta,t}) \frac{C_t^{-\gamma}}{P_t} = \beta \mathbb{E}_t \left[ \exp(\eta_{\beta,t+1}) \frac{C_{t+1}^{-\gamma}}{P_{t+1}} R_t \right]$$
$$1 = \beta \mathbb{E}_t \left[ \frac{\exp(\eta_{\beta,t+1})}{\exp(\eta_{\beta,t})} \frac{C_{t+1}^{-\gamma}}{C_t^{-\gamma}} \frac{R_t}{\Pi_{t+1}} \right].$$

I can further simplify the Euler equation by defining the (real) stochastic discount factor

$$M_{t+1} = \beta \frac{\exp(\eta_{\beta,t+1})}{\exp(\eta_{\beta,t})} \frac{C_{t+1}^{-\gamma}}{C_t^{-\gamma}}$$

After dividing through by  $\exp(\eta_{\beta,t})\lambda_t$  and re-arranging, the investment condition becomes

$$1 = Q_t \Phi' \left( \frac{X_t}{K_{t-1}} \right).$$

Finally, after dividing through by  $\exp(\eta_{\beta,t})\lambda_t$ , the first-order condition for next-period capital is

$$Q_t = \mathbb{E}_t \left[ M_{t+1} \left( R_{K,t+1} + Q_{t+1} \left( 1 - \delta + \Phi \left( \frac{X_{t+1}}{K_t} \right) - \Phi' \left( \frac{X_{t+1}}{K_t} \right) \frac{X_{t+1}}{K_t} \right) \right].$$

In summary, households' optimality conditions are

$$C_t^{-\gamma} W_t = \varphi \exp(\eta_{L,t}) L_t^{\nu},\tag{4}$$

$$M_{t+1} = \beta \frac{\exp(\eta_{\beta,t+1})}{\exp(\eta_{\beta,t})} \frac{C_{t+1}^{-\gamma}}{C_t^{-\gamma}},\tag{5}$$

$$1 = \beta \mathbb{E}_t \left[ M_{t+1} \frac{R_t}{\Pi_{t+1}} \right], \tag{6}$$

$$1 = Q_t \Phi'\left(\frac{X_t}{K_{t-1}}\right),\tag{7}$$

$$Q_{t} = \mathbb{E}_{t} \left[ M_{t+1} \left( R_{K,t+1} + Q_{t+1} \left( 1 - \delta + \Phi \left( \frac{X_{t+1}}{K_{t}} \right) - \Phi' \left( \frac{X_{t+1}}{K_{t}} \right) \frac{X_{t+1}}{K_{t}} \right) \right]. \tag{8}$$

### 1.2 Production

**Final Producers** There is a representative final goods firm which sells consumption goods in a competitive market. It aggregates intermediate goods using the CES technology

$$Y_t = \left(\int_0^1 Y_t(j)^{\frac{\epsilon - 1}{\epsilon}}\right)^{\frac{\epsilon}{\epsilon - 1}}$$

where  $\epsilon > 1$  so that inputs are substitutes. Profit maximization for the final good firm is

$$\max_{Y_t(j)} P_t \left( \int_0^1 Y_t(j)^{\frac{\epsilon-1}{\epsilon}} \right)^{\frac{\epsilon}{\epsilon-1}} - \int_0^1 P_t(j) Y_t(j) \, dj.$$

The FOC for  $Y_t(j)$  is

$$0 = P_t \frac{\epsilon}{\epsilon - 1} \left( \int_0^1 Y_t(j)^{\frac{\epsilon}{\epsilon - 1}} \right)^{\frac{1}{\epsilon - 1}} \frac{\epsilon - 1}{\epsilon} Y_t(j)^{-\frac{1}{\epsilon}} - P_t(j)$$

$$0 = \left( \int_0^1 Y_t(j)^{\frac{\epsilon}{\epsilon - 1}} \right)^{\frac{1}{\epsilon - 1}} Y_t(j)^{-\frac{1}{\epsilon}} - \frac{P_t(j)}{P_t}$$

$$0 = \left( \int_0^1 Y_t(j)^{\frac{\epsilon}{\epsilon - 1}} \right)^{-\frac{\epsilon}{\epsilon - 1}} Y_t(j) - \left( \frac{P_t(j)}{P_t} \right)^{-\epsilon}$$

$$Y_t(j) = \left( \frac{P_t(j)}{P_t} \right)^{-\epsilon} Y_t.$$

Plugging this quantity into the identity

$$P_t Y_t = \int_0^1 P_t(j) Y_t(j) \, dj$$

and simplifying yields the price index

$$P_t = \left(\int_0^1 P_t(j)^{1-\epsilon} dj\right)^{\frac{1}{1-\epsilon}}.$$

**Intermediate Producers** Intermediate goods are producing according to the Cobb-Douglas technology

$$Y_t(j) = \exp(\eta_{A,t}) K_{t-1}^{\alpha}(j) L_t^{1-\alpha}(j).$$

Intermediate producers minimize cost subject to the constraint of meeting demand and Calvo price rigidities. Formally,

$$\min_{K_{t-1}(j), L_t(j)} R_{K,t} K_{t-1}(j) + W_t L_t(j) \qquad \text{s.t.} \qquad \exp(\eta_{A,t}) K_{t-1}^{\alpha}(j) L_t^{1-\alpha}(j) \ge \left(\frac{P_t(j)}{P_t}\right)^{-\epsilon} Y_t.$$

The RHS of the inequality constraint is the demand from final goods producers for intermediate j. The Lagrangian is

$$\mathcal{L} = R_{K,t}K_{t-1}(j) + W_tL_t(j) + MC_t(j) \left( \left( \frac{P_t(j)}{P_t} \right)^{-\epsilon} Y_t - \exp(\eta_{A,t}) K_{t-1}^{\alpha}(j) L_t^{1-\alpha}(j) \right),$$

so the first-order conditions are

$$0 = R_{K,t} - MC_t(j)\alpha \exp(\eta_{A,t}) \left(\frac{L_t(j)}{K_{t-1}(j)}\right)^{1-\alpha}$$
$$0 = W_t - MC_t(j)(1-\alpha) \exp(\eta_{A,t}) \left(\frac{K_{t-1}(j)}{L_t(j)}\right)^{\alpha},$$

hence the optimal capital-labor ratio satisfies

$$\frac{R_{K,t}}{\alpha \exp(\eta_{A,t})(K_{t-1}(j)/L_t(j))^{\alpha-1}} = \frac{W_t}{(1-\alpha) \exp(\eta_{A,t})(K_{t-1}(j)/L_t(j))^{\alpha}}$$
$$\frac{K_{t-1}(j)}{L_t(j)} = \frac{\alpha}{1-\alpha} \frac{W_t}{R_{K,t}}.$$

Since the RHS does not vary with j, all firms choose the same capital-labor ratio. Given this optimal ratio, the marginal cost satisfies

$$MC_{t} = \frac{R_{K,t}}{\alpha \exp(\eta_{A,t})} \left(\frac{K_{t-1}}{L_{t}}\right)^{1-\alpha}$$

$$= \frac{R_{K,t}}{\alpha \exp(\eta_{A,t})} \left(\frac{\alpha}{1-\alpha} \frac{W_{t}}{R_{K,t}}\right)^{1-\alpha}$$

$$= \left(\frac{1}{1-\alpha}\right)^{1-\alpha} \left(\frac{1}{\alpha}\right)^{\alpha} \frac{W_{t}^{1-\alpha} R_{K,t}^{\alpha}}{\exp(\eta_{A,t})}.$$

It follows that

$$R_{K,t}K_{t-1} + W_tL_t = \left(\frac{R_{K,t}}{\exp(\eta_{A,t})} \left(\frac{K_{t-1}}{L_t}\right)^{1-\alpha} + \frac{W_t}{\exp(\eta_{A,t})} \left(\frac{L_t}{K_{t-1}}\right)^{\alpha}\right) (\exp(\eta_{A,t})K_{t-1}^{\alpha}L_t^{1-\alpha})$$

$$= (\alpha MC_t + (1-\alpha)MC_t)Y_t(j) = MC_tY_t(j).$$

Therefore, (real) profits for an intermediate producer become

$$F_t(j) = \frac{P_t(j)}{P_t} Y_t(j) - MC_t Y_t(j).$$

In addition to the capital-labor choice, firms also have the chance to reset prices in every period with probability  $1 - \theta$ . This problem can be written as

$$\max_{P_{t}(j)} \mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \frac{\exp(\eta_{\beta,t+s})}{\exp(\eta_{\beta,t})} \frac{u'(C_{t+s})}{u'(C_{t})} \left( \frac{P_{t}(j)}{P_{t+s}} \left( \frac{P_{t}(j)}{P_{t+s}} \right)^{-\epsilon} Y_{t+s} - mc_{t+s} \left( \frac{P_{t}(j)}{P_{t+s}} \right)^{-\epsilon} Y_{t+s} \right),$$

where I have imposed that intermediate output equals demand. The first-order condition is

$$0 = (1 - \epsilon)P_{t}(j)^{-\epsilon} \mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \frac{\exp(\eta_{\beta,t+s})}{\exp(\eta_{\beta,t})} \frac{u'(C_{t+s})}{u'(C_{t})} (P_{t+s})^{-(1-\epsilon)} Y_{t+s}$$
$$+ \epsilon P_{t}(j)^{-\epsilon-1} \mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \frac{\exp(\eta_{\beta,t+s})}{\exp(\eta_{\beta,t})} \frac{u'(C_{t+s})}{u'(C_{t})} m c_{t+s} P_{t+s}^{\epsilon} Y_{t+s}$$

Divide by  $P_t(j)^{-\epsilon}/(\exp(\eta_{\beta,t})u'(C_t))$ , apply the abuse of notation that  $\prod_{u=1}^0 \Pi_{t+u} = 1$ , and re-arrange to obtain

$$P_{t}(j) = \frac{\epsilon}{\epsilon - 1} \frac{\mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \exp(\eta_{\beta,t+s}) u'(C_{t+s}) m c_{t+s} P_{t+s}^{\epsilon} Y_{t+s}}{\mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \exp(\eta_{\beta,t+s}) u'(C_{t+s}) P_{t+s}^{\epsilon-1} Y_{t+s}}$$

$$= \frac{\epsilon}{\epsilon - 1} \frac{\mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \exp(\eta_{\beta,t+s}) u'(C_{t+s}) m c_{t+s} P_{t}^{\epsilon} (\prod_{u=1}^{s} \prod_{t+u})^{\epsilon} Y_{t+s}}{\mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \exp(\eta_{\beta,t+s}) u'(C_{t+s}) P_{t}^{\epsilon-1} (\prod_{u=1}^{s} \prod_{t+u})^{\epsilon-1} Y_{t+s}}$$

$$\frac{P_{t}(j)}{P_{t}} = \frac{\epsilon}{\epsilon - 1} \frac{\mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \exp(\eta_{\beta,t+s}) u'(C_{t+s}) m c_{t+s} (\prod_{u=1}^{s} \prod_{t+u})^{\epsilon} Y_{t+s}}{\mathbb{E}_{t} \sum_{s=0}^{\infty} (\beta \theta)^{s} \exp(\eta_{\beta,t+s}) u'(C_{t+s}) (\prod_{u=1}^{s} \prod_{t+u})^{\epsilon-1} Y_{t+s}}.$$

This expression gives the optimal (real) reset price  $P_t^* \equiv P_t(j)/P_t$  (note that the RHS does not depend on j). Define

$$S_{1,t} = \mathbb{E}_t \sum_{s=0}^{\infty} (\beta \theta)^s \exp(\eta_{\beta,t+s}) u'(C_{t+s}) m c_{t+s} Y_{t+s} \left( \prod_{u=1}^s \Pi_{t+s} \right)^{\epsilon},$$
  
$$S_{2,t} = \mathbb{E}_t \sum_{s=0}^{\infty} (\beta \theta)^s \exp(\eta_{\beta,t+s}) u'(C_{t+s}) Y_{t+s} \left( \prod_{u=1}^s \Pi_{t+s} \right)^{\epsilon-1}.$$

Using these definitions, I may write the optimal reset price more compactly as

$$P_t^* = \frac{\epsilon}{\epsilon - 1} \frac{S_{1,t}}{S_{2,t}}$$

where  $S_{1,t}$  and  $S_{2,t}$  satisfy the recursions

$$S_{1,t} = \exp(\eta_{\beta,t}) u'(C_t) M C_t Y_t + \theta \beta \mathbb{E}_t \Pi_{t+s}^{\epsilon} S_{1,t+1}$$
  

$$S_{2,t} = \exp(\eta_{\beta,t}) u'(C_t) Y_t + \theta \beta \mathbb{E}_t \Pi_{t+s}^{\epsilon-1} S_{2,t+1}.$$

These recursions can be further rewritten as

$$\frac{S_{1,t}}{\exp(\eta_{\beta,t})u'(C_t)} = MC_tY_t + \theta\beta\mathbb{E}_t \left[ \frac{\exp(\eta_{\beta,t+1})u'(C_{t+1})}{\exp(\eta_{\beta,t})u'(C_t)} \Pi_{t+s}^{\epsilon} \frac{S_{1,t+1}}{\exp(\eta_{\beta,t+1})u'(C_{t+1})} \right] 
\frac{S_{2,t}}{\exp(\eta_{\beta,t})u'(C_t)} = Y_t + \theta\beta\mathbb{E}_t \left[ \frac{\exp(\eta_{\beta,t+1})u'(C_{t+1})}{\exp(\eta_{\beta,t})u'(C_t)} \Pi_{t+s}^{\epsilon-1} \frac{S_{2,t+1}}{\exp(\eta_{\beta,t+1})u'(C_{t+1})} \right],$$

By defining  $\tilde{S}_{1,t} \equiv S_{1,t}/(\exp(\eta_{\beta,t})u'(C_t))$  and  $\tilde{S}_{2,t} \equiv S_{2,t}/(\exp(\eta_{\beta,t})u'(C_t))$ , I can simplify these recursions into the form I use for the numerical solution.

From this section, we obtain the following five equilibrium conditions:

$$MC_t = \left(\frac{1}{1-\alpha}\right)^{1-\alpha} \left(\frac{1}{\alpha}\right)^{\alpha} \frac{W_t^{1-\alpha} R_{K,t}^{\alpha}}{\exp(\eta_{A,t})},\tag{9}$$

$$\frac{K_{t-1}}{L_t} = \frac{\alpha}{1 - \alpha} \frac{W_t}{R_{K,t}},\tag{10}$$

$$P_t^* = \frac{\epsilon}{\epsilon - 1} \frac{\tilde{S}_{1,t}}{\tilde{S}_{2,t}},\tag{11}$$

$$\tilde{S}_{1,t} = MC_t Y_t + \theta \mathbb{E}_t [M_{t+1} \Pi_{t+1}^{\epsilon} \tilde{S}_{1,t+1}], \tag{12}$$

$$\tilde{S}_{2,t} = Y_t + \theta \mathbb{E}_t [M_{t+1} \Pi_{t+1}^{\epsilon - 1} \tilde{S}_{2,t+1}]. \tag{13}$$

### 1.3 Monetary Policy

I specify the monetary policy rule as the following Taylor rule

$$\frac{R_t}{R} = \left(\frac{R_{t-1}}{R}\right)^{\phi_R} \left(\left(\frac{\Pi_t}{\Pi}\right)^{\phi_\pi} \left(\frac{Y_t}{Y_{t-1}}\right)^{\phi_y}\right)^{1-\phi_R} \exp(\eta_{R,t}) \tag{14}$$

Any proceeds from monetary policy are distributed as lump sum to the representative household.

# 1.4 Aggregation

The price level is currently characterized as the integral

$$P_t^{1-\epsilon} = \int_0^1 P_t(j)^{1-\epsilon} \, dj.$$

To represent the model entirely in terms of aggregates, notice that, without loss of generality, we may re-order the fraction  $\theta$  of firms which cannot reset prices to the top of the interval so that

$$P_t^{1-\epsilon} = (1-\theta)(P_t^*)^{1-\epsilon} + \int_{1-\theta}^1 P_{t-1}(j)^{1-\epsilon} \, dj.$$

The latter term can be further simplified under the law of large numbers assumption that a positive measure of firms which cannot change their price still comprise a representative sample of all firms, yielding

$$P_t^{1-\epsilon} = (1-\theta)(P_t^*)^{1-\epsilon} + \theta \int_0^1 P_{t-1}(j)^{1-\epsilon} dj = (1-\theta)(P_t^*)^{1-\epsilon} + \theta P_{t-1}^{1-\epsilon}.$$

Dividing by  $P_{t-1}^{1-\epsilon}$  implies

$$\Pi_t^{1-\epsilon} = (1-\theta)(P_t^*\Pi_t)^{1-\epsilon} + \theta. \tag{15}$$

The price dispersion term can similarly be re-written in terms of aggregates by distinguishing which firms get to change prices.

$$V_{t}^{p} = \int_{0}^{1-\theta} (P_{t}^{*})^{-\epsilon} dj + \int_{1-\theta}^{1} \left(\frac{P_{t-1}(j)}{P_{t}}\right)^{-\epsilon} dj$$

$$= \int_{0}^{1-\theta} (P_{t}^{*}\Pi_{t})^{-\epsilon} \left(\frac{1}{\Pi_{t}}\right)^{-\epsilon} dj + \int_{1-\theta}^{1} \left(\frac{P_{t-1}(j)}{P_{t-1}}\right)^{-\epsilon} \left(\frac{P_{t-1}}{P_{t}}\right)^{-\epsilon} dj$$

$$= (1-\theta)(P_{t}^{*}\Pi_{t})^{-\epsilon}\Pi_{t}^{\epsilon} + \Pi_{t}^{\epsilon} \int_{1-\theta}^{1} \left(\frac{P_{t-1}(j)}{P_{t-1}}\right)^{-\epsilon} dj.$$

By invoking the law of large assumptions applied to any positive measure subset of firms, we must have

$$\int_{1-\theta}^{1} \left( \frac{P_{t-1}(j)}{P_{t-1}} \right)^{-\epsilon} dj = \theta \int_{0}^{1} \left( \frac{P_{t-1}(j)}{P_{t-1}} \right)^{-\epsilon} dj = \theta V_{t-1}^{p}.$$

Thus, we acquire

$$V_t^p = \Pi_t^{\epsilon} ((1 - \theta)(P_t^* \Pi_t)^{-\epsilon} + \theta V_{t-1}^p)$$
(16)

# 1.5 Equilibrium

To close the model, I need to specify the functional form for investment, aggregate shocks, and market-clearing conditions.

Following Jermann (1998), I assume the investment function takes the concave form

$$\Phi\left(\frac{X_t}{K_{t-1}}\right) = \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} \left(\frac{X_t}{K_{t-1}}\right)^{1 - 1/\chi} - \frac{\overline{X}}{\chi(\chi - 1)}$$

$$\tag{17}$$

where  $\overline{X} = \delta \chi / (\chi + 1)$  is the steady-state investment rate (per unit of capital). The first derivative of  $\Phi(\cdot)$  w.r.t.  $X_t/K_{t-1}$  is

$$\Phi'\left(\frac{X_t}{K_{t-1}}\right) = \overline{X}^{1/\chi} \left(\frac{X_t}{K_{t-1}}\right)^{-1/\chi}.$$
(18)

This functional form implies the law of motion

$$K_{t} = \left(1 - \delta + \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} \left(\frac{X_{t}}{K_{t-1}}\right)^{1 - 1/\chi} - \frac{\overline{X}}{\chi(\chi - 1)}\right) K_{t-1}$$

$$= \left(1 + \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} \left(\frac{X_{t}}{K_{t-1}}\right)^{1 - 1/\chi} - \delta \left(1 + \frac{1}{(\chi - 1)(\chi + 1)}\right)\right) K_{t-1}$$

$$= \left(1 + \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} \left(\frac{X_{t}}{K_{t-1}}\right)^{1 - 1/\chi} - \delta \left(\frac{\chi^{2}}{(\chi - 1)(\chi + 1)}\right)\right) K_{t-1}$$

$$= \left(1 + \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} \left(\frac{X_{t}}{K_{t-1}}\right)^{1 - 1/\chi} - \frac{\delta \chi^{2}}{\chi^{2}(1 - 1/\chi)(1 + 1/\chi)}\right) K_{t-1}$$

$$= \left(1 + \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} \left(\frac{X_{t}}{K_{t-1}}\right)^{1 - 1/\chi} - \frac{\overline{X}}{1 - 1/\chi}\right) K_{t-1}.$$

If  $K_t = K_{t-1} = K_{ss}$  and  $X_{ss}/K_{ss} = \overline{X}$ , then

$$1 + \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} \overline{X}^{1 - 1/\chi} - \frac{\overline{X}}{1 - 1/\chi} = 1 + \frac{\overline{X}}{1 - 1/\chi} - \frac{\overline{X}}{1 - 1/\chi} = 1,$$

thus verifying the original conjecture that  $\overline{X}$  represents the steady-state investment rate. There are four shocks in the model:  $\eta_{A,t}$ ,  $\eta_{\beta,t}$ ,  $\eta_{L,t}$ , and  $\eta_{R,t}$ . Without loss of generality, I assume all shocks follow AR(1) processes with persistence  $\rho_i$  and standard deviation  $\sigma_i$ .

Markets must clear for capital, labor, bonds, final goods, and intermediate goods, . The first three markets clear as a consequence of optimality conditions and the assumption that bonds have zero net supply. To clear the market for final goods, we set the sum of aggregate consumption demand  $C_t$  and investment demand  $X_t$  equal to aggregate supply  $Y_t$ , which satisfies

$$\int_{0}^{1} \exp(\eta_{A,t}) K_{t-1}^{\alpha} L_{t}^{1-\alpha} dj = \int_{0}^{1} \left(\frac{P_{t}(j)}{P_{t}}\right)^{-\epsilon} Y_{t} dj$$
$$\exp(\eta_{A,t}) K_{t-1}^{\alpha} L_{t}^{1-\alpha} = Y_{t} \int_{0}^{1} \left(\frac{P_{t}(j)}{P_{t}}\right)^{-\epsilon} dj = V_{t}^{p} Y_{t}.$$

Re-arranging yields the output market-clearing condition

$$C_t + X_t = Y_t, (19)$$

$$Y_t = \frac{\exp(\eta_{A,t}) K_{t-1}^{\alpha} L_t^{1-\alpha}}{V_t^p}.$$
 (20)

It can be shown that  $V_t^p \ge 1$  by applying Jensen's inequality. For our purposes, because the dimensionality of our model is not too large, we add the auxiliary  $Y_t$  variable, even though we could substitute it out of the system of equations.

All together, the full set of endogenous equilibrium conditions are

$$C_t^{-\gamma} W_t = \varphi \exp(\eta_{L,t}) L_t^{\nu}, \tag{21}$$

$$M_{t+1} = \beta \frac{\exp(\eta_{\beta,t+1})}{\exp(\eta_{\beta,t})} \frac{C_{t+1}^{-\gamma}}{C_t^{-\gamma}},\tag{22}$$

$$1 = \mathbb{E}_t \left[ M_{t+1} \frac{R_t}{\Pi_{t+1}} \right], \tag{23}$$

$$1 = Q_t \Phi'\left(\frac{X_t}{K_{t-1}}\right),\tag{24}$$

$$Q_{t} = \mathbb{E}_{t} \left[ M_{t+1} \left( R_{K,t+1} + Q_{t+1} \left( 1 - \delta + \Phi \left( \frac{X_{t+1}}{K_{t}} \right) - \Phi' \left( \frac{X_{t+1}}{K_{t}} \right) \frac{X_{t+1}}{K_{t}} \right) \right], \quad (25)$$

$$MC_t = \left(\frac{1}{1-\alpha}\right)^{1-\alpha} \left(\frac{1}{\alpha}\right)^{\alpha} \frac{W_t^{1-\alpha} R_{K,t}^{\alpha}}{\exp(\eta_{A,t})},\tag{26}$$

$$\frac{K_{t-1}}{L_t} = \frac{\alpha}{1 - \alpha} \frac{W_t}{R_{K_t}},\tag{27}$$

$$P_t^* = \frac{\epsilon}{\epsilon - 1} \frac{\tilde{S}_{1,t}}{\tilde{S}_{2,t}},\tag{28}$$

$$\tilde{S}_{1,t} = MC_t Y_t + \theta \mathbb{E}_t [M_{t+1} \Pi_{t+1}^{\epsilon} \tilde{S}_{1,t+1}], \tag{29}$$

$$\tilde{S}_{2,t} = Y_t + \theta \mathbb{E}_t[M_{t+1} \Pi_{t+1}^{\epsilon - 1} \tilde{S}_{2,t+1}], \tag{30}$$

$$\Pi_t^{1-\epsilon} = (1-\theta)(P_t^*\Pi_t)^{1-\epsilon} + \theta, \tag{31}$$

$$V_t^p = \Pi_t^{\epsilon} ((1 - \theta)(P_t^* \Pi_t)^{-\epsilon} + \theta V_{t-1}^p), \tag{32}$$

$$\frac{R_t}{R} = \left(\frac{R_{t-1}}{R}\right)^{\phi_R} \left(\left(\frac{\Pi_t}{\Pi}\right)^{\phi_\pi} \left(\frac{Y_t}{Y_{t-1}}\right)^{\phi_y}\right)^{1-\phi_R} \exp(\eta_{R,t}),\tag{33}$$

$$C_t + X_t = Y_t, (34)$$

$$Y_t = \frac{\exp(\eta_{A,t}) K_{t-1}^{\alpha} L_t^{1-\alpha}}{V_t^p},\tag{35}$$

as well as the law of motion for capital

$$K_{t} = \left(1 + \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} \left(\frac{X_{t}}{K_{t-1}}\right)^{1 - 1/\chi} - \frac{\overline{X}}{1 - 1/\chi}\right) K_{t-1}$$
(36)

and the four exogenous processes

$$\eta_{\beta,t+1} = \rho_{\beta}\eta_{\beta,t} + \sigma_{\beta}\varepsilon_{\beta,t+1},\tag{37}$$

$$\eta_{L,t+1} = \rho_L \eta_{L,t} + \sigma_L \varepsilon_{L,t+1},\tag{38}$$

$$\eta_{A,t+1} = \rho_A \eta_{A,t} + \sigma_A \varepsilon_{A,t+1},\tag{39}$$

$$\eta_{R,t+1} = \rho_R \eta_{R,t} + \sigma_R \varepsilon_{R,t+1}. \tag{40}$$

### 1.6 Deterministic Steady State

To provide an initial guess for the risk-adjusted linearization and to provide a verification that the model is coded correctly, I determine some reasonable guesses for the deterministic steady state.

Within this subsection, I denote the deterministic steady state values by an absence of a time subscript. The exogenous processes, by construction, have steady states of 0, i.e.  $\eta_{\beta} = \eta_{L} = \eta_{A} = \eta_{R} = 0$ . Further, A = 1.

Focusing now on the endogenous equilibrium conditions, from (21),

$$W = \frac{\varphi L^{\nu}}{C^{-\gamma}}.$$

From (22),

$$M = \beta$$
.

From (24), the fact that  $\overline{X}$  is the steady-state investment rate, and the fact that  $\Phi'(\overline{X}) = 1$ ,

$$Q=1.$$

From (25), first observing that,

$$\Phi(\overline{X}) = \frac{\overline{X}\chi}{\chi - 1} - \frac{\overline{X}}{\chi(\chi - 1)} = \overline{X}\frac{\chi^2 - 1}{\chi(\chi - 1)} = \overline{X}\frac{(\chi - 1)(\chi + 1)}{\chi(\chi - 1)} = \frac{\delta\chi}{\chi + 1}\frac{\chi + 1}{\chi} = \delta,$$

which ensures that K does indeed remain at steady state, it must be the cast that

$$1 = \beta (R_K + (1 - \delta + \Phi(\overline{X}) - \overline{X})$$
$$R_K = \frac{1}{\beta} + \overline{X} - 1.$$

Equation (26) remains as it is but with time subscripts removed. From (29),

$$\tilde{S}_1 = MC \cdot Y + \theta \beta \Pi^{\epsilon} \tilde{S}_1 \Rightarrow \tilde{S}_1 = \frac{MC \cdot Y}{1 - \theta \beta \Pi^{\epsilon}}.$$

From (30),

$$\tilde{S}_2 = Y + \theta \beta \Pi^{\epsilon - 1} \tilde{S}_2 \Rightarrow \tilde{S}_1 = \frac{Y}{1 - \theta \beta \Pi^{\epsilon - 1}}.$$

Thus,

$$P^* = \frac{\epsilon}{\epsilon - 1} MC \frac{1 - \theta \Pi^{\epsilon - 1}}{1 - \theta \Pi^{\epsilon}}.$$

From (31),

$$\Pi^{1-\epsilon} = (1-\theta) \left(P^*\Pi\right)^{1-\epsilon} + \theta$$

Note that  $P^*$  depends on MC and fundamental parameters, hence the above equation pins down MC, which then pins down the ratio of K to L. From (32),

$$V^{p} = \Pi^{\epsilon} \left( (1 - \theta)(P^{*}\Pi)^{-\epsilon} + \theta V^{p} \right)$$
$$V^{p} = \frac{(1 - \theta)(P^{*}\Pi)^{-\epsilon}}{\Pi^{-\epsilon} - \theta} = \frac{(1 - \theta)(P^{*})^{-\epsilon}}{1 - \theta\Pi^{\epsilon}}.$$

From the Taylor rule (33), the steady state interest and inflation rates are R and  $\Pi$ , respectively, and from the Euler equation (23), R must satisfy

$$R = \frac{\Pi}{\beta}.$$

From (34),

$$C + X = Y$$
.

From (35),

$$Y = \frac{K^{\alpha}L^{1-\alpha}}{V^p}.$$

As shown previously, the steady-state investment rate is  $\overline{X}$ , hence

$$X = \overline{X}K$$

Finally, I claim that the deterministic steady state reduces to a nonlinear equation in L. Using the aggregate supply and capital accumulation equations,

$$C + \overline{X}K = \frac{K^{\alpha}L^{1-\alpha}}{V^p}.$$

The optimal capital-labor ratio implies

$$K = \frac{\alpha}{1 - \alpha} \frac{W}{R_K} L,$$

$$C + \overline{X}K = \left(\frac{\alpha}{1 - \alpha}\right)^{\alpha} \left(\frac{W}{R_K}\right)^{\alpha} \frac{L}{V^p}.$$

The intratemporal condition for consumption and labor implies

$$K = \frac{\alpha}{1 - \alpha} \frac{\varphi L^{\nu}}{C^{-\gamma} R_K} L,$$

$$C + \delta K = \left(\frac{\alpha}{1 - \alpha}\right)^{\alpha} \left(\frac{\varphi L^{\nu}}{C^{-\gamma} R_K}\right)^{\alpha} \frac{L}{V^p}.$$

Given a guess for L, I can compute C using these two equations. Given C, I can compute W. Given the wage W, I can compute K and MC. Given the marginal cost MC, I can compute the inflation-related terms.

# 2 Risk-Adjusted Linearization

We now proceed to converting the equilibrium conditions into a suitable form for a risk-adjusted linearization. The system should conform to the representation

$$0 = \log \mathbb{E}_t \left[ \exp \left( \xi(z_t, y_t) + \Gamma_5 z_{t+1} + \Gamma_6 y_{t+1} \right) \right]$$
  
$$z_{t+1} = \mu(z_t, y_t) + \Lambda(z_t, y_t) (y_{t+1} - \mathbb{E}_t y_{t+1}) + \Sigma(z_t, y_t) \varepsilon_{t+1},$$

where  $z_t$  are (predetermined) state variables and  $y_t$  are (nondetermined) jump variables. For the remainder of this section, lower case variables are the logs of previously upper case variables, and with a small abuse of notation, let  $s_{1,t} = \log(\tilde{S}_{1,t})$  and  $s_{2,t} = \log(\tilde{S}_{2,t})$ . Additionally, let  $r_{k,t} = \log(R_{K,t})$  and  $v_t = \log(V_t^p)$ .

Equation (21) becomes

$$1 = \varphi \exp(\eta_{L,t}) \frac{L_t^{\nu}}{C_t^{-\gamma} W_t}$$

$$0 = \log \mathbb{E}_t \left[ \exp\left(\underbrace{\log(\varphi) + \eta_{L,t} + \nu l_t - (-\gamma c_t + w_t)}_{\xi}\right) \right].$$

Equation (22) will not be used in the system of equations for the risk-adjusted linearization, but it simplifies the other equations. Taking logs and re-arranging yields

$$0 = \log(\beta) + \eta_{\beta,t+1} + (-\gamma c_{t+1}) - \eta_{\beta,t} - (-\gamma c_t) - m_{t+1}$$

$$m_{t+1} = \underbrace{\log(\beta) - \eta_{\beta,t} + \gamma c_t}_{\text{forward-looking}} + \underbrace{\eta_{\beta,t+1} - \gamma c_{t+1}}_{\text{forward-looking}}.$$

Equation (23) becomes

$$0 = \log \mathbb{E}_t \left[ \exp \left( \underbrace{r_t}_{\xi} + \underbrace{m_{t+1}}_{\text{both}} - \underbrace{\pi_{t+1}}_{\text{forward-looking}} \right) \right].$$

Equation (24) becomes

$$0 = \log \mathbb{E}_t \left[ \exp \left( q_t + \log \left( \Phi' \left( \exp(x_t - k_{t-1}) \right) \right) \right) \right].$$

For equation (25), observe that the RHS is not log-linear in the forward-looking variables. To handle this case, I define the new variable

$$\Omega_{t} = R_{K,t} + Q_{t} \left( 1 - \delta + \Phi \left( \frac{X_{t}}{K_{t-1}} \right) - \Phi' \left( \frac{X_{t}}{K_{t-1}} \right) \frac{X_{t}}{K_{t-1}} \right). \tag{41}$$

Then (25) can be written as

$$1 = \mathbb{E}_t \left[ \frac{M_{t+1}\Omega_{t+1}}{Q_t} \right]$$
$$0 = \log \mathbb{E}_t \left[ \exp(m_{t+1} + \omega_{t+1} - q_t) \right].$$

Equation (26) becomes

$$0 = \log \mathbb{E}_t \left[ \exp \left( \underbrace{(1 - \alpha)w_t + \alpha r_{k,t} - a_t - (1 - \alpha)\log(1 - \alpha) - \alpha\log(\alpha) - mc_t}_{\xi} \right) \right].$$

Equation (27) becomes

$$0 = \log \mathbb{E}_t \left[ \exp \left( \underbrace{k_{t-1} - l_t - \log \left( \frac{\alpha}{1 - \alpha} \right) - (w_t - r_{k,t})}_{\xi} \right) \right].$$

Like the stochastic discount factor, equation (28) will not be used in the system of equations, but it will be useful to simplify other equations. Taking logs yields

$$p_t^* = \log\left(\frac{\epsilon}{\epsilon - 1}\right) + s_{1,t} - s_{2,t}.$$

Equation (29) becomes

$$\tilde{S}_{1,t} - MC_t Y_t$$

$$= \mathbb{E}_t \left[ \exp\left(\log(\theta) + m_{t+1} + \epsilon \pi_{t+1} + s_{1,t+1}\right) \right]$$

$$0 = \log \mathbb{E}_t \left[ \exp\left(\underbrace{\log(\theta) - \log(\exp(s_{1,t}) - \exp(mc_t) \exp(y_t)}_{\xi} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{\epsilon \pi_{t+1} + s_{1,t+1}}_{\text{forward-looking}} \right) \right]$$

and equation (30) becomes

$$0 = \log \mathbb{E}_t \left[ \exp \left( \underbrace{\log(\theta) - \log(\exp(s_{2,t}) - \exp(y_t))}_{\xi} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1} + s_{2,t+1}}_{\text{forward-looking}} \right) \right].$$

Equation (31) becomes

$$1 = \frac{\Pi_t^{1-\epsilon}}{(1-\theta)(P_t^*\Pi_t)^{1-\epsilon} + \theta}$$
  

$$0 = \log \mathbb{E}_t \left[ \exp\left( (1-\epsilon)\pi_t - \log((1-\theta)\exp((1-\epsilon)(p_t^* + \pi_t)) + \theta \right) \right].$$

Equation (32) becomes

$$1 = \frac{V_t^p}{\Pi_t^{\epsilon}((1 - \theta)(P_t^*\Pi_t)^{-\epsilon} + \theta V_{t-1}^p)}$$

$$0 = \log \mathbb{E}_t \left[ \exp \left( v_t - \epsilon \pi_t - \log((1 - \theta) \exp(-\epsilon(p_t^* + \pi_t)) + \theta \exp(v_{t-1}) \right) \right].$$

Equation (33) becomes

$$0 = \log \mathbb{E}_t \left[ \exp \left( \phi_R r_{t-1} + (1 - \phi_R) r + (1 - \phi_R) (\phi_\pi (\pi_t - \pi) + \phi_y (y_t - y_{t-1})) + \eta_{R,t} - r_t \right) \right].$$

Equation (34) becomes

$$0 = \log \mathbb{E}_t \left[ \exp(y_t - \log(\exp(c_t) + \exp(x_t))) \right].$$

Equation (35) becomes

$$0 = \log \mathbb{E}_t \left[ \exp(a_t + \alpha k_{t-1} + (1 - \alpha)l_t - v_t - y_t) \right].$$

Equation (36) becomes

$$k_t = \log \left( 1 + \frac{\overline{X}^{1/\chi}}{1 - 1/\chi} (\exp(x_t - k_{t-1}))^{1 - 1/\chi} - \frac{\overline{X}}{1 - 1/\chi} \right) + k_{t-1}.$$

The autoregressive processes (37) to (40) remain as they are.

The jump variables are  $y_t$ ,  $c_t$ ,  $l_t$ ,  $w_t$ ,  $r_t$ ,  $q_t$ ,  $x_t$ ,  $r_{k,t}$ ,  $\omega_t$ ,  $mc_t$ ,  $s_{1,t}$ ,  $s_{2,t}$ , and  $v_t$ . The state variables are  $k_{t-1}$ ,  $v_{t-1}$ ,  $r_{t-1}$ ,  $y_{t-1}$ , and the autoregressive processes. The equations defining the evolution of the lags  $v_{t-1}$ ,  $r_{t-1}$ , and  $y_{t-1}$  are obtained by the formula  $z_{(t-1)+1} = z_t$ . This system has three forward difference equations (25), (29), and (30). To ensure accuracy of the risk-adjusted linearization, I derive N-period ahead forward difference equations for all three.

First, redefine  $\Omega_t$  as

$$\Omega_t = 1 - \delta + \Phi\left(\frac{X_t}{K_{t-1}}\right) - \Phi'\left(\frac{X_t}{K_{t-1}}\right) \frac{X_t}{K_{t-1}}.$$

Then we can write (25) recursively as

$$Q_{t} = \mathbb{E}_{t}[M_{t+1}(R_{K,t+1} + Q_{t+1}\Omega_{t+1})]$$

$$= \mathbb{E}_{t}[M_{t+1}R_{K,t+1} + \Omega_{t+1}M_{t+1}\mathbb{E}_{t+1}[M_{t+2}(R_{K,t+2} + Q_{t+2}\Omega_{t+1})]]$$

$$= \mathbb{E}_{t}[M_{t+1}R_{K,t+1}] + \Omega_{t+1}\mathbb{E}_{t}\mathbb{E}_{t+1}[M_{t+1}M_{t+2}(R_{K,t+2} + Q_{t+2}\Omega_{t+2})].$$

By the tower property,

$$\begin{split} Q_t &= \mathbb{E}_t [M_{t+1} R_{K,t+1}] + \Omega_{t+1} \mathbb{E}_t [M_{t+1} M_{t+2} (R_{K,t+2} + Q_{t+2} \Omega_{t+2})] \\ &= \mathbb{E}_t \left[ \left( \sum_{s=1}^2 \left( \prod_{u=1}^{s-1} \Omega_{t+u} \right) \left( \prod_{u=1}^s M_{t+u} \right) R_{K,t+s} \right) + M_{t+1} M_{t+2} Q_{t+2} \Omega_{t+1} \Omega_{t+2} \right] \\ &= \mathbb{E}_t \left[ \left( \sum_{s=1}^2 \left( \prod_{u=1}^{s-1} \Omega_{t+u} \right) \left( \prod_{u=1}^s M_{t+u} \right) R_{K,t+s} \right) + \prod_{s=1}^2 (M_{t+s} \Omega_{t+s}) \mathbb{E}_{t+2} [M_{t+3} (R_{K,t+3} + Q_{t+3} \Omega_{t+3})] \right] \\ &= \mathbb{E}_t \left[ \left( \sum_{s=1}^3 \left( \prod_{u=1}^{s-1} \Omega_{t+u} \right) \left( \prod_{u=1}^s M_{t+u} \right) R_{K,t+s} \right) + \prod_{s=1}^3 (M_{t+s} \Omega_{t+s}) Q_{t+3} \right] \end{split}$$

and so on, with the abuse of notation that  $\prod_{u=1}^{0} \Omega_{t+u} = 1$ . Given this recursive structure, define  $D_{Q,t}^{(n)}$  and  $P_{Q,t}^{(n)}$  as

$$D_{Q,t}^{(n)} = \mathbb{E}_t \left[ \Omega_{t+1} M_{t+1} D_{Q,t+1}^{(n-1)} \right]$$
$$P_{Q,t}^{(n)} = \mathbb{E}_t \left[ \Omega_{t+1} M_{t+1} P_{Q,t+1}^{(n-1)} \right]$$

with boundary conditions

$$D_{Q,t}^{(0)} = \frac{R_{K,t+1}}{\Omega_{t+1}}$$
$$P_{Q,t}^{(0)} = Q_{t+1}.$$

Then I may write the N-period ahead recursive form of equation (25) as

$$Q_t = \sum_{n=1}^{N} D_{Q,t}^{(n)} + P_{Q,t}^{(N)}.$$

To see why this recursion works, it is simpler to first verify that  $P_{Q,t}^{(3)}$  is correct:

$$\begin{split} P_{Q,t}^{(1)} &= \mathbb{E}_t \left[ \Omega_{t+1} M_{t+1} Q_{t+1} \right] \\ P_{Q,t}^{(2)} &= \mathbb{E}_t \left[ \Omega_{t+1} M_{t+1} (\mathbb{E}_{t+1} [\Omega_{t+2} M_{t+2} Q_{t+2}]) \right] \\ &= \mathbb{E}_t \left[ \mathbb{E}_{t+1} \left[ \prod_{s=1}^2 (\Omega_{t+s} M_{t+s}) Q_{t+2} \right] \right] \\ &= \mathbb{E}_t \left[ \prod_{s=1}^2 (\Omega_{t+s} M_{t+s}) Q_{t+2} \right]. \end{split}$$

where the second equality for  $P_{Q,t}^{(2)}$  follows from the fact that  $M_{t+1}$  is measurable with respect to the information set at time t+1 and can therefore be moved insided the conditional

expectation  $\mathbb{E}_{t+1}[\cdot]$ . Continuing for one more recursion, I have

$$P_{Q,t}^{(3)} = \mathbb{E}_t \left[ \Omega_{t+1} M_{t+1} \mathbb{E}_{t+1} \left[ \prod_{s=1}^2 (\Omega_{t+1+s} M_{t+1+s}) Q_{t+3} \right] \right]$$
$$= \mathbb{E}_t \left[ \prod_{s=1}^3 (\Omega_{t+s} M_{t+s}) Q_{t+3} \right].$$

Similarly, for  $D_{Q,t}$ , I have

$$D_{Q,t}^{(1)} = \mathbb{E}_{t} \left[ \Omega_{t+1} M_{t+1} \frac{R_{K,t+1}}{\Omega_{t+1}} \right] = \mathbb{E}_{t} [M_{t+1} R_{K,t+1}]$$

$$D_{Q,t}^{(2)} = \mathbb{E}_{t} [\Omega_{t+1} M_{t+1} \mathbb{E}_{t+1} [M_{t+2} R_{K,t+2}]]$$

$$= \mathbb{E}_{t} [\Omega_{t+1} M_{t+1} M_{t+2} R_{K,t+2}]$$

$$D_{Q,t}^{(3)} = \mathbb{E}_{t} [\Omega_{t+1} M_{t+1} \mathbb{E}_{t+1} [\Omega_{t+2} M_{t+2} M_{t+3} R_{K,t+3}]]$$

$$= \mathbb{E}_{t} [\Omega_{t+1} \Omega_{t+2} M_{t+1} M_{t+2} M_{t+3} R_{K,t+3}].$$

Since  $P_{Q,t}^{(n)}$  and  $D_{Q,t}^{(n)}$  are time-t conditional expectations, they are measurable at time t, so they are not forward-looking variables. Thus, to get this version of (25) in the appropriate form, define  $d_{q,n,t} = \log(D_{Q,t}^{(n)})$  and  $p_{q,n,t} = \log(P_{Q,t}^{(n)})$ , and use the following 2N + 1 equations:

$$0 = \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\sum_{n=1}^{N} \exp(d_{q,n,t}) + \exp(p_{q,N,t})}_{\xi} \right) \right]$$

$$0 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{-d_{q,n,t} + \underbrace{m_{t+1} + \underbrace{\omega_{t+1} + d_{q,n-1,t+1}}_{\text{forward-looking}}} \right) \right] \text{ if } n > 1$$

$$0 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{-d_{q,1,t} + \underbrace{m_{t+1} + \underbrace{v_{k,t+1}}_{\text{forward-looking}}} \right) \right] \text{ if } n = 1,$$

$$0 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{-p_{q,n,t} + \underbrace{m_{t+1} + \underbrace{v_{t+1} + p_{q,n-1,t+1}}_{\text{forward-looking}}} \right) \right] \text{ if } n > 1$$

$$0 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{-p_{q,n,t} + \underbrace{m_{t+1} + \underbrace{\omega_{t+1} + p_{q,n-1,t+1}}_{\text{forward-looking}}} \right) \right] \text{ if } n > 1 \right\}$$

$$0 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{-p_{q,n,t} + \underbrace{m_{t+1} + \underbrace{\omega_{t+1} + p_{q,n-1,t+1}}_{\text{forward-looking}}} \right) \right] \text{ if } n > 1 \right\}$$

$$0 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{-p_{q,n,t} + \underbrace{m_{t+1} + \underbrace{\omega_{t+1} + q_{t+1}}_{\text{forward-looking}}} \right) \right] \text{ if } n > 1 \right\}$$

$$0 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{-p_{q,n,t} + \underbrace{m_{t+1} + \underbrace{\omega_{t+1} + q_{t+1}}_{\text{forward-looking}}} \right) \right] \text{ if } n > 1 \right\} \right\}$$

For (29), observe that

$$\begin{split} \tilde{S}_{1,t} &= MC_t Y_t + \theta \mathbb{E}_t [M_{t+1} \Pi_{t+1}^{\epsilon} (MC_{t+1} Y_{t+1} + \theta \mathbb{E}_{t+1} [M_{t+2} \Pi_{t+2}^{\epsilon} \tilde{S}_{1,t+2}])] \\ &= MC_t Y_t + \theta \mathbb{E}_t [M_{t+1} \Pi_{t+1}^{\epsilon} MC_{t+1} Y_{t+1} + \theta M_{t+1} \Pi_{t+1}^{\epsilon} M_{t+2} \Pi_{t+2}^{\epsilon} \tilde{S}_{1,t+2}] \\ &= MC_t Y_t + \mathbb{E}_t \left[ \sum_{s=1}^1 (\theta^s \prod_{u=1}^s (M_{t+u} \Pi_{t+u}^{\epsilon})) MC_{t+s} Y_{t+s} \right] + \mathbb{E}_t \left[ \prod_{s=1}^2 (\theta M_{t+s} \Pi_{t+s}^{\epsilon}) \tilde{S}_{1,t+2} \right]. \end{split}$$

Thus, define  $D_{S1,t}^{(n)}$  and  $P_{S1,t}^{(n)}$  as the recursions

$$D_{S1,t}^{(n)} = \mathbb{E}_t[\theta M_{t+1} \Pi_{t+1}^{\epsilon} D_{S1,t+1}^{(n-1)}],$$
  

$$P_{S1,t}^{(n)} = \mathbb{E}_t[\theta M_{t+1} \Pi_{t+1}^{\epsilon} P_{S1,t+1}^{(n-1)}],$$

with boundary conditions

$$D_{S1,t}^{(0)} = MC_t Y_t$$
$$P_{S1,t}^{(0)} = \tilde{S}_{1,t}.$$

Given these definitions, it follows that

$$D_{S1,t}^{(1)} = \mathbb{E}_{t}[\theta M_{t+1}\Pi_{t+1}^{\epsilon}MC_{t+1}Y_{t+1}]$$

$$P_{S1,t}^{(1)} = \mathbb{E}_{t}[\theta M_{t+1}\Pi_{t+1}^{\epsilon}\tilde{S}_{1,t+1}]$$

$$P_{S1,t}^{(2)} = \mathbb{E}_{t}[\theta M_{t+1}\Pi_{t+1}^{\epsilon}\mathbb{E}_{t+1}[\theta M_{t+2}\Pi_{t+2}^{\epsilon}\tilde{S}_{1,t+2}]]$$

$$= \mathbb{E}_{t}[\theta^{2}M_{t+1}\Pi_{t+1}^{\epsilon}M_{t+2}\Pi_{t+2}^{\epsilon}\tilde{S}_{1,t+2}].$$

Thus, defining  $d_{s1,t} = \log(D_{S1,t})$  and  $p_{s1,t} = \log(P_{S1,t})$ , the N-period ahead recursive form of

(29) results in the 2N + 1 equations

$$0 = \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\sum_{s_{1,t} - \log \left( \sum_{n=0}^{N-1} \exp(d_{s_{1,n,t}}) + \exp(p_{s_{1,N,t}}) \right)}_{\xi} \right) \right]$$

$$0 = \begin{cases} \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\log(\theta) - d_{s_{1,n,t}} + m_{t+1} + \epsilon \pi_{t+1} + d_{s_{1,n-1,t+1}}}_{\text{both}} \right) \right] & \text{if } n \geq 1 \\ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{d_{s_{1,0,t}} - mc_{t} - y_{t}}_{\xi} \right) \right] & \text{if } n = 0. \end{cases}$$

$$0 = \begin{cases} \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\log(\theta) - p_{s_{1,n,t}} + m_{t+1} + \epsilon \pi_{t+1} + p_{s_{1,n-1,t+1}}}_{\text{both}} \right) \right] & \text{if } n > 1 \\ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\log(\theta) - p_{s_{1,n,t}} + m_{t+1} + \epsilon \pi_{t+1} + p_{s_{1,n-1,t+1}}}_{\text{both}} \right) \right] & \text{if } n = 1. \end{cases}$$

$$(47)$$

It is straightforward to show that a similar recursive form applies to (30):

$$0 = \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\sum_{s_{2,t} - \log \left( \sum_{n=0}^{N-1} \exp(d_{s_{2,n,t}}) + \exp(p_{s_{2,N,t}}) \right)}_{\xi} \right) \right]$$

$$0 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\log(\theta) - d_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1} + d_{s_{2,n-1,t+1}}}_{\text{forward-looking}} \right) \right] \text{ if } n \ge 1$$

$$\log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\frac{d_{s_{2,0,t}} - y_{t}}{\xi}}_{\xi} \right) \right]$$

$$1 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}}}{\xi} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1} + p_{s_{2,n-1,t+1}}}_{\text{forward-looking}} \right) \right] \right]$$

$$1 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1} + p_{s_{2,n-1,t+1}}}_{\text{forward-looking}} \right) \right] \right\}$$

$$1 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1} + p_{s_{2,n-1,t+1}}}_{\text{forward-looking}} \right) \right] \right\}$$

$$1 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1} + p_{s_{2,n-1,t+1}}}_{\text{forward-looking}} \right) \right] \right\}$$

$$1 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1}}_{\text{forward-looking}} \right) \right\} \right\}$$

$$1 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1}}_{\text{forward-looking}} \right)}_{\text{forward-looking}} \right] \right\}$$

$$1 = \left\{ \log \mathbb{E}_{t} \left[ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1}}_{\text{forward-looking}} \right)}_{\text{forward-looking}} \right] \right\}$$

$$1 = \left\{ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1}}_{\text{forward-looking}} \right)}_{\text{forward-looking}} \right\} \right\}$$

$$1 = \left\{ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1}}_{\text{forward-looking}} \right)}_{\text{forward-looking}} \right\}$$

$$1 = \left\{ \exp \left( \underbrace{\frac{\log(\theta) - p_{s_{2,n,t}} + \underbrace{m_{t+1}}_{\text{both}} + \underbrace{(\epsilon - 1)\pi_{t+1}}_{\text{forward-looking}} \right)}_{\text{forward-looking}} \right\} \right\}$$

where terms and boundary conditions are analogously defined.