

A. $Y = K^\alpha H^\beta (AL)^{1-\alpha-\beta}$, $\tilde{y} = \tilde{k}^\alpha \tilde{h}^\beta$

이윤극대화의 일계조건 : $r_K = \alpha \frac{Y}{K} = \alpha \frac{\tilde{y}}{\tilde{k}}$, $r_H = \beta \frac{Y}{H} = \beta \frac{\tilde{y}}{\tilde{h}}$

Arbitrage(차익거래)가 없음 : $r_K = r_H \Rightarrow \tilde{h} = \frac{\beta}{\alpha} \tilde{k} \Rightarrow \tilde{y} = \lambda \tilde{k}^{\alpha+\beta}$, $\lambda = \left(\frac{\beta}{\alpha}\right)^\beta$

자본축적 방정식 : $\frac{\dot{\tilde{k}}}{\tilde{k}} = s_k \lambda \tilde{k}^{\alpha+\beta-1} - (n+g+d)$

B. 균형상태 근처에서 근사적으로 다음의 식이 성립함(테일러 전개)

$(f(x) \cong f(x^*) + f'(x)|_{x=x^*} (x-x^*))$, $\frac{d \log \tilde{k}}{dt} \equiv f(\log \tilde{k}) \equiv f(x) \Rightarrow$

$$\frac{\dot{\tilde{k}}}{\tilde{k}} = \frac{d \log \tilde{k}}{dt} \cong \frac{d \log \tilde{k}}{dt} \Big|_{\log \tilde{k} = \log \tilde{k}^*} + \frac{d \left(\frac{d \log \tilde{k}}{dt} \right)}{d \log \tilde{k}} \Big|_{\log \tilde{k} = \log \tilde{k}^*} (\log \tilde{k} - \log \tilde{k}^*)$$

$\left(\frac{d \log \tilde{k}}{dt} \Big|_{\log \tilde{k} = \log \tilde{k}^*} = 0 \Rightarrow \right)$

$$= \frac{d}{d \log \tilde{k}} \left[s_k \lambda \tilde{k}^{\alpha+\beta-1} - (n+g+d) \right] \Big|_{\log \tilde{k} = \log \tilde{k}^*} (\log \tilde{k} - \log \tilde{k}^*)$$

$(x = \exp(\log x) \Rightarrow)$

$$= \frac{d}{d \log \tilde{k}} \left[s_k \lambda \exp(\log \tilde{k}^{\alpha+\beta-1}) - (n+g+d) \right] \Big|_{\log \tilde{k} = \log \tilde{k}^*} (\log \tilde{k} - \log \tilde{k}^*)$$

$\left(\frac{d \exp(f(x))}{dx} = f'(x) \exp(f(x)) \Rightarrow \right)$

$$= (\alpha + \beta - 1) s_k \lambda \tilde{k}^{\alpha+\beta-1} \Big|_{\log \tilde{k} = \log \tilde{k}^*} (\log \tilde{k} - \log \tilde{k}^*)$$

$\left(s_k \lambda \tilde{k}^{\alpha+\beta-1} \Big|_{\log \tilde{k} = \log \tilde{k}^*} = n+g+d \text{ (균형상태이므로)} \Rightarrow \right)$

$$= (1 - \alpha - \beta)(n+g+d)(\log \tilde{k}^* - \log \tilde{k}) \equiv \gamma(\log \tilde{k}^* - \log \tilde{k})$$

$$C. \quad \tilde{y} = \lambda \tilde{k}^{\alpha+\beta} \Rightarrow \frac{\dot{\tilde{y}}}{\tilde{y}} = (\alpha + \beta) \frac{\dot{\tilde{k}}}{\tilde{k}},$$

$$\log \tilde{y}^* = \log \lambda + (\alpha + \beta) \log \tilde{k}^*, \quad \log \tilde{y} = \log \lambda + (\alpha + \beta) \log \tilde{k}$$

$$\Rightarrow \log \frac{\tilde{y}^*}{\tilde{y}} = (\alpha + \beta) \log \frac{\tilde{k}^*}{\tilde{k}}$$

$$\Rightarrow \frac{\dot{\tilde{y}}}{\tilde{y}} = (\alpha + \beta) \frac{\dot{\tilde{k}}}{\tilde{k}} = (\alpha + \beta) \gamma (\log \tilde{k}^* - \log \tilde{k})$$

$$= \gamma (\log \tilde{y}^* - \log \tilde{y}), \quad \gamma = (1 - \alpha - \beta)(n + g + d)$$

$$\Rightarrow \frac{d \log \tilde{y}}{dt} = \gamma (\log \tilde{y}^* - \log \tilde{y}) \quad (1 \text{계선형미분방적식})$$

$$\begin{aligned} \Rightarrow \log \tilde{y}(t) &= \log \tilde{y}^* + (\log \tilde{y}(0) - \log \tilde{y}^*) e^{-\gamma t} \\ &= (1 - e^{-\gamma t}) \log \tilde{y}^* + e^{-\gamma t} \log \tilde{y}(0) \end{aligned}$$