

## Second Order Conditions for Constrained Optimization

Name

Major

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Let  $f, h$  be  $C^2$  functions on  $R^2$ . To maximize  $f$  on the constraint set  $C_h = \{(x, y) | h(\vec{x}) = c\}$ , Form the Lagrangian

$$\mathcal{L}(\vec{x}, \vec{\mu}) = f(\vec{x}) - \mu[h(\vec{x}) - c].$$

$(x^*, y^*, \mu^*)$  is a local constrained max of  $f$  on  $C_h$  if there exists  $\mu^*$  such that  $(x^*, y^*, \mu^*)$  satisfies:

1. FOC:

$$\frac{\partial \mathcal{L}}{\partial x} = 0, \quad \frac{\partial \mathcal{L}}{\partial y} = 0, \quad \frac{\partial \mathcal{L}}{\partial \mu} = 0 \text{ at } (x^*, y^*, \mu^*)$$

2. Bordered Matrix satisfies:

$$\det \begin{pmatrix} 0 & \frac{\partial h}{\partial x} & \frac{\partial h}{\partial y} \\ \frac{\partial h}{\partial x} & \frac{\partial^2 \mathcal{L}}{\partial x^2} & \frac{\partial^2 \mathcal{L}}{\partial x \partial y} \\ \frac{\partial h}{\partial y} & \frac{\partial^2 \mathcal{L}}{\partial x \partial y} & \frac{\partial^2 \mathcal{L}}{\partial y^2} \end{pmatrix} > 0 \text{ at } (x^*, y^*, \mu^*)$$

Complete the following questions:

1. Suppose  $\frac{\partial h}{\partial x}(x^*, y^*) \neq 0$ . How does Simon and Blume (1994) prove the above Theorem?

2. Now suppose  $\frac{\partial h}{\partial x}(x^*, y^*) = 0$ . Write your own proof assuming instead that  $\frac{\partial h}{\partial y}(x^*, y^*) \neq 0$ .

3. Prove the inequality version of the above theorem:

Let  $f, g$  be  $C^2$  functions on  $R^2$ . To maximize  $f$  on the constraint set  $C_g = \{(x, y) \mid g(\vec{x}) \leq b\}$ , Form the Lagrangian

$$\mathcal{L}(\vec{x}, \vec{\mu}) = f(\vec{x}) - \lambda[g(\vec{x}) - b].$$

Then,  $(x^*, y^*, \lambda^*)$  is a local constrained max of  $f$  on  $C_g$  if there exists  $\lambda^*$  such that  $(x^*, y^*, \lambda^*)$  satisfies:

(a) FOC:

$$\frac{\partial \mathcal{L}}{\partial x} = 0, \quad \frac{\partial \mathcal{L}}{\partial y} = 0, \quad \lambda^* \geq 0, \quad \lambda^*[g(\vec{x}^*) - b] = 0 \text{ at } (x^*, y^*, \mu^*)$$

(b) Suppose  $g$  is binding at  $(x^*, y^*)$ , then Bordered Matrix satisfies:

$$\det \begin{pmatrix} 0 & \frac{\partial g}{\partial x} & \frac{\partial g}{\partial y} \\ \frac{\partial g}{\partial x} & \frac{\partial^2 \mathcal{L}}{\partial x^2} & \frac{\partial^2 \mathcal{L}}{\partial x \partial y} \\ \frac{\partial g}{\partial y} & \frac{\partial^2 \mathcal{L}}{\partial x \partial y} & \frac{\partial^2 \mathcal{L}}{\partial y^2} \end{pmatrix} > 0 \text{ at } (x^*, y^*, \lambda^*)$$

Otherwise, if  $g$  is not binding, the Hessian of the Lagrangian is negative definite, or

$$\frac{\partial^2 \mathcal{L}}{\partial x^2} < 0, \text{ and } \det \begin{pmatrix} \frac{\partial^2 \mathcal{L}}{\partial x^2} & \frac{\partial^2 \mathcal{L}}{\partial x \partial y} \\ \frac{\partial^2 \mathcal{L}}{\partial x \partial y} & \frac{\partial^2 \mathcal{L}}{\partial y^2} \end{pmatrix} > 0 \text{ at } (x^*, y^*, \lambda^*).$$