

**Mathematics for Economists**

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**Solutions to the problem set 4:**

**Question 1:**

$$\begin{aligned} \max f(x, y) &= xy \\ \text{subject to: } x + y &\leq 100 \\ x &\leq 40 \\ x, y &\geq 0 \end{aligned}$$

We first form the Lagrangian:

$$L = xy - \lambda_1(x + y - 100) - \lambda_2(x - 40) + \lambda_3x + \lambda_4y = 0$$

The first order conditions:

$$\frac{\partial L}{\partial x} = 0 \Rightarrow y - \lambda_1 - \lambda_2 + \lambda_3 = 0 \quad (1)$$

$$\frac{\partial L}{\partial y} = 0 \Rightarrow x - \lambda_1 + \lambda_4 = 0 \quad (2)$$

$$\lambda_1(x + y - 100) = 0 \quad (3)$$

$$\lambda_2(x - 40) = 0 \quad (4)$$

$$\lambda_3x = 0, \lambda_4y = 0$$

$$x + y \leq 100, \quad x \leq 40, \quad x, y \geq 0$$

If  $x, y > 0$ , then  $\lambda_3, \lambda_4 = 0$ . Putting this into (1) and (2) we have:

$$x = \lambda_1$$

$$y - x = \lambda_2$$

Using (3) we have:

$$x(x + y - 100) = 0$$

And  $x > 0$ , so first condition is binding and:

$$x + y = 100 \quad (5)$$

Using (4)

$$(y - x)(x - 40) = 0 \Rightarrow \begin{cases} y = x & \text{not valid because of (5) and the fact that } x \leq 40 \\ x = 40 & \text{valid, so the second condition is binding} \end{cases}$$

So  $x^* = 40$ . And using (5) we have

$$y^* = 100 - 40 = 60$$

**Question 2:**

$$x \in \mathbb{R}_+^n, p_i > 0, w > 0 \text{ and}$$

$$p \cdot x \leq w$$

$$\sum_{i=1}^n p_i x_i \leq w$$

Since  $p_i$ 's and  $x_i$ 's are both positive, the maximum value of each  $x_i$  is when we set all the other elements,  $x_j$  for  $j \neq i$  equal to 0, so:

$$p_1 x_1 + p_2 x_2 + \dots + p_n x_n = w, \quad p_i, x_i > 0$$

$$\Rightarrow \bar{x}_i = \max(x_i) = \frac{w}{p_i} \text{ when } x_j = 0 \text{ for all } j \neq i$$

Now what we are looking for is a bound for  $\|x\| = \sum_{i=1}^n x_i^2$ , but we know the upper limit for each  $x_i$  so:

$$\forall x_i : x_i \leq \frac{w}{p_i} \Rightarrow \sum_i x_i^2 \leq \sum_i \frac{w^2}{p_i^2} \Rightarrow \|x\| \leq \frac{w}{\|p\|}$$

So the set B is bounded.

Extra credit:

To prove that a set is closed we should prove that every limit point of the set is in the set. We name the set of  $x \in \mathbb{R}_+^n$  for which we have  $p \cdot x \leq w$  to be A, so:

$$x \in A : p \cdot x \leq w$$

Now assume that  $\hat{x}$  is a limit point of the set A, then there is a sequence  $\{x_n\}_{n \geq 0} \subset A$  such that: (Lecture 7, page 13, theorem iv)

$$\lim_{n \rightarrow \infty} x_n = \hat{x}$$

$$\text{As } \{x_n\}_{n \geq 0} \subset A \Rightarrow p \cdot x_n \leq w, \forall n$$

This gives:

$$\lim_{n \rightarrow \infty} p \cdot x_n \leq w \Rightarrow p \cdot \hat{x} \leq w$$

So  $\hat{x}$  is in the set A, so the set A is closed.

**Question 3:**

a)

$$\max_{(x_1, \dots, x_n) \in \mathbb{R}_+^n} (x_1 \cdot x_2 \cdot \dots \cdot x_n)^{\frac{1}{n}}$$

subject to

$$\sum_{i=1}^n x_i = y$$

i)

We form the Lagrangian first:

$$L(x_1, \dots, x_n, \lambda) = (x_1 \cdot x_2 \cdot \dots \cdot x_n)^{\frac{1}{n}} + \lambda \left( \sum_{i=1}^n x_i - y \right)$$

The objective function is obviously continuous. Moreover, all the  $x_i$ s are positive and sum of them is equal to  $y$ , so:  $0 \leq x_i \leq y$ , so the feasible set is compact and the optimization have solutions. Then we take derivatives of the Lagrangian to get the FOCs.

ii)

$$\frac{\partial L}{\partial x_1} = (x_2 \dots x_n)^{\frac{1}{n}} \cdot \frac{1}{n} x_1^{\frac{1-n}{n}} - \lambda = 0$$

⋮

$$\frac{\partial L}{\partial x_n} = (x_1 \dots x_{n-1})^{\frac{1}{n}} \cdot \frac{1}{n} x_n^{\frac{1-n}{n}} - \lambda = 0$$

$$\sum_{i=1}^n x_i - y = 0$$

iii, iv)

from the first and the second foc:

$$(x_2 \dots x_n)^{\frac{1}{n}} \cdot \frac{1}{n} x_1^{\frac{1-n}{n}} = (x_1, x_3 \dots x_n)^{\frac{1}{n}} \cdot \frac{1}{n} x_2^{\frac{1-n}{n}}$$

$$x_1^{\frac{1-n}{n}} x_2^{\frac{1}{n}} = x_2^{\frac{1-n}{n}} x_1^{\frac{1}{n}} \Rightarrow x_1 = x_2$$

Using the rest of the constraints we get

$$x_1 = x_2 = \dots = x_n$$

And finally using the last constraint we get

$$x_1^* = x_2^* = \dots = x_n^* = \frac{y}{n}$$

And this is the unique solution of the problem, and in this optimal solution the two averages are equal to each other.

$$\text{Arithmetic mean} = \frac{1}{n} \sum_{i=1}^n x_i = \frac{y}{n}$$

$$\text{Geometric mean} = (\prod_{i=1}^n x_i)^{\frac{1}{n}} = \frac{y}{n}$$

b)

$$\max_{(x_1, \dots, x_n) \in R_+^n} \frac{n}{\sum_{i=1}^n \frac{1}{x_i}}$$

subject to

$$\sum_{i=1}^n x_i = y$$

We first form the Langrangian. So:

$$L = \frac{n}{\sum_{i=1}^n \frac{1}{x_i}} + \lambda \left( \sum_{i=1}^n x_i - y \right)$$

$$\frac{\partial L}{\partial x_1} = n \left( \frac{\frac{1}{x_1^2}}{\left( \sum_{i=1}^n \frac{1}{x_i} \right)^2} \right) + \lambda = 0$$

$$\vdots$$

$$\frac{\partial L}{\partial x_n} = n \left( \frac{\frac{1}{x_n^2}}{\left( \sum_{i=1}^n \frac{1}{x_i} \right)^2} \right) + \lambda = 0$$

$$\sum_{i=1}^n x_i = y$$

Just like before, all the constraints we get:

$$x_1^* = x_2^* = \dots = x_n^* = \frac{y}{n}$$

And finally at the optimum point we have:

$$\text{Arithmetic mean} = \frac{1}{n} \sum_{i=1}^n x_i = \frac{y}{n}$$

$$\text{harmonic mean} = \frac{n}{\sum_{i=1}^n \frac{1}{x_i}} = \frac{y}{n}$$

#### Question 4:

a) For convex functions we have:

$$f(\theta a + (1 - \theta)b) \leq \theta f(a) + (1 - \theta)f(b)$$

And for concave functions:

$$f(\theta a + (1 - \theta)b) \geq \theta f(a) + (1 - \theta)f(b)$$

So if the function f is convex and concave at the same time then:

$$f(\theta a + (1 - \theta)b) = \theta f(a) + (1 - \theta)f(b)$$

To prove that the only function that has the above characteristics is the affine function we should prove that:

$$1- f(\alpha x) = \alpha f(x) \text{ and}$$

$$2- f(a + b) = f(a) + f(b)$$

Without loss of generality we assume that  $f(0) = 0$  (if not we can easily assume function g, where  $g(x) = f(x) - f(0)$  and we need the lemma that if function f is both concave and convex then the function  $f(x) - f(0)$  is both convex and concave and then prove the characteristics for function g.)

$$1- f(\alpha x) = f(\alpha x + (1 - \alpha) \cdot 0) = \alpha f(x) + (1 - \alpha)f(0) = \alpha f(x)$$

$$2- f\left(\frac{1}{2} \cdot 2a + \frac{1}{2} \cdot 2b\right) = \frac{1}{2}f(2a) + \frac{1}{2}f(2b) = f(a) + f(b)$$

So f is an affine function.

b) According to the definitions of the quasiconcave and quasiconvex functions:

Quasi concave functions:  $f(\alpha x + (1 - \alpha)y) \geq \min \{f(x), f(y)\}$

Quasi convex functions:  $f(\alpha x + (1 - \alpha)y) \leq \max \{f(x), f(y)\}$

So any increasing or decreasing function  $f$  in  $R_+$  is both quasicconcave and quasicconvex, such as:

$$f(x) = x^2 \quad \text{or} \quad f(x) = e^x \quad \dots$$

c) Assume the point to be:

$$p_0 = (x_0, y_0, z_0)$$

And the plane  $N$  to be:

$$N: Ax + By + Cz + D = 0$$

So the optimization problem is:

$$\begin{aligned} \min_{(x,y,z) \in N} & (x - x_0)^2 + (y - y_0)^2 + (z - z_0)^2 \\ \text{subject to} & Ax + By + Cz + D = 0 \end{aligned}$$

We can equivalently write it as:

$$\begin{aligned} \max_{(x,y,z) \in N} & - [(x - x_0)^2 + (y - y_0)^2 + (z - z_0)^2] \\ \text{subject to} & Ax + By + Cz + D = 0 \end{aligned}$$

Now the objective function is a concave function and the constraint is the set of all the points on a plain which is obviously a convex set.

d)

$f(x)$  is a concave function so:  $f''(x) \leq 0$

$g(y)$  is increasing and concave so  $g'(x) \geq 0$  and  $g''(x) \leq 0$

$$\frac{d}{dx} g(f(x)) = f'(x)g'(f(x))$$

$$\frac{d^2}{d^2x} g(f(x)) = f''(x)g'(f(x)) + (f'(x))^2 g''(f(x))$$

Form the assumptions that  $f(x)$  is a concave and  $g(y)$  is increasing and concave we have:

$$f''(x)g'(f(x)) \leq 0 \quad \text{and}$$

$$(f'(x))^2 g''(f(x)) \leq 0$$

So after all

$$\frac{d^2}{d^2x} g(f(x)) \leq 0$$

And the  $g(f(x))$  is also concave.

### Question 5:

a)

We form the Hessian matrix of the function  $u$ :

$$u(f, c, s) = \alpha_f \ln(f) + \alpha_c \ln(c) + \alpha_s \ln(s)$$

So

$$H_u = \begin{bmatrix} \frac{\partial^2 u}{\partial f^2} & \frac{\partial^2 u}{\partial f \partial c} & \frac{\partial^2 u}{\partial f \partial s} \\ \frac{\partial^2 u}{\partial c \partial f} & \frac{\partial^2 u}{\partial c^2} & \frac{\partial^2 u}{\partial c \partial s} \\ \frac{\partial^2 u}{\partial s \partial f} & \frac{\partial^2 u}{\partial s \partial c} & \frac{\partial^2 u}{\partial s^2} \end{bmatrix} = \begin{bmatrix} -\frac{\alpha_f}{f^2} & 0 & 0 \\ 0 & -\frac{\alpha_c}{c^2} & 0 \\ 0 & 0 & -\frac{\alpha_s}{s^2} \end{bmatrix}$$

Which is obviously negative definite for all the  $(f, c, s)$  in the domain so the function  $u$  is strictly concave.

b)

So the time of the father is allocated to three different tasks:

$$s + c + h_w = 24$$

Where  $h_w$  is the working time. Moreover the income should be equal to the expenses so:

$$h_w \cdot w \geq f$$

So over all

$$(24 - s - c) \cdot w - f \geq 0$$

c)

The feasible set is:

$$g(f, c, s) = (24 - s - c) \cdot w - f \geq 0$$

$$0 < s \leq 24$$

$$0 < c \leq 24$$

$$0 < f \leq 24w$$

d)

$$\max_{(f, c, s) \in \mathbb{R}_+^3} \alpha_f \ln(f) + \alpha_c \ln(c) + \alpha_s \ln(s)$$

Subject to:

$$g(f, c, s) = (24 - s - c) \cdot w - f \geq 0$$

$$s, c, f > 0$$

Now we form the Lagrangian:

$$L = \alpha_f \ln(f) + \alpha_c \ln(c) + \alpha_s \ln(s) + \lambda((24 - s - c) \cdot w - f)$$

Then the first order conditions:

$$\frac{\partial L}{\partial f} = \frac{\alpha_f}{f} - \lambda = 0$$

$$\frac{\partial L}{\partial c} = \frac{\alpha_c}{c} - \lambda w = 0$$

$$\frac{\partial L}{\partial s} = \frac{\alpha_s}{s} - \lambda w = 0$$

$$\lambda[(24 - s - c).w - f] = 0$$

e,f) Obviously none of the non-negativity constraints are binding because if they do then the function u will be infinite ( $\lim_{x \rightarrow 0} \ln(x) = -\infty$ )

Using the first three focs:

$$c = \frac{f \alpha_c}{w \alpha_f}$$

$$s = \frac{f \alpha_s}{w \alpha_f}$$

and  $\lambda \neq 0$  so the budget constraint is binding

$$\left(24 - \frac{f \alpha_s}{w \alpha_f} - \frac{f \alpha_c}{w \alpha_f}\right) w - f = 0$$

$$\frac{24w \alpha_f - f \alpha_s - f \alpha_c}{\alpha_f} - f = 0$$

$$24w \alpha_f = f(\alpha_s + \alpha_c + \alpha_f)$$

$$f^* = \frac{24w \alpha_f}{\alpha_s + \alpha_c + \alpha_f}$$

$$c^* = \frac{24 \alpha_c}{\alpha_s + \alpha_c + \alpha_f}$$

$$s^* = \frac{24 \alpha_s}{\alpha_s + \alpha_c + \alpha_f}$$

The Weierstrass' theorem is not satisfied because the set E of the feasible solutions is not compact, because it is not closed.

We try to answer this question for the variable s, but it is the same for other variables f and c. We know from the form of the utility function that  $\alpha_s$  shows the importance of the sleeping in the father's utility function, and we know that it is always positive so assume that  $\alpha_s = \delta$ , which is a very small amount. So we have:

$$s^* = \frac{24\delta}{\alpha_s + \alpha_f}$$

And we know that utility function u is not continuous at  $(f = 0, c = 0, s = 0)$ , BUT it is always possible to consider small enough  $\varepsilon$  such that:

$$s = \varepsilon < \frac{24\delta}{\alpha_s + \alpha_f}$$

And the utility function will always be continuous at this point, so we will always have an interior solution to our problem no matter how small  $\alpha_s, \alpha_c, \alpha_f$  be.