

Planner's problem:

$$\max_{\{c_{it}(s^+)\}} \sum_i \sum_t \sum_{s^t} \alpha_i \beta^+ \pi(s^+) u(c_{it}(s^+))$$

$$\text{s.t. } \sum_i c_{it}(s^+) \leq \sum_i y_{it}(s^+) \quad \forall t, s^+$$

F. O. C.

$$\frac{\alpha_i \beta^+ \pi(s^+) u_c^*(c_{it}(s^+))}{\alpha_j \beta^+ \pi(s^+) u_c^*(c_{jt}(s^+))} = \frac{\lambda_t(s^+)}{\lambda_t(s^+)}$$

~~OK~~

$$\Rightarrow \frac{u_c(c_{it}(s^+))}{u_c(c_{jt}(s^+))} = \frac{\alpha_j}{\alpha_i}$$

CRA:

$$\frac{u_c(c_{it}(s^+))}{u_c(c_{jt}(s^+))} = \frac{c_{it}(s^+)^{-\frac{1}{6}}}{c_{jt}(s^+)^{-\frac{1}{6}}}$$

Full insurance implies

$$c_{it}(s^+) = \left(\frac{\alpha_j}{\alpha_i}\right)^{-\frac{1}{6}} c_{jt}(s^+)$$

$$\Rightarrow \sum_i c_{it}(s^+) = \sum_i \left(\frac{\alpha_j}{\alpha_i}\right)^{-\frac{1}{6}} c_{jt}(s^+)$$

$$<= > G_b(s^+) = \alpha_j^{-\frac{1}{6}} c_{jt}(s^+) \cdot \sum_i \left(\frac{1}{\alpha_i}\right)^{-\frac{1}{6}}$$

$$<= > c_{jt}(s^+) = \underbrace{\frac{\alpha_j^{\frac{1}{6}}}{\sum_i \alpha_i^{\frac{1}{6}}} G(s^+)}_{\Theta_j}$$

(1)

- Define

$$RHS(a_1; \varepsilon) = \frac{1}{2} f(a_1; \varepsilon) + \frac{1}{2} f(a_1; -\varepsilon)$$

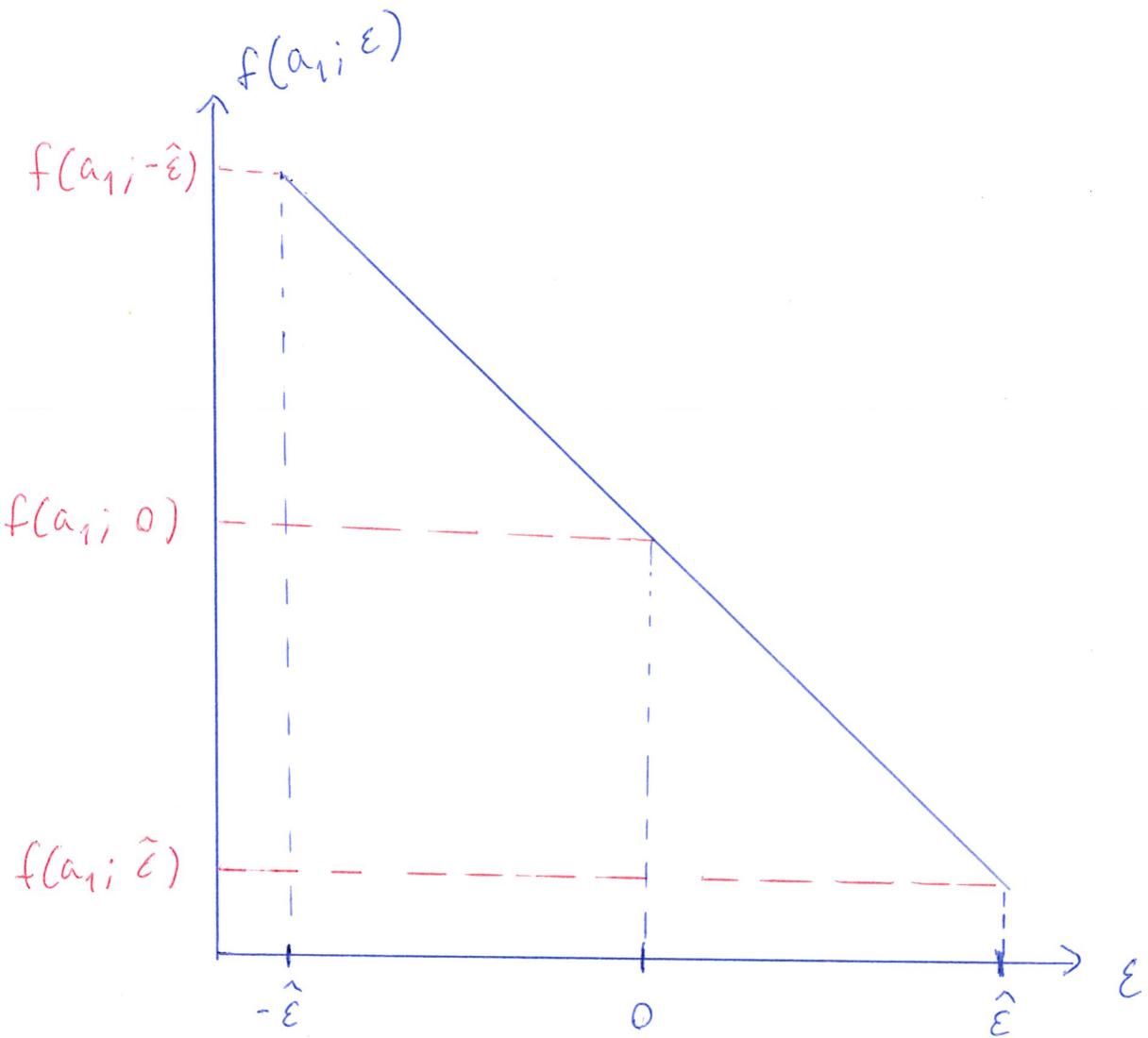
where

$$f(a_1; \varepsilon) = u^T (Ra_1 + \bar{y}_1 + \varepsilon)$$

- suppose $a_1(\varepsilon)$ solves

$$u^T (\gamma_0 - a_1) = RHS(a_1, \varepsilon) \quad (EE)$$

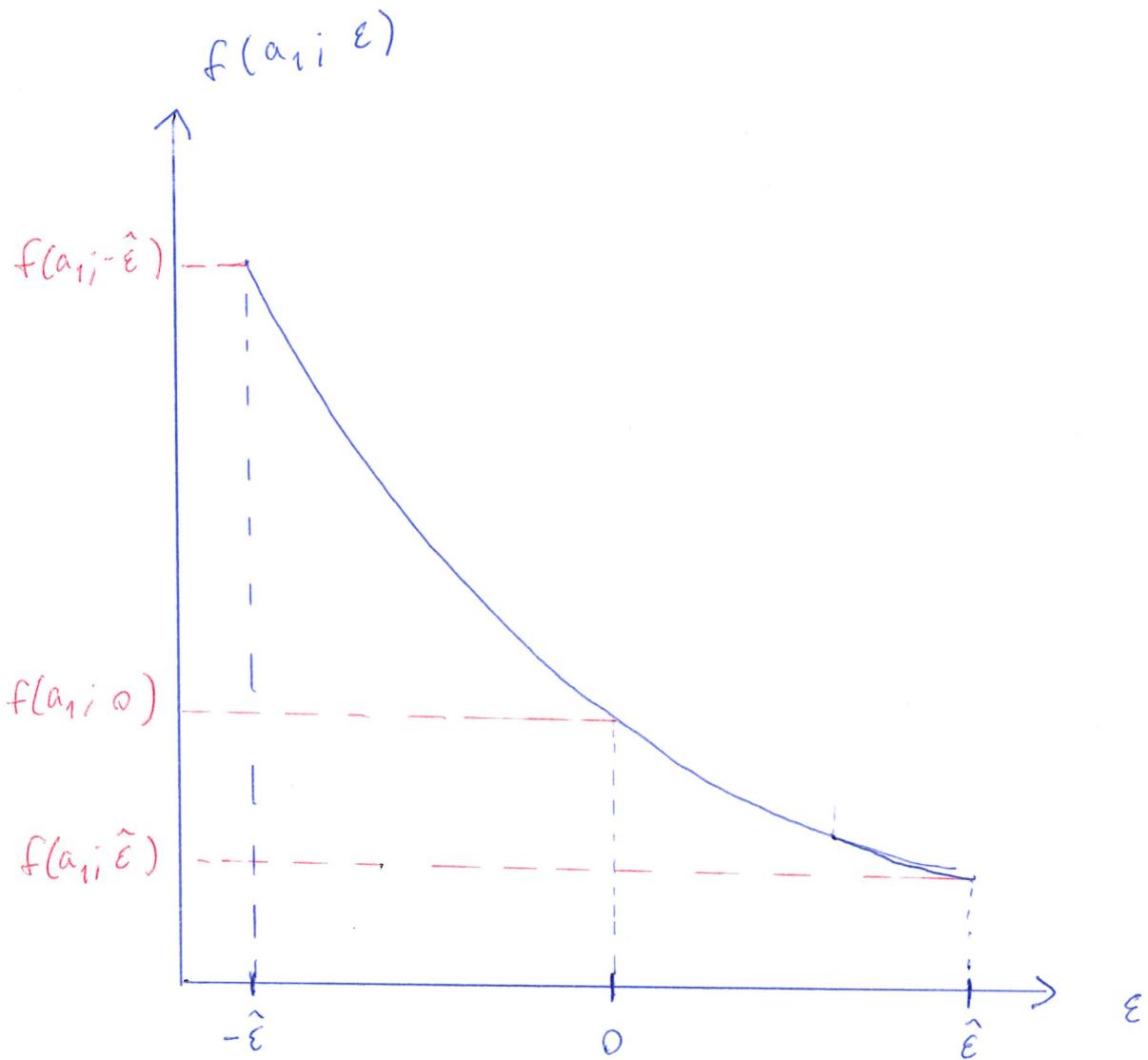
(2)

Case 1: $u^{m_1} = 0$ 

$$\begin{aligned}
 RHS(a_1; 0) &= f(a_1; 0) \\
 &= \frac{1}{2} f(a_1; -\hat{\epsilon}) + \frac{1}{2} f(a_1; \hat{\epsilon}) \\
 &\stackrel{?}{=} RHS(a_1; \hat{\epsilon})
 \end{aligned}$$

$$\Rightarrow a_1(\hat{\epsilon}) = a_1(0)$$

(3)

Case 2: $u''' > 0$ 

$$\text{RHS}(a_1, 0) = f(a_1; 0)$$

$$< \frac{1}{2} f(a_1; \hat{\epsilon}) + \frac{1}{2} f(a_1; -\hat{\epsilon})$$

$$= \text{RHS}(a_1, \hat{\epsilon})$$

$$\Rightarrow a_1(\hat{\epsilon}) > a_1(0)$$

Case 3: $u''' < 0$

$$\Rightarrow a_1(\hat{\epsilon}) < a_1(0)$$

At the optimum:

$$c = c(x)$$

$$\Rightarrow V_x(x) = u(c(x)) + \beta E V(R(x - c(x)) + y')$$

$$\Rightarrow V_x(x) = u_c(c) c_x(x) + \underbrace{R \beta E V_x(x') [R(1 - c_x(x))]}_{u_c(c) - R\mu}$$

$$\Rightarrow V_x(x) = u_c(c) c_x(x) + u_c(c) \cancel{+ R\mu} (1 - c_x(x)) - R\mu (1 - c_x(x))$$

$$= u_c(c) - R\mu (1 - c_x(x))$$

If c binding: $c_x(x) = 1$

If not: $\mu = 0$

$$\Rightarrow V_x(x) = u_c(c)$$