

Exam #2 Solution Outline

1. a. True. b. True. c. True.

d. False. For each $i = 1, 2, \dots, n$: $\frac{\partial F}{\partial x_i}(x^*) - \lambda^* \frac{\partial g}{\partial x_i}(x^*) \leq 0$.

e. True.

2. a. $L(x_1, x_2; \lambda) = x_2 e^{x_1} + \lambda(M - p_1 x_1 - p_2 x_2)$

$$\text{I. } \frac{\partial L}{\partial x_1} = x_2 e^{x_1} - p_1 \lambda \leq 0, \quad x_1 \geq 0, \quad (x_2 e^{x_1} - p_1 \lambda) \cdot x_1 = 0$$

$$\text{II. } \frac{\partial L}{\partial x_2} = e^{x_1} - p_2 \lambda \leq 0, \quad x_2 \geq 0, \quad (e^{x_1} - p_2 \lambda) \cdot x_2 = 0$$

$$\text{III. } \frac{\partial L}{\partial \lambda} = M - p_1 x_1 - p_2 x_2 \geq 0, \quad \lambda \geq 0, \quad (M - p_1 x_1 - p_2 x_2) \cdot \lambda = 0$$

b. Marginal condition in II. requires $\lambda^* > 0$. Combined with complementary slackness condition in III., this gives:

$$p_1 x_1^* + p_2 x_2^* = M \quad (\text{budget constraint binding})$$

With budget constraint binding, cannot have $x_1^* = x_2^* = 0$.

Possible to have $x_1^* > 0, x_2^* = 0$?

Then complementary slackness condition in I. implies $\lambda^* = 0$, but this violates marginal condition in II.

Possible to have $x_1^* = 0, x_2^* > 0$?

$$\text{From I.: } x_2^* - p_1 \lambda^* \leq 0$$

$$\text{From II: } 1 - p_2 \lambda^* = 0 \Rightarrow \lambda^* = \frac{1}{p_2}$$

$$\text{So } x_2^* \leq \frac{p_1}{p_2}. \text{ Budget constraint } \Rightarrow x_2^* = \frac{M}{p_2}, \text{ requiring } M \leq p_1.$$

Possible to have $x_1^* > 0, x_2^* > 0$?

$$\text{I. } \Rightarrow x_2^* e^{x_1^*} - p_1 \lambda^* = 0$$

$$\text{II. } \Rightarrow e^{x_1^*} - p_2 \lambda^* = 0$$

$$\text{Taking ratio: } x_2^* = \frac{p_1}{p_2}. \text{ From budget constraint: } x_1^* = \frac{M - p_1}{p_1} > 0, \text{ requiring } p_1 < M. \text{ (Note also: } \lambda^* > 0.)$$

Summary: Two types of solutions are possible:

$$\text{When } M \leq p_1: x_1^* = 0, x_2^* = \frac{M}{p_2} > 0, p_1 x_1^* + p_2 x_2^* = M.$$

$$\text{When } M > p_1: x_1^* = \frac{M - p_1}{p_1} > 0, x_2^* = \frac{p_1}{p_2} > 0, p_1 x_1^* + p_2 x_2^* = M.$$

3. Proof: Form a sequence of distinct points in B : $\{x_n\}_{n=1}^{\infty}$. (For each n , can choose $x_n \neq x_m$ for all $m = 1, 2, \dots, n-1$, because B is infinite.)

$\{x_n\}_{n=1}^{\infty}$ is a sequence in a compact subset of \mathfrak{R}^n . By Bolzano-Weierstrass, $\{x_n\}_{n=1}^{\infty}$ has a convergent subsequence:

$$\{x_{n_k}\}_{k=1}^{\infty} \rightarrow x.$$

Claim: x is a cluster point of B .

x is the limit of a sequence in a compact (hence closed) set B so $x \in B$. For each $\varepsilon > 0$, there exists K such that

$$x_{n_k} \in B_{\varepsilon}(x) \text{ for all } k \geq K.$$

Because the x_{n_k} 's are all distinct, at least one is different from x . Q.E.D.

4. (Graph follows)

a. Features of graph:

(i.) All three functions have same value at $p = p^0$;

$$\pi^*(p^0) = \pi(p^0; y^*(p^0), x_1^*(p^0), x_2^*(p^0)) \quad (\text{by definition of LR value function})$$

$$\pi^s(p^0; x_2^*(p^0)) = \pi(p^0; y^s(p^0; x_2^*(p^0)), x_1^s(p^0; x_2^*(p^0)), x_2^*(p^0))$$

(by definition of SR value function)

$$= \pi(p^0; y^*(p^0), x_1^*(p^0), x_2^*(p^0))$$

(because $y^s(p^0; x_2^*(p^0)) = y^*(p^0)$ and $x_1^s(p^0; x_2^*(p^0)) = x_1^*(p^0)$. Same result when optimizing in 2 steps or 1.)

$$(ii.) \pi(p; y^*(p^0), x_1^*(p^0), x_2^*(p^0)) = p \cdot y^*(p^0) - x_1^*(p^0) - x_2^*(p^0)$$

Linear in p with slope $y^*(p^0)$.

(iii.) Curvatures reflect: For $p \neq p^0$:

$$\pi(p; y^*(p^0), x_1^*(p^0), x_2^*(p^0)) \leq \pi^s(p; x_2^*(p^0)) \leq \pi^*(p)$$

because removal of a constraint cannot reduce profit.

b. The curvatures of the three functions imply:

$$\frac{d^2}{dp^2} \pi^*(p^0) \geq \frac{\partial^2}{\partial p^2} \pi^s(p^0; x_2^*(p^0)) \geq \frac{\partial^2}{\partial p^2} \pi(p^0; y^*(p^0), x_1^*(p^0), x_2^*(p^0)) = 0 \quad (*)$$

The Lagrangians for the SR and LR problems are the same -- the only difference is whether x_2 is a choice variable or a parameter:

$$L(y, x_1, x_2; p; \lambda) = p \cdot y - x_1 - x_2 + \lambda(y - f(x_1, x_2))$$

By envelope theorem:

$$\frac{d}{dp} \pi^*(p) = \frac{\partial}{\partial p} L(\text{"*" values}) = y^*(p)$$

$$\frac{\partial}{\partial p} \pi^s(p; x_2^*(p^0)) = \frac{\partial}{\partial p} L(\text{"s" values}) = y^s(p; x_2^*(p^0))$$

Taking a second derivative w.r.t. p and evaluating at $p = p^0$:

$$\frac{d^2}{dp^2} \pi^*(p^0) = \frac{d}{dp} y^*(p^0)$$

$$\frac{\partial^2}{\partial p^2} \pi^s(p^0; x_2^*(p^0)) = \frac{\partial}{\partial p} y^s(p^0; x_2^*(p^0))$$

From inequality (*):

$$\frac{d}{dp} y^*(p^0) \geq \frac{\partial}{\partial p} y^s(p^0; x_2^*(p^0)) \geq 0 \quad (\text{Supply more elastic in LR.})$$

