

# “Quantitative Macroeconomics & Social Insurance - TA 3”

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# 1 Consumption, Saving, and Portfolio Choice

Consider the household problem

$$\begin{aligned} v_t(w_t) &= \max_{c_t, \alpha_t, w_{t+1}} \{u(c_t) + \beta \mathbb{E}_t v_{t+1}(w_{t+1})\}, \\ \text{s.t. } w_{t+1} &= (w_t - c_t) R_{t+1}^p, \\ R_{t+1}^p &= R^f + \alpha_t (R_{t+1} - R^f), \quad \ln R_{t+1} \sim N(\mu, \sigma^2) \\ w_0 &\text{ given, } w_{T+1} = 0. \end{aligned}$$

- Finite-horizon  $t = 0, 1, 2, \dots, T$ .
- Non-tradable asset  $h_t$  equal to present value of deterministic income  $\{y_t\}$
- Total wealth  $w_t$  is the sum of cash-on-hand  $x_t = a_t R_{t+1}^p + y_t$  and  $h_t$
- Natural debt limit  $a_t \geq -y_t - h_t$  if binding then  $c_t = 0, \forall t$

## 1. Life cycle numerical examples in partial equilibrium.

- Individuals start to work at age 20, retire at 65, and exit the model at 80.
- Solve the problem with VFI or by guess and verify (see lecture notes)
- The portfolio choice reduces to  $\max_{\alpha_t} g(R_{t+1}^p)$ 
  - Exploit log-normality to rewrite  $g$ , approximate  $R^p$ , solve FOC.
- Solution:

$$v_t = (1 - \theta)^{-1} m_t^{-\theta} w_t^{1-\theta},$$

$$c_t = m_t w_t,$$

$$m_t = (1 + b_t)^{-1}, \quad \forall t \in [0, T - 1], m_T = 1,$$

$$b_t = [\beta \mathbf{E}_t(R_{t+1}^p(\alpha)^{1-\theta}) m_{t+1}^{-\theta}]^{1/\theta},$$

$$\alpha = \frac{\ln(1 + \mu) - \ln R^f + \sigma^2/2}{\theta \sigma^2}.$$

Set up:

- Time  $s = 1, \dots, 46, \dots, 61$  and initial condition at  $s = 1$
- Fix parameters values and the income stream  $(y_0, \dots, y_T)$

Implementation:

1. Compute policy  $\alpha$ , and also  $m_t$  solving backward  $b_t$  given  $m_T = 1$
2. Solve backward  $h_{t+1} = Rh_t - y_{t+1}$  given  $h_T = 0$
3. Simulate  $(\ln R_0, \dots, \ln R_T)$  from  $\ln(1 + \mu) + \sigma\epsilon$ ,  $\epsilon \sim N(0, 1)$
4. Solve forward budget constraint given  $R_{t+1}^p$  and  $w_0 = c_0 + h_0$ ,  $c_0 = 1$ 
  - $c_t = m_t w_t$ ,  $x_t = w_t - h_t$
  - $s_t = w_t - c_t$ ,  $s_t^f = x_t - c_t$ ,  $\hat{\alpha}_t = \alpha(s_t/s_t^f)$ ,  $w_{t+1} = s_t R_{t+1}^p$
5. Simulate  $N$  times and compute aggregates  $c_t = N^{-1} \sum_{i=1}^N c_t^i$

## 2. Consumption over the life-cycle.

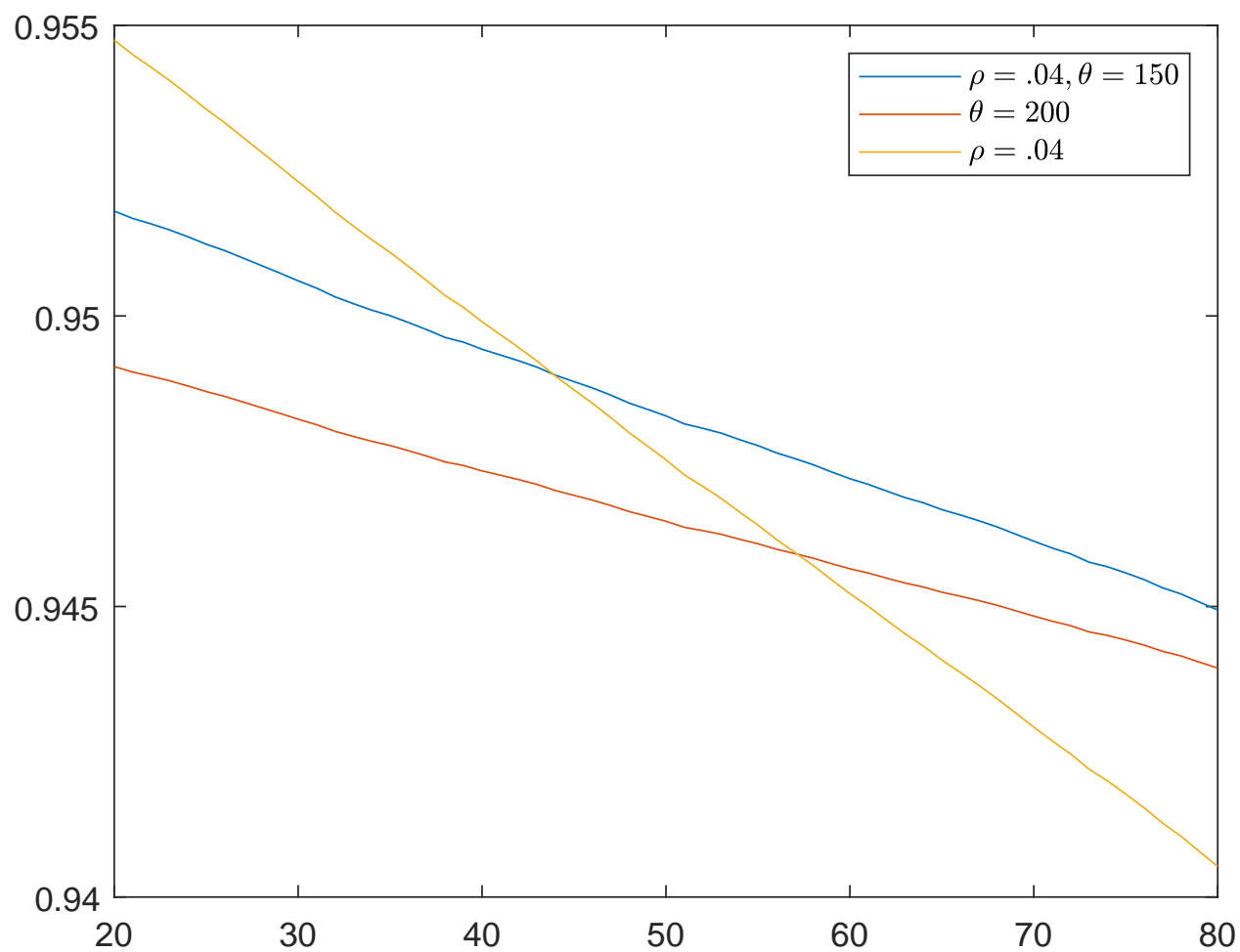


Figure 1: Average consumption

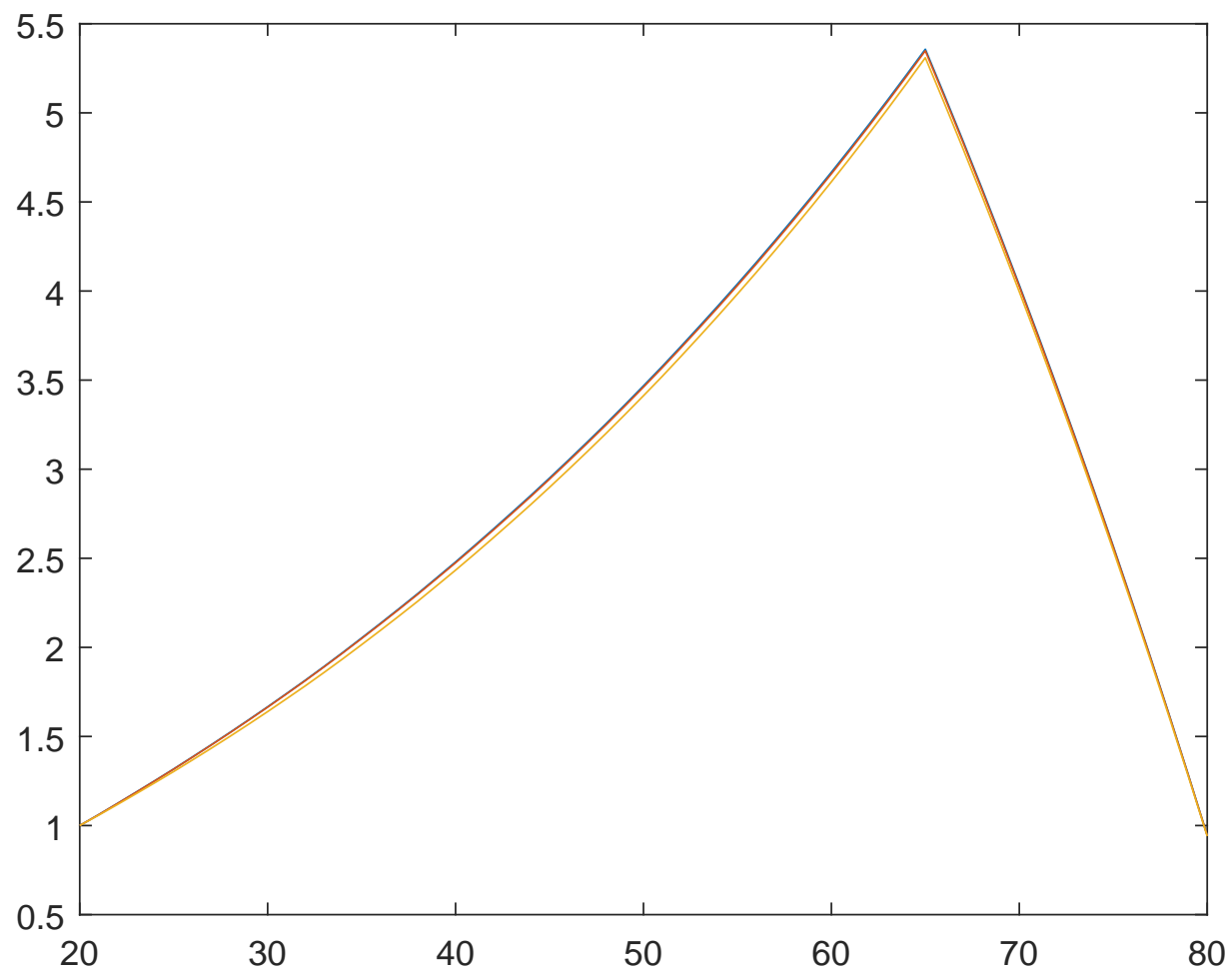


Figure 2: Average cash on hand

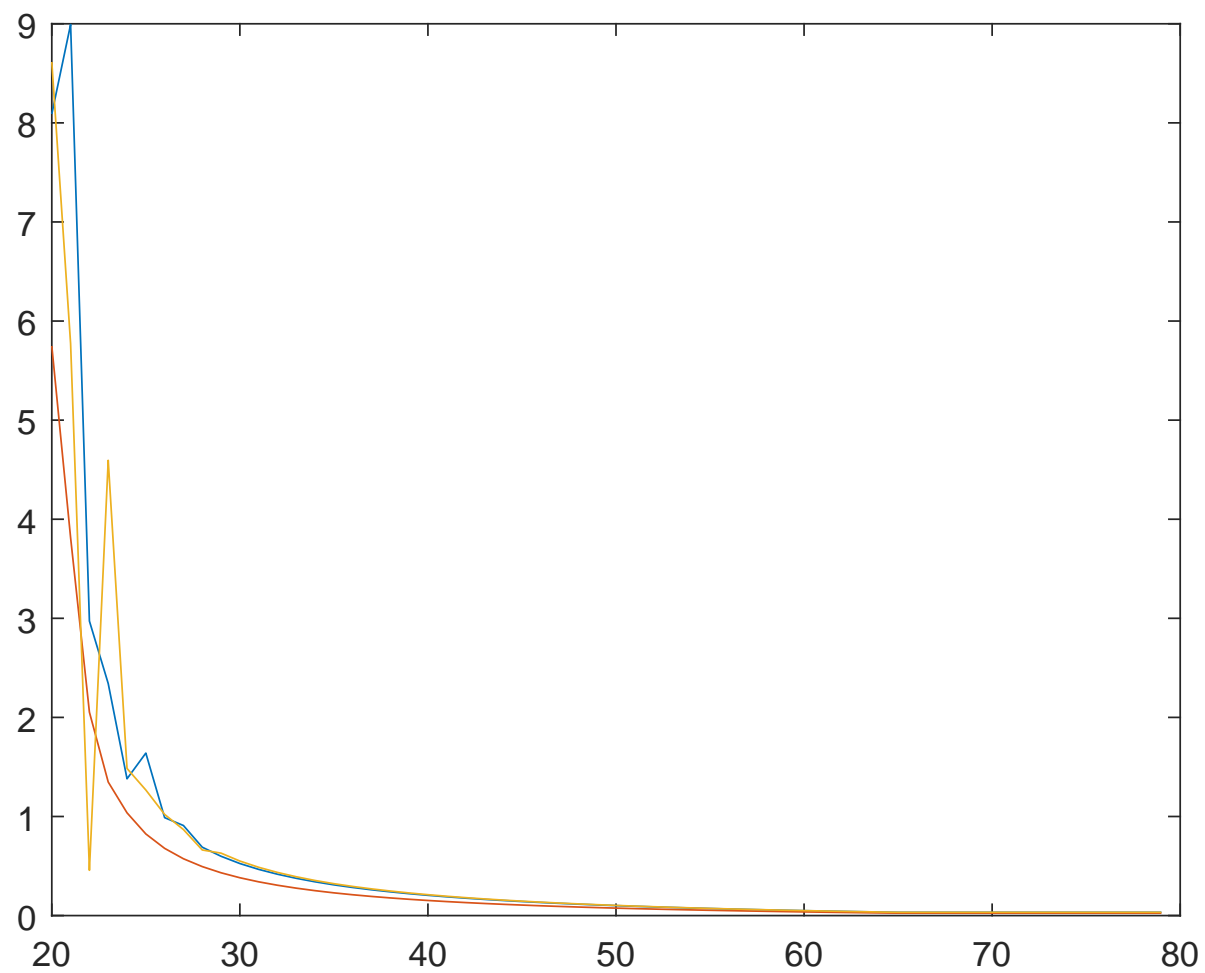


Figure 3: Average risky share

## Remarks:

- $\hat{\alpha}_t = \alpha(1 + h_t/s_t^f)$  where  $h$  is decreasing and  $s^f$  increasing over time
- Since  $\downarrow h_t, \uparrow$  risk-free asset,  $\alpha_t$  is decreasing over time
- Borrow from future income to invest in risky asset  $\alpha_t > 1$
- Consumption decreases over time due to the precautionary saving motive.
- In the data consumption is hump-shaped with a peak around forty/fifty.
- Literature often set  $\theta$  around 2. Using 150 lead to much less volatility
- $\uparrow \theta$  (high RRA, lower IES),  $\downarrow \alpha_t$ , lower and less steep consumption profile.
- $\uparrow \rho$  (more impatience), steeper consumption profile, almost no effect on  $\alpha_t$ .



### 3. Rule of thumb portfolio.

$$\bar{a}_j = 1 - \frac{j}{100},$$

$$\bar{a}_{20} = .8,$$

$$\bar{a}_{65} = .2,$$

$$\bar{a}_{80} = .35.$$

Compared to Figure 3 this allocation is clearly suboptimal:

- Very flat over life-cycle, below 1 when young and above 0 when old.

## 4. Epstein-Zin-Weil (EZW) preferences

Remarks:

- Typical RRA estimates put  $\theta$  around 2 and IES  $\eta$  below 1
- However, some studies find very low IES, but then  $\theta = \eta^{-1} = 10$

Consider the household problem

$$v_t(w_t) = \max_{c_t, \alpha_t, w_{t+1}} \left\{ (1 - \beta) c_t^{\frac{1-\theta}{\gamma}} + \beta \mathbf{E}_t(v_{t+1}(w_{t+1})^{1-\theta})^{\frac{1}{\gamma}} \right\}^{\frac{\gamma}{1-\theta}},$$
$$\text{s.t. } w_{t+1} = (w_t - c_t) R_{t+1}^p,$$

where

$$\gamma = \frac{1 - \theta}{1 - \frac{1}{\psi}},$$

$\theta$  is the relative risk aversion coefficient and  $\psi$  is the intertemporal elasticity of substitution. If  $\gamma = 1 \iff \theta = 1/\psi$ . We solve this by value function iteration.

Notice that  $w_{T+1} = 0 \Rightarrow c_T = w_T$ . Therefore,

$$v_T(w_T) = \left[ (1 - \beta) c_T^{\frac{1-\theta}{\gamma}} \right]^{\frac{\gamma}{1-\theta}} = \left[ (1 - \beta) (w_{T-1} - c_{T-1})^{\frac{1-\theta}{\gamma}} (R_T^p)^{\frac{1-\theta}{\gamma}} \right]^{\frac{\gamma}{1-\theta}}$$

Hence,

$$\begin{aligned} v_{T-1}(w_{T-1}) &= \max_{c_{T-1}, \alpha_{T-1}, w_T} \left\{ (1 - \beta) c_{T-1}^{\frac{1-\theta}{\gamma}} + \beta \left( \mathbf{E}_{T-1}(v_T(w_T)^{1-\theta})^{\frac{1}{\gamma}} \right) \right\}^{\frac{\gamma}{1-\theta}} \\ &= \max_{c_{T-1}} \left\{ (1 - \beta) c_{T-1}^{\frac{1-\theta}{\gamma}} + \beta (1 - \beta) (w_{T-1} - c_{T-1})^{\frac{1-\theta}{\gamma}} \max_{\alpha_{T-1}} \left( \mathbf{E}_{T-1}[(R_T^p)^{1-\theta}]^{\frac{1}{\gamma}} \right) \right\}^{\frac{\gamma}{1-\theta}} \end{aligned}$$

Notice that if  $V = \max_{x,y} (A(x) + B(x, y))^\rho$ . Then,

$$\frac{\partial V}{\partial y} = \rho(A + B)^{\rho-1} \frac{\partial B}{\partial y} = 0 \iff \frac{\partial B}{\partial y} = 0.$$

Assuming without loss of generality  $\rho(A + B)^{\rho-1} \neq 0$ .

Given the optimal share  $\alpha$  we solve for  $c_{T-1}$  using the first order condition

$$\left(\frac{1-\theta}{\gamma}\right)(1-\beta)c_{T-1}^{\frac{1-\theta}{\gamma}-1} - \left(\frac{1-\theta}{\gamma}\right)\beta(1-\beta)(w_{T-1} - c_{T-1})^{\frac{1-\theta}{\gamma}-1}A_{T-1} = 0.$$

$$c_{T-1}^{\frac{1-\theta}{\gamma}-1} = \beta A_{T-1}(w_{T-1} - c_{T-1})^{\frac{1-\theta}{\gamma}-1},$$

$$c_{T-1} = (\beta A_{T-1})^{\frac{1}{\frac{1-\theta}{\gamma}-1}}(w_{T-1} - c_{T-1}).$$

$$c_{T-1} = \frac{1}{1 + B_{T-1}}w_{T-1} = m_{T-1}w_{T-1},$$

where  $B_{T-1}^{-1} := (\beta A_{T-1})^{\frac{1}{\frac{1-\theta}{\gamma}-1}}$ ,  $A_{T-1} := \mathbf{E}_{T-1}[(R_T^p(\alpha))^{1-\theta}]^{\frac{1}{\gamma}}$ .

Going back to the value function

$$\begin{aligned}
v_{T-1}(w_{T-1}) &= \left\{ (1 - \beta) \left[ \frac{w_{T-1}}{1 + B_{T-1}} \right]^{\frac{1-\theta}{\gamma}} + \beta(1 - \beta) \left( w_{T-1} - \frac{w_{T-1}}{1 + B_{T-1}} \right)^{\frac{1-\theta}{\gamma}} A_{T-1} \right\}^{\frac{\gamma}{1-\theta}} \\
&= \left\{ (1 - \beta) \left( \frac{1}{1 + B_{T-1}} \right)^{\frac{1-\theta}{\gamma}} + \beta(1 - \beta) \left( \frac{B_{T-1}}{1 + B_{T-1}} \right)^{\frac{1-\theta}{\gamma}} A_{T-1} \right\}^{\frac{\gamma}{1-\theta}} w_{T-1} \\
&= \left\{ (1 - \beta)(1 + B_{T-1}) \left( \frac{1}{1 + B_{T-1}} \right)^{\frac{1-\theta}{\gamma}} \right\}^{\frac{\gamma}{1-\theta}} w_{T-1} \\
&= (1 - \beta)^{\frac{\gamma}{1-\theta}} m_{T-1}^{\left( \frac{1-\theta}{\gamma} - 1 \right) \left( \frac{\gamma}{1-\theta} \right)} w_{T-1} \\
&= (1 - \beta)^{\frac{\gamma}{1-\theta}} m_{T-1}^{\omega} w_{T-1},
\end{aligned}$$

where  $\beta A_{T-1} = B_{T-1}^{-\left( \frac{1-\theta}{\gamma} - 1 \right)}$ ,  $m = 1/(1 + B)$ ,  $\eta = 1/\psi$ ,  $\omega = -\eta/(1 - \eta)$ .

Proceeding backwards yields

$$v_t = (1 - \beta)^{\frac{\gamma}{1-\theta}} m_t^\omega w_t,$$

$$c_t = m_t w_t,$$

$$m_t = (1 + B_t)^{-1}, \quad \forall t \in [0, T - 1], m_T = 1,$$

$$B_t = [\beta(1 - \beta)(\mathbf{E}_t(R_{t+1}^p(\alpha)^{1-\theta} m_{t+1}^{\omega(1-\theta)})^{\frac{1}{\gamma}}]^{-\gamma/(1-\theta-\gamma)},$$

$$\alpha = \frac{\ln(1 + \mu) - \ln R^f + \sigma^2/2}{\theta\sigma^2}.$$

Assuming *iid* returns  $\mathbf{E}_t(m_{t+1} R_{t+1}^p) = \mathbf{E}_t(m_{t+1}) \mathbf{E}_t(R_{t+1}^p)$

$$\max_{\alpha_t} \left\{ \left( \mathbf{E}_t((m_{t+1}^\omega R_{t+1}^p)^{1-\theta}) \right)^{\frac{1}{\gamma}} \right\} \Rightarrow \frac{\partial \mathbf{E}_t(R_{t+1}^p(\alpha_t)^{1-\theta})}{\partial \alpha_t} = 0.$$

Notice that we can solve the portfolio choice using the standard approximation.