



MTM4502-Optimization Techniques

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- The conditions known as the KKT Conditions were first published in 1951 by Princeton University professors, American mathematician Harold William Kuhn and Canadian mathematician Albert William Tucker.
-  [H. W. Kuhn and A. W. Tucker. Nonlinear Programming. Proceedings of the Second Berkeley Symposium on Mathematical Statistics and Probability, 481–492, University of California Press, Berkeley, California, 1951.](#)

- Then, over time, it was realized that the necessary conditions of the nonlinear optimization problem were stated in the 1939 master's thesis of William Karush, who was then a graduate student at the University of Chicago.
-  [W. Karush.](#)
Minima of Functions of Several Variables with Inequalities as Side Constraints.
[MSc Thesis, Chicago University, Dept. of Mathematics, Chicago, Illinois.](#)

- The mathematical formulation of the problem of finding the minimum (maximum) of a given function under equality and inequality constraints is as follows:

$$\underset{\mathbf{x} \in \mathbb{R}^n}{\text{minimize}} \quad f(\mathbf{x})$$

$$\text{subject to} \quad g_i(\mathbf{x}) \leq 0, \quad i = 1, \dots, m$$

$$h_j(\mathbf{x}) = 0, \quad j = 1, \dots, \ell$$

- The necessary conditions for the optimal solution of the nonlinear optimization problem are examined in four main conditions:
 - Stationarity Condition:
 - To minimize $f(\mathbf{x})$:

$$\nabla f(\mathbf{x}^*) - \sum_{i=1}^m \mu_i \nabla g_i(\mathbf{x}^*) - \sum_{j=1}^{\ell} \lambda_j \nabla h_j(\mathbf{x}^*) = 0,$$
 - To maximize $f(\mathbf{x})$:

$$-\nabla f(\mathbf{x}^*) - \sum_{i=1}^m \mu_i \nabla g_i(\mathbf{x}^*) - \sum_{j=1}^{\ell} \lambda_j \nabla h_j(\mathbf{x}^*) = 0,$$
 - Primary Feasibility Condition:

$$g_i(\mathbf{x}^*) \leq 0, i = 1, \dots, m$$

$$h_j(\mathbf{x}^*) = 0, j = 1, \dots, \ell$$
 - Dual Feasibility Condition:

$$\mu_i \geq 0, i = 1, \dots, m$$
 - Complementary Slackness Condition:

$$\mu_i g_i(\mathbf{x}^*) = 0, i = 1, \dots, m.$$

- Moreover,
 - if $f(\nu\mathbf{x}_1 + (1 - \nu)\mathbf{x}_2) \leq \nu f(\mathbf{x}_1) + (1 - \nu)f(\mathbf{x}_2)$ holds for any $\mathbf{x}_1 \neq \mathbf{x}_2$ from the domain of the function f with some $\nu \in [0, 1]$, i.e. $f(\mathbf{x})$ is a convex function,
 - if $g_i(\nu_i\mathbf{x}_1 + (1 - \nu_i)\mathbf{x}_2) \leq \nu_i g_i(\mathbf{x}_1) + (1 - \nu_i)g_i(\mathbf{x}_2)$ holds for any $\mathbf{x}_1 \neq \mathbf{x}_2$ from the domain of each g_i function with some $\nu_i \in [0, 1]$, i.e. $g_i(\mathbf{x})$'s are convex functions,
 - if h_i 's are linear functions,

then, these conditions are also sufficient conditions.

Example 1

Examine whether the minimization problem given below satisfies the KKT conditions under the given constraints.

$$\min f(x_1, x_2) = 4x_1^2 + 2x_2^2$$

$$\text{subject to } 3x_1 + x_2 = 8$$

$$2x_1 + 4x_2 \leq 15$$

Example 1

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$$\min f(x_1, x_2) = 4x_1^2 + 2x_2^2$$

$$\text{subject to } 3x_1 + x_2 = 8 \rightarrow h(x_1, x_2) = 3x_1 + x_2 - 8 = 0,$$

$$2x_1 + 4x_2 \leq 15 \rightarrow g(x_1, x_2) = 2x_1 + 4x_2 - 15 \leq 0.$$

- Stationarity Condition:

$$\nabla f - \mu \nabla g - \lambda \nabla h = \begin{bmatrix} 8x_1 - 2\mu - 3\lambda \\ 4x_2 - 4\mu - \lambda \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

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- Primal Feasibility Condition:

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$$\mu \geq 0$$

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- Primal Feasibility Condition:

$$3x_1 + x_2 = 8$$

$$2x_1 + 4x_2 \leq 15$$

- Dual Feasibility Condition:

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$$\mu(2x_1 + 4x_2 - 15) = 0.$$

There will be two cases, based on the dual feasibility condition and the complementary slackness condition:

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- $\mu > 0$

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- The solution will be $x_1 = \frac{17}{10}$, $x_2 = \frac{29}{10}$, $\mu = \frac{53}{25}$ and $\lambda = \frac{78}{25}$.

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- The solution will be $x_1 = \frac{17}{10}$, $x_2 = \frac{29}{10}$, $\mu = \frac{53}{25}$ and $\lambda = \frac{78}{25}$.
- $f(17/10, 29/10) = \frac{1419}{50}$

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- The solution will be $x_1 = \frac{24}{11}$, $x_2 = \frac{16}{11}$, $\mu = 0$ and $\lambda = \frac{64}{11}$.
- $f(24/11, 16/11) = \frac{256}{11} \rightarrow$ Global Minimum ✓

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Here,

- $f(x_1, x_2) = 4x_1^2 + 2x_2^2$ is a convex function,
- $g(x_1, x_2) = 2x_1 + 4x_2 - 15$ is a convex function,
- $h(x_1, x_2) = 3x_1 + x_2 - 8$ is a linear function.

Example 2

Examine whether the following KKT conditions are met for the revenue optimization problem of a company trying to maximize its revenues ($R(Q)$) under a certain minimum profit constraint ($G_{\min} \leq R(Q) - C(Q)$).

$$\min f(Q) = -R(Q)$$

$$\text{subject to } G_{\min} \leq R(Q) - C(Q)$$

$$Q \geq 0$$

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$$\min f(Q) = -R(Q)$$

$$\text{subject to } G_{\min} \leq R(Q) - C(Q) \rightarrow g_1(Q) = G_{\min} - R(Q) + C(Q) \leq 0,$$

$$Q \geq 0$$

$$\rightarrow g_2(Q) = -Q \leq 0.$$

- Stationarity Condition:

$$-\frac{dR}{dQ} - \mu_1 \left[-\frac{dR}{dQ} + \frac{dC}{dQ} \right] - \mu_2(-1) = 0$$

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- Stationarity Condition:

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$$\rightarrow (\mu_1 - 1) \frac{dR}{dQ} - \mu_1 \frac{dC}{dQ} + \mu_2 = 0,$$

- Primal Feasibility Condition:

$$G_{\min} - R(Q) + C(Q) \leq 0,$$

$$-Q \leq 0.$$

- Stationarity Condition:

$$-\frac{dR}{dQ} - \mu_1 \left[-\frac{dR}{dQ} + \frac{dC}{dQ} \right] - \mu_2(-1) = 0$$

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- Primal Feasibility Condition:

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- Dual Feasibility Condition:

$$\mu_1, \mu_2 \geq 0,$$

- Stationarity Condition:

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- Dual Feasibility Condition:

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- Complementary Slackness Condition:

$$\mu_1(G_{\min} - R(Q) + C(Q)) = 0,$$

$$\mu_2 Q = 0.$$

- Stationarity Condition:

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$$\mu_1, \mu_2 \geq 0,$$

- Complementary Slackness Condition:

$$\mu_1(G_{\min} - R(Q) + C(Q)) = 0,$$

$$\mu_2 Q = 0. \rightarrow \mu_2 = 0.$$

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- $\frac{dC}{dQ} = \frac{\mu_1 - 1}{\mu_1} \frac{dR}{dQ},$

Since $\mu_2 = 0$ is satisfied, we have

- $\frac{dC}{dQ} = \frac{\mu_1 - 1}{\mu_1} \frac{dR}{dQ}$,
- which indicates that marginal revenue of the company is greater than its marginal costs.



H. W. Kuhn and A. W. Tucker.

Nonlinear Programming.

[Proceedings of the Second Berkeley Symposium on Mathematical Statistics and Probability](#), 481–492, University of California Press, Berkeley, California, 1951.



W. Karush.

Minima of Functions of Several Variables with Inequalities as Side Constraints.

[MSc Thesis](#), Chicago University, Dept. of Mathematics, Chicago, Illinois.



A. R. Parkinson.

Optimization Techniques in Engineering

[Lecture Notes](#), Brigham Young University, Dept. of Civil Engineering, Provo, Utah.



A. E. Bryson.

Applied Optimal Control

[CRC Press](#), 1975.



Wikipedia.

http://en.wikipedia.org/wiki/KarushKuhnTucker_conditions

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