

The Optimum Quantity of Capital and Debt*

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February 15, 2015

Abstract

In this paper we consider an optimal taxation problem in an incomplete markets model to study the optimal quantity of capital and debt. The government commits itself ex-ante to a tax schedule and government debt. In contrast to most of the existing literature these instruments are chosen to maximize agents' discounted present value of lifetime utility.

Whereas the literature mainly focuses on characterizing the steady state which maximizes welfare, we characterize and compute the optimal policy along the full transition path. In particular our characterization takes into account that the optimal long-run policy depends on capital, debt and taxation during the transition path.

We show theoretically that it is optimal to equalize the pre-tax return on capital and the rate of time preference in the long-run, i.e. the capital stock satisfies the modified golden-rule.

Quantitatively we find that the tax on capital is around 3 percent in the long-run. Labor is taxed at a much higher rate where the precise number depends on the labor supply elasticity. For standard choices for this elasticity we find a labor tax rate of almost 40 percent to be optimal in the long-run. The reason for such a high tax rate on labor income is that labor income is risky. Taxing this risky income and redistributing it back through lump-sum transfers improves ex-ante welfare in the long-run.

Transfers and the optimal level of debt along the transition are chosen to equalize the amount of redistribution over time. Initially capital is taxed higher than in the long-run since it is inelastically supplied whereas labor is taxed less than in steady state.

*We thank seminar participants at the University of Oslo for helpful comments.

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Keywords: Optimal Debt, Incomplete Markets, Capital Taxation

JEL codes:

1 Introduction

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Whereas the literature mainly focuses on characterizing the steady state which maximizes welfare, we characterize and compute the optimal policy along the full transition path. In particular our characterization takes into account that the optimal long-run policy depends on capital, debt and taxation during the transition path.

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2 The Model

2.1 Households' Problem

Following Acikgoz 2014, we formulate a Household's problem as:

$$V^H(a_0; \bar{r}, \bar{w}) = \max_{\{a_{t+1}(h^t), c_t(h^t)\}} \sum_{t=0}^{\infty} \beta^t \sum_{h^t} \Pi(h^t) u(c_t(h^t), n_t(h^t))$$

subject to

$$\begin{aligned} c_t(h^t) + a_{t+1}(h^t) &\leq a_t(h^{t-1})(1 + \bar{r}_t) + \bar{w}_t e_t(h^t) n_t(h^t) + T_t, \\ a_{t+1}(h^t) &\geq -\underline{a}. \end{aligned}$$

We use the utility function $u(c, n) = \frac{c^{1-\sigma}}{1-\sigma} - \chi \frac{n^{1+\phi}}{1+\phi}$, so a household's labor supply can be expressed as

$$\begin{aligned}
u_c(c_t, n_t) e_t \bar{w}_t + u_n(c_t, n_t) &= 0 \Rightarrow \\
\frac{-u_n(c_t, n_t)}{u_c(c_t, n_t)} &= e_t \bar{w}_t \Rightarrow \\
\frac{\chi n_t^\phi}{c_t^{-\sigma}} &= e_t \bar{w}_t \Rightarrow \\
n_t &= \left(\chi^{-1} e_t \bar{w}_t c_t^{-\sigma} \right)^{\frac{1}{\phi}}, \\
y_t &= \left(\chi^{-1} e_t^{1+\phi} \bar{w}_t^{1+\phi} c_t^{-\sigma} \right)^{\frac{1}{\phi}}.
\end{aligned}$$

Moreover,

$$e_t w_t u_{ct} + u_{nt} = 0$$

will be a useful expression to simplify expressions later. Keeping the expressions of n and y in mind, the households' policy functions solve the following system of necessary conditions

$$\begin{aligned}
u'(c_t(h^t)) &\geq \beta (1 + \bar{r}_{t+1}) \sum_{h^{t+1}} \Pi(h^{t+1}|h^t) u'(c_{t+1}(h^{t+1})), \\
0 &= (a_{t+1}(h^t) + \underline{a}) \left(u'(c_t(h^t)) - \beta (1 + \bar{r}_{t+1}) \sum_{h^{t+1}} \Pi(h^{t+1}|h^t) u'(c_{t+1}(h^{t+1})) \right) \\
c_t(h^t) + a_{t+1}(h^t) &\leq a_t(h^{t-1}) (1 + \bar{r}_t) + y_t \\
a_{t+1}(h^t) + \underline{a} &\geq 0.
\end{aligned}$$

2.2 Ramsey Problem

The Ramsey problem in Aiyagari (1994) can be formed as

$$V(a_0, B_0) = \max_{\{\bar{r}_t, \bar{w}_t, B_{t+1}, T_t, a_{t+1}(h^t), c_t(h^t)\}} \sum_{t=0}^{\infty} \beta^t \sum_{h^t} \Pi(h^t) u(c_t(h^t), n_t(h^t))$$

subject to

$$c_t(h^t) + a_{t+1}(h^t) \leq a_t(h^{t-1})(1 + \bar{r}_t) + y_t(h^t) + T_t$$

$$u'(c_t(h^t)) \geq \beta(1 + \bar{r}_{t+1}) \sum_{h^{t+1}} \Pi(h^{t+1}|h^t) u'(c_{t+1}(h^{t+1})) \quad (1)$$

$$0 = (a_{t+1}(h^t) + \underline{a}) \quad (2)$$

$$(u'(c_t(h^t)) - \beta(1 + \bar{r}_{t+1}) \sum_{h^{t+1}} \Pi(h^{t+1}|h^t) u'(c_{t+1}(h^{t+1}))) \quad (3)$$

$$a_{t+1}(h^t) + \underline{a} \geq 0,$$

$$G_t + T_t + (1 + \bar{r}_t) B_t + \bar{r}_t K_t + \bar{w}_t N_t \leq F(K_t, N_t) + B_{t+1},$$

$$K_{t+1} = \sum_{h^t} \Pi(h^t) a_{t+1}(h^t) - B_{t+1},$$

$$N_t = \sum_{h^t} \Pi(h^t) e_t n_t(h^t, \bar{w}_t, c_t).$$

Let $\beta^t \Pi(h^t) \theta_{t+1}(h^t)$ and $\beta^t \Pi(h^t) \eta_{t+1}(h^t)$ represent the Lagrange multipliers for (1) and (3) respectively, and define $\lambda_{t+1}(h^t) \equiv \eta_{t+1}(h^t) (a_{t+1}(h^t) + \underline{a}) - \theta_{t+1}(h^t)$, then we can write down the Lagrangian as

$$\begin{aligned} L &= \sum_{t=0}^{\infty} \beta^t \sum_{h^t} \Pi(h^t) u(c_t(h^t)) \\ &+ \sum_{t=0}^{\infty} \beta^t \sum_{h^t} \Pi(h^t) \theta_{t+1}(h^t) \left(u'(c_t(h^t)) - \beta(1 + \bar{r}_{t+1}) \sum_{h^{t+1}} \Pi(h^{t+1}|h^t) u'(c_{t+1}(h^{t+1})) \right) \\ &- \sum_{t=0}^{\infty} \beta^t \sum_{h^t} \Pi(h^t) \eta_{t+1}(h^t) (a_{t+1}(h^t) + \underline{a}) \\ &\left(u'(c_t(h^t)) - \beta(1 + \bar{r}_{t+1}) \sum_{h^{t+1}} \Pi(h^{t+1}|h^t) u'(c_{t+1}(h^{t+1})) \right) \\ &= \sum_{t=0}^{\infty} \beta^t \sum_{h^t} \Pi(h^t) [u(c_t(h^t)) + u'(c_t(h^t)) \\ &[\theta_{t+1}(h^t) - \theta_t(h^{t-1})(1 + \bar{r}_t) - \eta_{t+1}(h^t)(a_{t+1}(h^t) + \underline{a}) + \eta_t(h^{t-1})(a_t(h^{t-1}) + \underline{a})(1 + \bar{r}_t)]] \\ &= \sum_{t=0}^{\infty} \beta^t \sum_{h^t} \Pi(h^t) [u(c_t(h^t)) + u'(c_t(h^t)) [\lambda_t(h^{t-1})(1 + \bar{r}_t) - \lambda_{t+1}(h^t)]] \end{aligned}$$

subject to

$$\begin{aligned}
c_t(h^t) + a_{t+1}(h^t) &\leq a_t(h^{t-1})(1 + \bar{r}_t) + y(h^t, \bar{w}_t, c_t) + T_t, \\
a_{t+1}(h^t) + \underline{a} &\geq 0, \\
G_t + T_t + (1 + \bar{r}_t)B_t + \bar{r}_tK_t + \bar{w}_tN_t &\leq F(K_t, N_t) + B_{t+1}, \\
K_{t+1} &= \sum_{h^t} \Pi(h^t) a_{t+1}(h^t) - B_{t+1}, \\
N_t &= \sum_{h^t} \Pi(h^t) e_t n_t(h^t, \bar{w}_t).
\end{aligned}$$

given the known forms of $y(h^t)$ and $n(h^t)$, with initial conditions $a_0(h^{-1}) = a_0, B_0$ and $\lambda_0(h^{-1}) = 0$.

2.3 Recursive Form

To write the problem in the recursive form, follow Marcet (2011) to expand the state space to include Lagrange multipliers of the dynamic implementability constraints to recover stationarity. Index all households by $(s, e) \equiv (a, \lambda, e)$ and denote μ as the corresponding probability measure. Then we have $V(a_0, B_0) = W(\mu_0, B_0)$ which solves:

$$W(\mu, B) = \min_{\theta'(\cdot), \eta'(\cdot) \geq 0, \bar{r}, \bar{w}, T, B', a'(\cdot), c(\cdot)} \max_{\sum_e \int u(c(\cdot), n(\cdot)) + u_c(c(\cdot)) [\lambda(1 + \bar{r}) - \lambda'(\cdot)] \mu(ds, e) + \beta W(\mu', B')}$$

subject to

$$\begin{aligned}
c(\cdot) + a'(\cdot) &\leq a(1 + \bar{r}) + y(\cdot) + T & (4) \\
a' + \underline{a} &\geq 0 \\
G + T + (1 + \bar{r})B + \bar{r}K + \bar{w}N &\leq F(K, N) + B' \\
K &= \sum_e \int a \mu(ds, e) - B \\
N &= \sum_e \int e n(\cdot) \mu(ds, e) = \sum_e \pi_e e n(e, \bar{w}) \\
\mu'(S', e') &= \sum_e \pi_{ee'} \int I[(a'(\cdot), \lambda'(\cdot)) \in S'] \mu(ds, e) \\
\lambda'(\cdot) &= \eta'(\cdot)(a'(\cdot) + \underline{a}) - \theta'.
\end{aligned}$$

2.4 Interior Solution

Denote the multiplier for government budget constraint as γ . Then we treat c as a function of other control variables, which requires substituting the expressions of n and y :

$$\begin{aligned} c + a' &= a(1 + \bar{r}) + y + T \Rightarrow \\ c + a' &= a(1 + \bar{r}) + (\chi^{-1} \bar{w}^{1+\phi} e^{1+\phi} c^{-\sigma})^{\frac{1}{\phi}} + T. \end{aligned}$$

We can get the following useful expressions on how n and y respond to c , which of course respond to other choice variable $x \in \{T, B', \bar{r}, \bar{w}\}$,

$$\begin{aligned} \frac{\partial n}{\partial c} &= -\frac{\sigma}{\phi} (\chi^{-1} \bar{w} e)^{\frac{1}{\phi}} c^{-\frac{\sigma}{\phi}-1} \\ &= -\frac{\sigma n}{\phi c}, \\ \frac{\partial n}{\partial \bar{w}} &= \frac{1}{\phi} \frac{n}{\bar{w}}, \\ \frac{\partial y}{\partial c} &= -\frac{\sigma}{\phi} (\chi^{-1} \bar{w}_t^{1+\phi} e_t^{1+\phi})^{\frac{1}{\phi}} c_t^{-\frac{\sigma}{\phi}-1} \\ &= -\frac{\sigma y}{\phi c} \\ &= -\frac{\sigma n}{\phi c} e w \\ &= e w \frac{\partial n}{\partial c}, \\ \frac{\partial c}{\partial T} &= e \bar{w} \frac{\partial n}{\partial c} \frac{\partial c}{\partial T} + 1 \Rightarrow \\ \frac{\partial c}{\partial T} \left(1 - e \bar{w} \frac{\partial n}{\partial c} \right) &= 1, \\ \frac{\partial u(c, n)}{\partial T} &= u_c \frac{\partial c}{\partial T} + u_n \frac{\partial n}{\partial c} \frac{\partial c}{\partial T} \\ &= u_c \frac{\partial c}{\partial T} \left(1 + \frac{u_n}{u_c} \frac{\partial n}{\partial c} \right) \\ &= u_c \frac{\partial c}{\partial T} \left(1 - e \bar{w} \frac{\partial n}{\partial c} \right) \\ &= u_c. \end{aligned}$$

This is essentially the envelop theorem. Similarly, we can also show that

$$\begin{aligned}\frac{\partial u(c, n)}{\partial a'} &= u_c, \\ \frac{\partial u(c, n)}{\partial \bar{r}} &= au_c, \\ \frac{\partial u(c, n)}{\partial \bar{w}} &= enu_c, \\ \frac{\partial u(c, n)}{\partial a} &= (1 + \bar{r})u_c,\end{aligned}$$

It is not possible to directly solve for the expression of c , but we can derive the necessary partial derivatives to know how c responds to other control variables, including $T, a', \bar{r}, \bar{w}, a$,

as follows:

$$\begin{aligned}
\frac{\partial c}{\partial T} &= \frac{\partial y}{\partial c} \frac{\partial c}{\partial T} + 1 \Rightarrow \\
\frac{\partial c}{\partial T} &= \frac{1}{1 - \frac{\partial y}{\partial c}} \\
&= \frac{1}{1 + \frac{\sigma y}{\phi c}} \\
\frac{\partial c}{\partial a'} + 1 &= \frac{\partial y}{\partial c} \frac{\partial c}{\partial a'} \Rightarrow \\
\frac{\partial c}{\partial a'} &= \frac{1}{\frac{\partial y}{\partial c} - 1} \\
&= \frac{1}{\frac{\sigma y}{\phi c} - 1} \\
\frac{\partial c}{\partial \bar{w}} &= en + e\bar{w} \frac{\partial n}{\partial \bar{w}} + e\bar{w} \frac{\partial n}{\partial c} \frac{\partial c}{\partial \bar{w}} \Rightarrow \\
\frac{\partial c}{\partial \bar{w}} &= en + e\bar{w} \frac{1}{\phi} \frac{n}{\bar{w}} + e\bar{w} \left(-\frac{\sigma n}{\phi c} \right) \frac{\partial c}{\partial \bar{w}} \Rightarrow \\
\frac{\partial c}{\partial \bar{w}} &= \frac{1 + \phi}{\phi} \frac{en}{1 - \frac{\sigma}{\phi} e\bar{w}nc^{-1}}, \\
\frac{\partial c}{\partial \bar{r}} &= a + \frac{\partial y}{\partial c} \frac{\partial c}{\partial \bar{r}} \Rightarrow \\
\frac{\partial c}{\partial \bar{r}} &= \frac{a}{1 - \frac{\partial y}{\partial c}} \\
&= \frac{a}{1 + \frac{\sigma y}{\phi c}}, \\
\frac{\partial c}{\partial a} &= (1 + \bar{r}) + \frac{\partial y}{\partial c} \frac{\partial c}{\partial a} \Rightarrow \\
\frac{\partial c}{\partial a} &= \frac{1 + \bar{r}}{1 - \frac{\partial y}{\partial c}} \\
&= \frac{1 + \bar{r}}{1 + \frac{\sigma y}{\phi c}}.
\end{aligned}$$

Then the interior solution to the Ramsey problem satisfies the following conditions.

$$\begin{aligned}
\partial\lambda' : u_c(c) &\geq \beta(1 + \bar{r}') \mathbb{E}[u_c(c') | e] \text{ with equality if } a' > -\underline{a} \\
\partial a' : u_c(c) + \frac{\partial c}{\partial a'} u_{cc}(c) [\lambda(1 + \bar{r}) - \lambda'] &= \beta \mathbb{E} \left[(1 + \bar{r}') u_c(c') + \frac{\partial c'}{\partial a'} u_{cc}(c') [\lambda'(1 + \bar{r}') - \lambda''] | e \right] \\
&\quad + \beta \gamma' (F_K(K', N') - \bar{r}') \text{ if } a' > -\underline{a}, \text{ otherwise } \lambda' = 0 \\
\partial T : \gamma &= \sum_e \int \left(u_c(c) + \frac{\partial c}{\partial T} u_{cc}(c) [\lambda(1 + \bar{r}) - \lambda'] \right) \mu(ds, e) \\
&\quad + \gamma (F_N(K, N) - \bar{w}) \frac{\partial N}{\partial T} \\
\partial B' : \gamma &= \beta(1 + F_K(K', N')) \gamma' \\
\partial \bar{r} : \gamma A &= \sum_e \int u_c(c) \lambda \mu(ds, e) \\
&\quad + \gamma (F_N(K, N) - \bar{w}) \frac{\partial N}{\partial \bar{r}} \\
&\quad + \sum_e \int a u_c(c) + \frac{\partial c}{\partial \bar{r}} u_{cc}(c) [\lambda(1 + \bar{r}) - \lambda'] \mu(ds, e) \\
\partial \bar{w} : \gamma N &= \gamma (F_N(K, N) - \bar{w}) \frac{\partial N}{\partial \bar{w}} \\
&\quad + \sum_e \int e n(e, \bar{w}, c) u_c(c) + \frac{\partial c}{\partial \bar{w}} u_{cc}(c) [\lambda(1 + \bar{r}) - \lambda'] \mu(ds, e),
\end{aligned}$$

where

$$\begin{aligned}
\frac{\partial N}{\partial \bar{w}} &= \sum_e \int \left(\frac{\partial n}{\partial \bar{w}} + \frac{\partial n}{\partial c} \frac{\partial c}{\partial \bar{w}} \right) \mu(ds, e), \\
\frac{\partial N}{\partial T} &= \sum_e \int \frac{\partial n}{\partial c} \frac{\partial c}{\partial T} \mu(ds, e), \\
\frac{\partial N}{\partial \bar{r}} &= \sum_e \int \frac{\partial n}{\partial c} \frac{\partial c}{\partial \bar{r}} \mu(ds, e).
\end{aligned}$$

Here the unknowns are $\lambda'(\cdot), a'(\cdot), B', T, \bar{r}, \bar{w}$, while $c(\cdot), \mu(\cdot), K, N, A$ are considered as functions of these unknowns, from the household and the government budget constraint, the market clearing conditions and so on. We denote the p.d.f of the distribution of (a, λ, e) as $p(a, \lambda, e)$, while its probabilistic measure is μ . Similarly, we denote the p.d.f. of the distribution of (a, e) as $m(a, e)$. Moreover, as we know that c_t and a_{t+1} should only depend on (a_t, e_t) but not λ_t because they should satisfy the households' optimization which doesn't consider λ_t , we can use $g_{a,t+1}(a_t, e_t)$ as the policy function of a_{t+1} , $g_{c,t}(a_t, e_t)$ the policy function of c_t , and $g_{\lambda,t+1}(a_t, \lambda_t, e_t)$ the policy function of λ_{t+1} . Now all the unknowns and their related

equations can be written as follows.

$$u_c(c_t) \geq \beta(1 + \bar{r}_{t+1}) \mathbb{E}[u_c(c_t) | e_{t+1}] \text{ with equality if } a_{t+1} > -\underline{a} \quad (5)$$

$$u_c(c_t) + \frac{\partial c_t}{\partial a_{t+1}} u_{cc}(c_t) [\lambda_t(1 + \bar{r}_t) - \lambda_{t+1}] \quad (6)$$

$$= \beta \mathbb{E} \left[(1 + \bar{r}_{t+1}) u_c(c_{t+1}) + \frac{\partial c_t}{\partial a_{t+1}} u_{cc}(c_{t+1}) [\lambda_{t+1}(1 + \bar{r}_{t+1}) - \lambda_{t+2}] | e_{t+1} \right] \\ + \beta \gamma_{t+1} (F_K(K_{t+1}, N_{t+1}) - \bar{r}_{t+1}) \text{ if } a_{t+1} > -\underline{a}, \text{ otherwise } \lambda_{t+1} = 0 \quad (7)$$

$$\gamma_t = \beta(1 + F_K(K_{t+1}, N_{t+1})) \gamma_{t+1} \quad (8)$$

$$\gamma_t = \sum_e \int \int \left(u_c(c_t) + \frac{\partial c_t}{\partial T_t} u_{cc}(c_t) [\lambda_t(1 + \bar{r}_t) - \lambda_{t+1}] \right) p(a_t, \lambda_t, e_t) da_t d\lambda_t \quad (9)$$

$$+ \gamma_t (F_N(K_t, N_t) - \bar{w}_t) \frac{\partial N_t}{\partial T_t},$$

$$\gamma_t A_t = \sum_{e_t} \int \int u_c(c_t) \lambda_t p_t(a_t, \lambda_t, e_t) da_t d\lambda_t \\ + \gamma_t (F_N(K_t, N_t) - \bar{w}_t) \frac{\partial N_t}{\partial \bar{r}_t} \\ + \sum_{e_t} \int \int a_t (u_c(c_t) + u_{cc}(c_t) [\lambda_t(1 + \bar{r}_t) - \lambda_{t+1}]) p(a_t, \lambda_t, e_t) da_t d\lambda_t \quad (10)$$

$$\gamma_t N_t = \gamma_t (F_N(K_t, N_t) - \bar{w}_t) \frac{\partial N_t}{\partial \bar{w}_t} \\ + \sum_{e_t} \int \int e_t n(e_t, \bar{w}_t) (u'(c_t) + u''(c_t) [\lambda_t(1 + \bar{r}_t) - \lambda_{t+1}]) \quad (11)$$

$$p_t(a_t, \lambda_t, e_t) da_t d\lambda_t$$

subject to

$$c_t + a_{t+1} = a_t (1 + \bar{r}_t) + y(e_t, \bar{w}_t) \quad (12)$$

$$G_t + (1 + \bar{r}_t) B_t + \bar{r}_t K_t + \bar{w}_t N_t \leq F(K_t, N_t) + B_{t+1} \quad (13)$$

$$K_t = A_t - B_t \quad (14)$$

$$A_t = \sum_{e_t} \int \int a_t p_t(a_t, \lambda_t, e_t) da_t d\lambda_t \quad (15)$$

$$N_t = \sum_{e_t} \pi(e_t) e_t n(e_t, \bar{w}_t) \quad (16)$$

$$p_{t+1}(a_{t+1}, \lambda_{t+1}, e_{t+1}) = \sum_{e_t} \pi(e_{t+1}|e_t) \int \int I[g_{a,t+1}(a_t, e_t) = a_{t+1}, g_{\lambda,t+1}(a_t, \lambda_t, e_t) = \lambda_{t+1}] \quad (17)$$

$$m_{t+1}(a_{t+1}, e_{t+1}) = \sum_{e_t} \pi(e_{t+1}|e_t) \int I[g_{a,t+1}(a_t, e_t) = a_{t+1}] m_t(a_t, e_t) da_t \quad (18)$$

$$\frac{\partial N_t}{\partial T_t} = \sum_e \int \int \frac{\partial n_t}{\partial c_t} \frac{\partial c_t}{\partial T_t} p_t(a_t, \lambda_t, e_t) da_t d\lambda_t \quad (19)$$

$$\frac{\partial N_t}{\partial \bar{r}_t} = \sum_e \int \int \frac{\partial n_t}{\partial c_t} \frac{\partial c_t}{\partial \bar{r}_t} p_t(a_t, \lambda_t, e_t) da_t d\lambda_t \quad (20)$$

$$\frac{\partial N_t}{\partial \bar{w}_t} = \sum_e \int \int \left(\frac{\partial n_t}{\partial \bar{w}_t} + \frac{\partial n_t}{\partial c_t} \frac{\partial c_t}{\partial \bar{w}_t} \right) p_t(a_t, \lambda_t, e_t) da_t d\lambda_t. \quad (21)$$

Now there are 12 sets of unknowns $g_{a,t+1}(a_t, e_t)$, $g_{\lambda,t+1}(a_t, \lambda_t, e_t)$, $g_{c,t}(a_t, e_t)$, $p_t(a_t, \lambda_t, e_t)$, $m_t(a_t, e_t)$ and $\gamma_t, \bar{r}_t, \bar{w}_t, K_t, N_t, A_t, B_t, T_t$. We have 12 sets of equations to solve them.

3 The Algorithm

3.1 Steady State

Here we illustrate the algorithm to compute the steady state using a simplified example as in Aiyagari (1994) and Acikgoz (2014) where the labor supply depends on the tradeoff between the market wage and home production. In the steady state, we can use the knowledge that all the variables $\bar{r}, \bar{w}, \gamma, K, N, B$ are constant and all the policy functions do not change over

time, to simplify the set of equations that we need to solve as follows:

$$u'(c) \geq \beta(1 + \bar{r}) \mathbb{E}[u'(c') | e] \text{ with equality if } a' > -\underline{a} \quad (22)$$

$$u'(c) + u''(c) [\lambda(1 + \bar{r}) - \lambda'] = \beta(1 + \bar{r}) \mathbb{E}[u'(c') + u''(c') [\lambda'(1 + \bar{r}') - \lambda''] | e] \\ + \beta\gamma(F_K(K, N) - \bar{r}) \text{ if } a' > -\underline{a}, \text{ otherwise } \lambda' = 0 \quad (23)$$

$$1 = \beta(1 + F_K(K, N)) \quad (24)$$

$$\gamma = \sum_e \int (u'(c) + u''(c) [\lambda(1 + \bar{r}) - \lambda']) \mu(ds, e) \quad (25)$$

$$\gamma A = \sum_e \int u'(c) \lambda \mu(ds, e) \\ + \sum_e \int a(u'(c) + u''(c) [\lambda(1 + \bar{r}) - \lambda']) \mu(ds, e) \quad (26)$$

$$\gamma N = \gamma(F_N(K, N) - \bar{w}) N'(\bar{w}) \\ + \sum_e \int en(e, \bar{w}) (u'(c) + u''(c) [\lambda(1 + \bar{r}) - \lambda']) \mu(ds, e) \quad (27)$$

$$c + a' = a(1 + \bar{r}) + y(e, \bar{w}) + T \quad (28)$$

$$G + T + \bar{r}B + \bar{r}K + \bar{w}N \leq F(K, N) \quad (29)$$

$$K = A - B \quad (30)$$

$$N = \sum_e \pi_e en(e, \bar{w}) \quad (31)$$

$$A = \sum_e \int a \mu(ds, e) \quad (32)$$

$$p(a', \lambda', e') = \sum_e \pi_{ee'} \int I[g_{a'}(a, e) = a', g_{\lambda'}(a, \lambda, e) = \lambda'] p(a, \lambda, e) da d\lambda \quad (33)$$

$$m(a', e') = \sum_e \pi_{ee'} \int I[g_{a'}(a, e) = a'] m(a, e) da. \quad (34)$$

1. Guess T .

2. Guess \bar{w} . Solve for $\bar{r}(w)$ following Aiyagari (1994)

(a) Solve for K from (24).

(b) Guess \bar{r} and solve the household's problem: solve for $c = g_c(a, e)$, $a' = g_{a'}(a, e)$ from (22) and (28), keeping in mind that $n = (\chi^{-1} \bar{w} e c^{-\sigma})^{\frac{1}{\phi}}$, $y = (\chi^{-1} \bar{w}^{1+\phi} e^{1+\phi} c^{-\sigma})^{\frac{1}{\phi}}$.

(c) Compute N from (31).

- (d) Solve for $m(a, e)$ or equivalently $m(\cdot, e)$ from (34).
- (e) Solve for A from (32).
- (f) Solve for B from (30).
- (g) Verify \bar{r} using (29). If the equation is not satisfied, update \bar{r} .
3. Define $q \equiv \lambda/\gamma$, and solve for $q' = g_{q'}(a, q, e)$ from (23).
- (a) Approach 1. Guess $q'(a, q, e) = g_{q'}^0(a, q, e)$, and then use equation 23 to find the new $q'(a, q, e) = g_{q'}^1(a, q, e)$.
- (b) Approach 2. Guess $q' = g_{q'}(a, q, e) = \alpha_0(a, e)q + \frac{\alpha_1(a, e)}{(1+\bar{r})u''(e)}$ and solve for $\alpha_0(a, e)$ and $\alpha_1(a, e)$, as in Acikgoz (2014).
4. Solve for γ from (26)
5. Check whether (27) is satisfied.
- (a) Simplify the expression of N as

$$N = (F_N(K, N) - \bar{w})N'(\bar{w}) + \frac{1}{\gamma} \sum_e \int en(e, \bar{w})u'(c)p(a, \lambda, e)dad\lambda$$

$$+ \sum_e \int en(e, \bar{w})u''(c)[q(1+\bar{r}) - q']p(a, \lambda, e)dad\lambda \quad (35)$$

$$= (F_N(K, N) - \bar{w})N'(\bar{w}) + \frac{1}{\gamma} \sum_e \int en(e, \bar{w})u'(c)p_q(a, q, e)dadq \quad (36)$$

$$+ \sum_e \int en(e, \bar{w})u''(c)[q(1+\bar{r}) - q']p_q(a, q, e)dadq$$

$$= (F_N(K, N) - \bar{w})N'(\bar{w}) + \frac{1}{\gamma} \sum_e \int en(e, \bar{w})u'(c)m(a, e)da \quad (37)$$

$$+ \sum_e \int en(e, \bar{w})u''(c)(1+\bar{r})\mathbb{E}[q|a, e]m(a, e)da \quad (38)$$

$$- \sum_e \int en(e, \bar{w})u''(c)\mathbb{E}[q'|a, e]m(a, e)da. \quad (39)$$

- (b) If equation (39) holds, stop.
- (c) If not, update \bar{w} .
- i. If LHS of (39) $>$ RHS, adjust \bar{w} down.
- ii. If LHS $<$ RHS, adjust \bar{w} up.

- iii. One potential new candidate of \bar{w} can be computed as follows: use RHS of (39) to represent the new N to backout the new \bar{w} from (31).

6. Verify T using (25).

3.2 Transition Dynamics

We consider three approaches to solve the transition dynamics. The first one is to guess the distribution of households on a, λ, e and then update it. The second is to guess the price sequences and then to iterate on the price sequences, similar to how transition dynamics of Aiyagari model without taxes is solved. The last one is to use backward induction from the steady state. Let us first consider the transition dynamics without transfer.

3.2.1 Guess the Distribution

Main idea

- Guess the distribution $p_t(a_t, \lambda_t, e_t)$
- Solve from backward for policy functions and Lagrangian multipliers using the guessed p_t
- Compute forward the new distribution, starting from $p_0(a_t, \lambda_t, e_t)$
- Iterate until $p_t(a_t, \lambda_t, e_t)$ converge

Steps

1. Set all variables after T at the steady state
2. Guess $\{p_t(a_t, \lambda_t, e_t)\}_{t=1}^{T-1}$
3. Starting from $t = T - 1$, solve for variables at t using variables from $t + 1$ on. We use guess and verify to find \bar{w}_t and then other variables at t .
 - (a) Compute γ_t from (8).
 - (b) Guess \bar{w}_t .
 - (c) Guess \bar{r}_t .
 - (d) Solve for $c_t = g_{c,t}(a_t, e_t)$ and $a_{t+1} = g_{a,t+1}(a_t, e_t)$ from (5) and (12).

- (e) Solve for $\lambda_{t+1} = g_{\lambda,t+1}(a_t, \lambda_t, e_t)$ from (7).
 - (f) Compute A_t and N_t from (15) and (16).
 - (g) Check whether \bar{r}_t satisfies (10). Otherwise update on \bar{r}_t
 - (h) Solve for K_t and B_t from (13) and (14).
 - (i) Verify (11) and update \bar{w}_t .
 - i. If LHS of (11) $>$ RHS, reduce \bar{w}_t as the new guess, and vice versa.
 - ii. A possible choice of the new guess of \bar{w}_t is to use the RHS of (11) to get a new N_t and then back out the new guess of \bar{w}_t using (16).
4. Continue the backward induction until $t = 0$.
 5. Starting from $p_0(a_0, \lambda_0, e_t)$, compute the new $p_t(a_t, \lambda_t, e_t)$ using the new policy functions $a_{t+1} = g_{a,t+1}(a_t, e_t)$ and $\lambda_{t+1} = g_{\lambda,t+1}(a_t, \lambda_t, e_t)$.
 6. Iterate until $p_t(a_t, \lambda_t, e_t)$ converge for each t .

3.2.2 Guess the Price Sequences and Iterate on Them

Main idea

- Guess $\{\bar{w}_t, \bar{r}_t\}_{t=0}^{\infty}$
- Solve the households' problem on $a_{t+1}(a_t, e_t)$ from backward
- Compute $m_t(a_t, e_t), A_t, B_t, K_t$ from forward
- Solve for γ_t and $\lambda_{t+1}(a_t, e_t)$ from backward
- Compute $p_t(a_t, \lambda_t, e_t)$ from forward
- Check and update $\{\bar{w}_t, \bar{r}_t\}_{t=0}^{\infty}$

The steps are the following:

1. Guess $\{\bar{w}_t, \bar{r}_t\}_{t=0}^{\infty}$
2. Solve households' problem by backward induction, as in Aiyagari model
 - (a) Compute $n(e_t, \bar{w}_t)$ and $y(e_t, \bar{w}_t)$

- (b) Solve for $a_{t+1}(a_t, e_t)$ by backward induction from from (5) and (12).
3. Compute $m_t(a_t, e_t)$, starting from $m_0(a_0, e_0)$ using (18) or simulation.
 4. Compute A_t and N_t from (15) and (16).
 5. Compute K_t and B_{t+1} forward using (14) and (13), namely,

$$\begin{aligned} K_t &= A_t - B_t \\ B_{t+1} &= F(K_t, N_t) - G_t + (1 + \bar{r}_t) B_t + \bar{r}_t K_t + \bar{w}_t N_t. \end{aligned}$$

6. Compute γ_t backward using

$$\gamma_t = \beta (1 + F_K(K_{t+1}, N_{t+1})) \gamma_{t+1}$$

7. Solve for $\lambda_{t+1}(a_t, \lambda_t, e_t)$ from

$$\begin{aligned} &u'(c_t) + u''(c_t) [\lambda_t (1 + \bar{r}_t) - \lambda_{t+1}] \\ &= \beta (1 + \bar{r}_{t+1}) \mathbb{E}[u'(c_{t+1}) + u''(c_{t+1}) [\lambda_{t+1} (1 + \bar{r}_{t+1}) - \lambda_{t+2}] | e_{t+1}] \\ &+ \beta \gamma_{t+1} (F_K(K_{t+1}, N_{t+1}) - \bar{r}_{t+1}) \\ &\text{if } a_{t+1} > -\underline{a}, \text{ otherwise } \lambda_{t+1} = 0. \end{aligned}$$

Notice that λ_{t+2} is a function of λ_{t+1} , namely, $\lambda_{t+2} = \lambda_{t+2}(a_{t+1}, \lambda_{t+1}, e_{t+1})$.

8. Compute p_t forward by simulaitons using p_0 and the policy functions: $a_{t+1}(a_t, e_t)$ and $\lambda_{t+1}(a_t, \lambda_t, e_t)$.
9. Check whether the guessed $\{\bar{w}_t, \bar{r}_t\}_{t=0}^{\infty}$ is the solution, and update if not.

- (a) Check the equation

$$\begin{aligned} \gamma_t N_t &= \gamma_t (F_N(K_t, N_t) - \bar{w}_t) N'(\bar{w}_t) \\ &+ \sum_{e_t} \int \int e_t n(e_t, \bar{w}_t) (u'(c_t) + u''(c_t) [\lambda_t (1 + \bar{r}_t) - \lambda_{t+1}]). \end{aligned}$$

If it is not satisfied, update \bar{w}_t , for example by solving

$$\begin{aligned}\gamma_t N_t (\bar{w}_t^{new}) &= \gamma_t (F_N (K_t, N_t) - \bar{w}_t) N' (\bar{w}_t) \\ &+ \sum_{e_t} \int \int e_t n (e_t, \bar{w}_t) (u' (c_t) + u'' (c_t) [\lambda_t (1 + \bar{r}_t) - \lambda_{t+1}]).\end{aligned}$$

(b) Check the equation

$$\begin{aligned}\gamma_t A_t &= \sum_{e_t} \int \int u' (c_t) \lambda_t p_t (a_t, \lambda_t, e_t) da_t d\lambda_t \\ &+ \sum_{e_t} \int \int a_t (u' (c_t) + u'' (c_t) [\lambda_t (1 + \bar{r}_t) - \lambda_{t+1}]) p (a_t, \lambda_t, e_t) da_t d\lambda_t\end{aligned}$$

If it is not satisfied, update \bar{r}_t , for example by solving \bar{r}_t^{new} from the above equation.

3.2.3 Backward Induction

Knowing the steady state, we can solve for the transition dynamics, given the initial bond B_0 and the initial asset and productivity distribution $m_0 (a, e)$, by backward induction, with the following steps.

1. Set all variables After $T + 1$ at the steady state levels.
2. At T , all variables except B_T and p_T are at the steady state levels. Guess B_T and p_T .
3. Solve for variables at t using variables from $t + 1$ on. We use guess and verify to find \bar{w}_t and then other variables at t .
 - (a) Compute γ_t from (8).
 - (b) Guess \bar{w}_t .
 - (c) Guess \bar{r}_t .
 - (d) Solve for $c_t = g_{c,t} (a_t, e_t)$ and $a_{t+1} = g_{a,t+1} (a_t, e_t)$ from (5) and (12).
 - (e) Solve for $\lambda_{t+1} = g_{\lambda,t+1} (a_t, \lambda_t, e_t)$ from (7).
 - (f) Solve for $p_t (a_t, \lambda_t, e_t)$ from (17). One choice is to solve in two steps.
 - i. Solve for $m_t (a_t, e_t)$ from (18), using the knowledge of $g_{a,t+1} (a_t, e_t)$, $\pi (e_{t+1}|e_t)$ and $m_{t+1} (a_{t+1}, e_{t+1})$.

- ii. Solve for $p_t(\lambda_t|a_t, e_t)$ from (17), using the knowledge of $g_{\lambda,t+1}(a_t, \lambda_t, e_t)$, $p_{t+1}(\lambda_{t+1}|a_{t+1}, e_{t+1})$ and $m_t(a_t, e_t)$.
- (g) Compute A_t and N_t from (15) and (16).
- (h) Check whether \bar{r}_t satisfies (10). Otherwise update on \bar{r}_t
- (i) Solve for K_t and B_t from (13) and (14). We can substitute $B_t = A_t - K_t$ into (13) and get

$$G_t + (1 + \bar{r}_t) A_t - K_t + \bar{w} N_t \leq F(K_t, N_t) + B_{t+1}$$

which leads to a unique solution of K_t .

- (j) Verify (11) and update \bar{w}_t .
 - i. If LHS of (11) $>$ RHS, reduce \bar{w}_t as the new guess, and vice versa.
 - ii. A possible choice of the new guess of \bar{w}_t is to use the RHS of (11) to get a new N_t and then back out the new guess of \bar{w}_t using (16).
4. Continue the backward induction until $t = 0$.
5. Check whether $p_0(\lambda, a, e)$ satisfies the initial conditions
- (a) $m_0(a, e) = \int p_0(\lambda, a, e) d\lambda$ equals the initial asset and productivity distribution
 - (b) $p_0(\lambda|a, e)$ is degenerate distribution at $\lambda = 0$
 - (c) If not, update the guess of B_T and p_T . One possible way to get the new B_T and p_T is to compute the transition forward from the real B_0 and p_0 , with the policy functions calculated. Then we will get a new sequence of p_t and B_t , which give us a new guess of p_T and B_T .