

PS 2



Exercise 1

$$\underset{t \geq 0}{\text{Max}} \sum_{t=0}^{\infty} \beta^t \log c_t$$

s.t. $\sum_{t=0}^{\infty} c_t \leq k_0$ we already know that the whole cake is going to be consumed

$$c_t \geq 0, k_0 \geq 0$$

k_0 given

$$L = \sum_{t=0}^{\infty} \beta^t \ln(c_t) - \lambda \left(\sum_{t=0}^{\infty} c_t - k_0 \right)$$

$$\text{FOC: } \frac{\partial L}{\partial c_t} = 0 \quad \frac{\beta^t}{c_t} = \lambda$$

$$\cdot \frac{\partial L}{\partial \lambda} = 0 \quad \sum_{t=0}^{\infty} c_t - k_0 = 0$$

Euler Equation:

$$\begin{cases} \lambda = \frac{\beta^t}{c_t} \\ \lambda = \frac{\beta^{t+1}}{c_{t+1}} \end{cases} \quad \frac{c_{t+1}}{c_t} = \beta$$

2) The complementary slackness implies

$$\sum_{t=0}^{\infty} c_t = k_0 \quad \text{NB: } c_t = \beta^t c_0$$

$$\sum_{t=0}^{\infty} \beta^t c_0 = k_0 \Rightarrow \frac{c_0}{1-\beta} = k_0 \quad c_0 = (1-\beta)k_0$$

$$\text{and } c_t = \beta^t c_0 = \beta^t (1-\beta) k_0$$

* Short. focus on the multiplier λ

In an optimization problem with objective function $f(x_1, x_2, \dots, x_n)$ subject to a constraint $g(x_1, x_2, \dots, x_n)$, the Lagrange multiplier, tells us the rate of change of the objective function with respect to a change in the constraint.

In other words, λ measures how much the value of the optimal solution would change if the constraint were relaxed (high λ) or tightened slightly.

In our case:

The total amount of the cake is K_0 . λ represents the marginal utility of an additional unit of K_0 .

From FOC: $\frac{\beta^+}{C^+} = \lambda = \frac{\beta^+}{K_0(l-\beta)\beta^+} = \frac{l}{K_0(l-\beta)} > 0$

If K_0 is small, λ is very large meaning that the agent would gain a lot with an additional unit of K_0 .

As K_0 increases (relaxing the BC) λ will decrease which is coherent with decreasing marginal utility
 \Rightarrow THE AGENT GAINS MARGINALLY LESS FOR EACH UNIT INCREASE IN K_0

3) How much cake left at time T?

$$\sum_{t=0}^T c_t = \sum_{t=0}^T \beta^t k_0 = \sum_{t=0}^T \beta^t (1-\beta) k_0 = \cancel{(1-\beta)} k_0 \frac{(1-\beta^{T+1})}{1-\beta}$$
$$= k_0 (1-\beta^{T+1})$$

$$\text{Cake left} = k_0 - k_0 (1-\beta^{T+1}) = \underbrace{k_0 \beta^{T+1}}$$

5) Transversality Condition

$$\lim_{T \rightarrow \infty} \beta^T u'(c_T) k_T = \beta \underbrace{\frac{1}{C_T}}_{\beta^T} k_T = \lim_{T \rightarrow \infty} \beta^T \frac{1}{\beta^T k_0} k_0 \beta^{T+1} = 0 \quad \checkmark$$

6) If the resource constraint is not satisfied
in this problem we would have that

$\lambda \left(\sum_{t=0}^T c_t - k_0 \right) \neq 0$ violating the
complementary slackness condition since λ
is always positive $\lambda = \frac{\beta}{C_T} > 0$

In this case we should adopt different preferences
for the individual to make the problem
respect the condition to find an optimum

1) Cake-Eating Problem: A more general approach

$$\sum_{t=0}^{T-1} \beta^t \ln(c_t)$$

s.t. $k_{t+1} = k_t - c_t \Rightarrow$ you can rewrite infinitely many budget constraint

$$k_0 - c_0 - k_1 \geq 0 \quad \text{also } k_t \geq 0, c_t \geq 0$$

$$k_1 - c_1 - k_2 \geq 0$$

\vdots

$$k_t - c_t - k_{t+1} \geq 0$$

\vdots

$$k_T - c_T - k_{T+1} \geq 0$$

|| End of the world, we need a transversality condition

$$\Rightarrow \text{This is the same as } c + c_1 + c_2 + \dots + c_T = k_0$$

$$y = \sum_{t=0}^{T-1} \beta^t \left[\ln(c_t) + \lambda_t (k_t - c_t - k_{t+1}) + \mu_{T+1}^k \right]$$

N.B.: Here we have infinitely many budget constraint
and so λ depends on time.

FOCs

$$1) \frac{\frac{\partial y}{\partial c_t}}{\frac{\partial c_t}{\partial c_t}} = \frac{\beta^t}{c_t} = \beta^t \lambda_t \Rightarrow \frac{1}{c_t} = \lambda_t$$

THERE ARE TWO CONSECUTIVE BUDGET CONSTRAINT WHERE k_{T+1} IS PRESENT

$$2) \frac{\frac{\partial y}{\partial k_{t+1}}}{\frac{\partial k_{t+1}}{\partial k_{t+1}}} = -\beta^t \lambda_t + \beta^{t+1} \lambda_{t+1} \Rightarrow \lambda_t = \beta \lambda_{t+1}$$

Substitute 2) in 1)

$$\frac{1}{c_t} = \beta \frac{1}{c_{t+1}} \Rightarrow \frac{c_{t+1}}{c_t} = \beta \Rightarrow \begin{array}{l} \text{CONSUMPTION GROWTH IS} \\ \text{CONSTANT} \end{array}$$

$$3) \frac{\partial y}{\partial k_{T+1}} = 0 \quad \mu - \beta^T \lambda_T = 0$$

$$4) \sum_{t=0}^{T-1} \lambda_t (k_t - c_t - k_{t+1}) = 0 \Rightarrow \begin{array}{l} \text{COMPLEMENTARY SLACKNESS} \\ \text{CONDITIONS} \end{array}$$

$$\cdot \mu k_{T+1} = 0$$

- $M - \beta^T \lambda_T = 0$
 - $M k_{T+1} = 0$
- Combining the two gives
- $$M k_{T+1} - \beta^T \lambda_T k_{T+1} = 0$$
- 0
- $\boxed{\beta^T \lambda_T k_{T+1} = 0}$ \Rightarrow This implies $k_{T+1} = 0$
- $\boxed{\text{So}}$
- In infinite horizon, a natural extension is the so called transversality condition
- $$\lim_{t \rightarrow \infty} \beta^t \lambda_t k_{t+1} = 0 = \lim_{t \rightarrow \infty} \beta^t u'(c(t)) k_{t+1} = 0$$

Optimal Consumption profile (finite HORIZON T)

$$\frac{k_{t+1}}{c_t} = \beta$$

$$\Rightarrow \text{Start at } T+1, k_{T+1} = 0 \text{ by slackness.}$$

$$k_{T+1} = k_T - c_T = 0 \Rightarrow k_T = c_T$$

$$\frac{c_T}{c_{T-1}} = \beta$$

$$c_{T-1} = \frac{1}{\beta} \frac{k_T}{k_{T-1}} \quad k_T = k_{T-1} - c_{T-1}$$

$$c_{T-1} = \frac{1}{\beta} (k_{T-1} - c_{T-1}) \Rightarrow c_{T-1} = \frac{1}{1+\beta} k_{T-1}$$

$$c_{T-2} = \frac{1}{\beta} c_{T-1} \Rightarrow c_{T-2} = \frac{1}{\beta} \frac{1}{1+\beta} k_{T-1};$$

$$k_{T-1} = k_{T-2} - c_{T-2}$$

$$C_{T-2} = \frac{1}{\beta + \beta^2} (K_{T-2} - C_{T-2}) \Rightarrow C_{T-2} = \frac{1}{1 + \beta + \beta^2} K_{T-2}$$

⋮
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$$C_0 = \frac{1}{1 + \beta + \dots + \beta^{T+1}} K_0 = \frac{1 - \beta}{1 - \beta^{T+1}} K_0$$

another way

$$C_T = \beta C_{T-1} = \beta \beta^{T-1} C_0 = \beta^T C_0$$

Now take the total sum

$$\underbrace{\sum_{t=0}^{\infty} C_T}_{1} = \sum_{t=0}^{\infty} \beta^T C_0 = \frac{1 - \beta^{T+1}}{1 - \beta} C_0$$

$$\left. \begin{aligned} K_0 & \\ K_0 = \frac{1 - \beta^{T+1}}{1 - \beta} C_0 & \end{aligned} \right\} \Rightarrow \text{Path } f_i, C_0$$

Now let's see it w/ INFINITE HORIZON ($t = \infty$)

$$Q = \sum_{t=0}^{\infty} \beta^t \ln(c_t)$$

$$K_{t+1} = K_t - c_t$$

$$K_{t+0} = k_0$$

$$Q = \sum_{t=0}^{\infty} \beta^t \ln(c_t) - \lambda_t (K_{t+1} - K_t + c_t)$$

$$\text{FOC } \frac{\partial Q}{\partial c_t} = \beta \frac{\lambda_t}{c_t} = \lambda_t + \beta$$

$$\frac{\partial Q}{\partial K_{t+1}} = -\beta \lambda_t + \beta \lambda_{t+1} = 0$$

$$\left. \begin{array}{l} \\ \end{array} \right\} \Rightarrow \frac{c_{t+1}}{c_t} = \beta$$

$$+ \lim_{t \rightarrow \infty} \beta^t \lambda_t \underline{K_{t+1}} = 0 \Rightarrow \text{TRANSVERSALITY}$$

Now I put you two ways of solving the problem, so you see where I need the TVC.

$$k_1 = k_0 - c_1 = k_0 - c_0 - \beta c_0 = k_0 - (1-\beta)c_0$$

$$k_2 = k_1 - c_2 = k_1 - c_1 - \beta c_1 = k_0 - (1-\beta)c_0 - \beta^2 c_0$$

$$k_{T+1} = k_0 - \frac{(1-\beta)}{1-\beta^{T+1}} c_0$$

$$c_0 = \frac{1-\beta}{1-\beta^{T+1}} k_0 - \frac{1-\beta}{(1-\beta)^{T+1}} k_{T+1}$$

We can't just get rid of k_{T+1} , we need an additional condition
 $\Rightarrow \text{TRANSVERSALITY!}$

$$\lim_{T \rightarrow \infty} \beta^T M'(c_T) K_{T+1} = 0 \Rightarrow K_{T+1} \text{ is the cake left and we computed it before}$$

$$= \lim_{T \rightarrow \infty} \beta^T \frac{1}{\beta^T G} \cdot \underline{K_0 \beta^{T+1}} = 0 \quad K_{T+1} = K_0 \beta^{T+1}$$

Here I proved you that the TVC holds. this is a necessary condition, if
should hold **EX-ANTE!**

$$c_0 = \underbrace{\frac{(1-\beta)}{1-\beta} K_0}_{\rightarrow (1-\beta) \rightarrow 1} + \underbrace{\frac{(1-\beta)}{1-\beta} K_{T+1}}_{\rightarrow 0 \text{ by TVC}}$$

$$c_0 = (1-\beta) K_0$$

What about c_T ? Use Euler!

$$\frac{c_T}{c_0} = \beta^t = \beta^t c_0 = \beta^t (1-\beta) K_0$$

Simpler version that you can use (bc. TVC holds)

$$G = \sum_{t=0}^{\infty} c_t \Rightarrow \text{Now I take the infinite sum}$$

$$\sum_{t=0}^{\infty} c_t = \sum_{t=0}^{\infty} \beta^t c_0 \Rightarrow c_0 = (1-\beta) K_0$$

$$\bullet \frac{c_t}{c_{t-1}} = \beta$$

This two equations describe the whole model

2) Exponential Utility

$$U(C) = -\frac{1}{\theta} e^{-\theta C} \Rightarrow \text{Utility of households over infinite horizon}$$

a) Relate θ to the concavity of utility functions: check 2nd derivative to be < 0

$$\cdot \frac{\partial U(C)}{\partial C} = \frac{\partial}{\partial C} \left(-\frac{1}{\theta} e^{-\theta C} \right) = e^{-\theta C} > 0 \quad \checkmark$$

$$\cdot \frac{\partial^2 U(C)}{\partial C^2} = \frac{\partial}{\partial C} \left(e^{-\theta C} \right) = -\theta e^{-\theta C} < 0 \quad \text{for } \theta > 0$$

Parameter θ governs the degree of concavity; $\uparrow \theta \uparrow$ concavity

b) Intertemporal Elasticity of substitution

$$IES = -\frac{U'(C)}{C U''(C)} = \frac{e^{-\theta C}}{C - \theta e^{-\theta C}} = \frac{1}{\theta C} \Rightarrow \begin{array}{l} \text{higher } \theta \Rightarrow \downarrow IES \\ \text{higher } C \Rightarrow \downarrow IES \end{array}$$

It depends on both θ and C

2) Set up the problem:

The household maximizes:

$$\max \left\{ e^{-pt} U(C(t)) \right\}$$

$$\text{s.t. } \dot{K}(t) = w(t) + r(t)K(t) - c(t) \Rightarrow \text{in per capita terms, assuming population constant } L(t) = L$$

and No-Panti condition:

$$\lim_{t \rightarrow \infty} \int_t^{\infty} K_s \times \exp \left[- \int_0^s (r(s)) ds \right] \geq 0 \Rightarrow \text{Always positive or zero capital in the limit}$$

REMEMBER: 1) The transversality condition (TVC) is about optimal behavior

2) The No-Panti condition is a condition to make sure that the flow budget constraints are consistent w/ a lifetime budget constraint
(i.e. aggregated across periods)

N.B.

Set up the Hamiltonian:

$$\gamma(-e^{-kt} u(c(t))) + \bar{m}(t) \left[w(t) + \gamma(t) K(t) - c(t) \right]$$

i) $H_c(c, k, \bar{m}) = 0 \Rightarrow e^{-kt} \bar{m}'(c(t)) = \bar{m}(t)$

ii) $H_k(c, k, \bar{m}) = -\dot{m}(t) \Rightarrow \bar{m}(t) \cdot (1 + \gamma(t)) = -\dot{m}(t)$

$$H_{\bar{m}}(c, k, \bar{m}) = k(t)$$

$\lim_{t \rightarrow \infty} e^{-kt} \bar{m}(t) k(t) = 0$ + No positive scheme condition
and initial condition $k(0) = k_0$

To compute $\dot{m}(t)$ take the true derivative of the first FOC
 $\dot{m}(t) = \frac{d}{dt} \bar{m}(t) = \frac{d}{dt} e^{-kt} \bar{m}'(c(t))$

TVC Now put 1) + 2) together

$$e^{-kt} (1 + \gamma(t)) \bar{m}'(c(t)) = e^{-kt} \bar{m}'(c(t)) - e^{-kt} \bar{m}''(c(t)) \dot{c}(t)$$

$$\frac{\bar{m}'(c(t))}{\bar{m}''(c(t))} [\gamma(t) - \ell] = -\dot{c}(t) \Rightarrow \frac{\dot{c}(t)}{c} = \sigma (1 + \gamma(t)) - \ell$$

$$w/\sigma = \left(-\frac{\bar{m}'(c(t))}{\bar{m}''(c(t))} \right)$$

This is the IES

Consumption grows as long as $1 + \gamma(t) > \ell$ and how much it grows depends on the IES (σ). To see this function Just plug-in the utility

3) The fun: $\max_{k, L} F(k, L) - w(t)L - R(t)K$

$$w(t) = F_L(k, L) = F_L(k, 1)$$

$$R(t) = F_K(k, L) = F_K(k, 1)$$

$$\gamma(t) = R(t) - \delta \Rightarrow \text{Total return of capital}$$

$$w(t) = F_L(k, L) = f(k) - f'(k)$$

From the capital accumulation

$$K(t) = \overline{[w(t) + \gamma(t)] K(t)} - c(t)$$

We know that
 $\gamma(t) = F_K(k, L) - \delta$

$$\dot{k}(t) = f(k) - \delta k(t) - c(t) \quad \bar{c}(t) = -\frac{u'(c)}{u''(c)c}$$

$$\dot{c}(t) = (\gamma c(t) - \ell) \bar{c}(t) = (f'(k) - \ell - \delta) \bar{c}(t)$$

$$\frac{\dot{c}(t)}{c} = \left[\gamma c(t) - \ell \right] \frac{e^{-\theta c}}{c^\theta e^{-\theta c}} = \frac{\gamma c(t) - \ell}{c^\theta}$$

$$\frac{\dot{c}(t)}{c} = \frac{[f'(k) - \ell - \delta]}{c^\theta}$$

Steady State

$$k(t) = 0 \quad \dot{c}(t) = 0$$

$$f(k^*) - \delta k^* = c^* \quad || \text{ Net output is consumed}$$

$$\dot{c}^*(t) = 0$$

$$f'(k^*) = \ell + \delta + \gamma$$

Growth rate of capital is equal to the MP

S) During the transition towards the ss, the parameter θ governs the speed of consumption growth. A higher θ means lower $\dot{c}(t)$ meaning that the individual are more reluctant to adjust consumption in response to changes in the interest rate.

Remember $\text{ES} = \frac{1}{CRRA}$ (CRRA = coefficient of risk aversion). So the individual is more risk averse.

- If you want to add population growth the model becomes:

$$\max \mathcal{U} = \int_0^{\infty} \left(\frac{1}{\theta} \right) e^{-\theta c} e^{-(h-\rho)t} dt$$

s.t. $\dot{k} = w(t) + r k(t) - c(t) - n k(t)$ | per capita
term

The Hamiltonian

$$H = -\left(\frac{1}{\theta}\right) e^{-\theta c} e^{-(h-\rho)t} + \mu(t) (r k(t) + w(t) - c(t) - n k(t))$$

you'll find the same solution only the evolution of \dot{k} is also influenced by n .

$$\dot{k} = f(k) - c - (h+\delta)k$$

$$\dot{c} = \frac{1}{\theta c} (f'(k) - \delta - \rho)$$

~~Behavior of the model.~~

Exercise 3

From: $Y_{SJ} = A_{SJ} k_{SS}^{\alpha} L_{SS}^{1-\alpha}$

$$\text{Sector: } Y_S = \left(\sum_{S=1}^M Y_{SS}^{\frac{\alpha-1}{\alpha}} \right)^{\frac{\alpha}{\alpha-1}}$$

Demand for variety J in sector S :

$$Y_J = \left(\frac{P_J}{P} \right)^{\frac{1}{\alpha}} Y \quad P_J = P \left(\frac{Y_J}{Y} \right)^{\frac{1}{\alpha}} ; \text{ normalize } P = 1$$

We are considering just one sector

Firm's maximization problem

1) $\max_{K_S, L_S} p_S Y_S (1 - \gamma_{S,y}) - w L_S - r (1 + \gamma_{K,S}) K_S$ $\left. \begin{array}{l} p_S = \left(\frac{y_S}{Y} \right)^{\frac{1}{\alpha}} \\ Y_S = A_S K_S^{\alpha} L_S^{1-\alpha} \end{array} \right\} \text{Substitute inside the objective function}$

$$\max_{K_S, L_S} y_S^{\frac{1-\frac{1}{\alpha}}{\alpha}} (1 - \gamma_{S,y}) - w L_S - r (1 + \gamma_{K,S}) K_S = \left[A_S K_S^{\alpha} L_S^{1-\alpha} \right]^{\frac{1}{\alpha}} y^{\frac{1}{\alpha}} - w L_S - r (1 + \gamma_{K,S}) K_S$$

KOC $\frac{\partial \uparrow}{\partial K_S} = (1 - \gamma_{S,y}) \left(1 - \frac{1}{\alpha} \right) y_S^{\frac{1}{\alpha}} y^{\frac{1}{\alpha}} \alpha K_S^{\alpha-1} L_S^{1-\alpha} - r (1 + \gamma_{K,S}) = 0 \quad 1)$

$$\frac{\partial \uparrow}{\partial L_S} = (1 - \gamma_{S,y}) \left(1 - \frac{1}{\alpha} \right) y_S^{\frac{1}{\alpha}} y^{\frac{1}{\alpha}} (1 - \alpha) \alpha K_S^{\alpha} L_S^{\alpha-1} - w = 0 \quad 2)$$

3) $\frac{r (1 + \gamma_{K,S})}{\alpha} = (1 - \gamma_{S,y}) y_S^{\frac{1}{\alpha}} y^{\frac{1}{\alpha}} \alpha K_S^{\alpha-1} L_S^{1-\alpha}$

4) $\frac{w}{1-\alpha} = (1 - \gamma_{S,y}) \left(1 - \frac{1}{\alpha} \right) y_S^{\frac{1}{\alpha}} y^{\frac{1}{\alpha}} K_S^{\alpha-1} L_S^{\alpha-1} A_S$

remember: $y_S^{\frac{1}{\alpha}} y^{\frac{1}{\alpha}} = p_S$

$$\frac{r (1 + \gamma_{K,S})}{\alpha} = \frac{w}{1-\alpha} \frac{L_S}{K_S}$$

$$(1 + \gamma_{K,S}) = \frac{w \alpha}{r} \frac{L_S}{1-\alpha} \frac{K_S}{K_S}$$

Back to 1) $(1 - \gamma_{S,y}) \left(1 - \frac{1}{\alpha} \right) p_S \alpha \frac{A_S K_S^{\alpha-1} L_S^{1-\alpha}}{K_S} = \gamma_S \cdot \frac{w \alpha}{r} \frac{L_S}{1-\alpha} \frac{K_S}{K_S}$

$$(1 - \pi_{s,y}) \left(\frac{1-\alpha}{\alpha} \right) p_J y_J = \frac{1}{1-\alpha} w L_J$$

$$(1 - \pi_{s,y}) = \left(\frac{1}{1-\alpha} \right) \left(\frac{1}{1-\alpha} \right) \frac{w L_J}{p_J y_J}$$

3) Revenue Productivity: $p_J A_J$

• Remember from FOC

$$\begin{aligned} & \bullet (1 - \pi_{s,y}) \alpha \left(1 - \frac{1}{\alpha} \right) p_J y_J = MPK k_J \\ & \bullet (1 - \pi_{s,y}) (1 - \alpha) \left(1 - \frac{1}{\alpha} \right) p_J y_J = MPL L_J \end{aligned} \quad \left. \begin{aligned} \frac{\alpha}{1-\alpha} &= \frac{MPK}{MPL} \cdot \frac{k_J}{L_J} \\ \frac{k_J}{L_J} &= \frac{MPL}{MPK} \frac{\alpha}{1-\alpha} \end{aligned} \right\}$$

$$(1 - \pi_{s,y}) \alpha \left(1 - \frac{1}{\alpha} \right) p_J y_J + (1 - \pi_{s,y}) (1 - \alpha) \left(1 - \frac{1}{\alpha} \right) p_J y_J = MPK k_J + MPL L_J$$

$$7) (1 - \pi_{s,y}) \left(1 - \frac{1}{\alpha} \right) p_J y_J = MPK k_J + MPL L_J$$

$$\bullet \text{Remember } \frac{p_{s,y}}{p_J} : A_J \left(\frac{k_J}{L_J} \right)^{\frac{\alpha}{1-\alpha}} L_J = A_J \left(\frac{L_J}{k_J} \right)^{1-\alpha} k_J$$

$$y_J = A_J k_J^{\alpha} L_J^{1-\alpha} = A_J \left(\frac{k_J}{L_J} \right)^{\frac{\alpha}{1-\alpha}} L_J = A_J \left(\frac{L_J}{k_J} \right)^{1-\alpha} k_J$$

$$8) k_J = \frac{y_J}{A_J} \left(\frac{k_J}{L_J} \right)^{1-\alpha} = \frac{y_J}{A_J} \left(\frac{\alpha}{1-\alpha} \frac{MPL}{MPK} \right)^{-\alpha}; 9) L_J = \frac{y_J}{A_J} \left(\frac{\alpha}{1-\alpha} \frac{MPK}{MPL} \right)^{-\alpha}$$

$$\text{RHS of 7): } MPK \frac{y_J}{A_J} \left(\frac{\alpha}{1-\alpha} \frac{MPL}{MPK} \right)^{-\alpha} + MPL \frac{y_J}{A_J} \left(\frac{\alpha}{1-\alpha} \frac{MPK}{MPL} \right)^{-\alpha}$$

$$= \frac{y_J}{A_J} \left[MPK \left(\frac{\alpha}{1-\alpha} \frac{MPL}{MPK} \right)^{-\alpha} + MPL \left(\frac{\alpha}{1-\alpha} \frac{MPK}{MPL} \right)^{-\alpha} \right] =$$

$$= \frac{y_J}{A_J} \left[MPK \frac{\alpha}{1-\alpha} \frac{MPL}{MPK} \left(\frac{\alpha}{1-\alpha} \frac{MPL}{MPK} \right)^{-\alpha} + MPL \left(\frac{\alpha}{1-\alpha} \frac{MPK}{MPL} \right)^{-\alpha} \right]$$

$$= \frac{y_J}{A_J} \left[\left(\frac{\alpha}{1-\alpha} \frac{MPL}{MPK} \right)^{-\alpha} \left(\frac{\alpha MPL}{1-\alpha} + MPL \right) \right] = \frac{y_J}{A_J} \left[\left(\frac{\alpha}{1-\alpha} \frac{MPL}{MPK} \right)^{-\alpha} \frac{MPL}{1-\alpha} \right]$$

$$= \left(\frac{MP_L}{1-\alpha} \right)^{1-\alpha} \left(\frac{MP_K}{\alpha} \right)^\alpha \frac{y_J}{A_J} = MP_K K_J + MP_L L_J$$

Now plug the above expression into 7)

$$(1 - \gamma_{S,y}) \left(\frac{\sigma - 1}{\sigma} \right) P_J Y_J = \left(\frac{MP_L}{1-\alpha} \right)^{1-\alpha} \left(\frac{MP_K}{\alpha} \right)^\alpha \frac{y_J}{A_J}$$

$$MP_L = w$$

$$MP_K = (1 + \gamma_{S,K}) z$$

$$P_J = \frac{\sigma}{\sigma-1} \left(\frac{w}{1-\alpha} \right)^{1-\alpha} \left(\frac{z}{\alpha} \right)^\alpha \frac{(1 + \gamma_{S,K})^\alpha}{1 - \gamma_{S,y}} \cdot \frac{1}{A_J}$$

$$q) \gamma_{S,K} = 0 ; \gamma_{S,y} = 0$$

$$\text{W/out distortions } P_J A_J = \frac{\sigma}{\sigma-1} \left(\frac{w}{1-\alpha} \right)^{1-\alpha} \left(\frac{z}{\alpha} \right)^\alpha \frac{1}{1} \quad \checkmark$$

$$\Rightarrow \frac{(1 + \gamma_{S,K})^\alpha}{1 - \gamma_{S,y}} \Rightarrow \text{Factor of distortion, thus the output decrease w/ distortion}$$

by which