# MTM4502-Optimization Techniques

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KKT Week - 28/04/2025



### Historical Background

- 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.

### Historical Background

- ➤ 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.

#### **Problem Statement**

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

#### KKT Conditions

- The necessary conditions for the optimal solution of the nonlinear optimization problem are examined in four main conditions:
  - Stationarity Condition:
    - To minimize  $f(\mathbf{x})$ :  $\nabla f(\mathbf{x}^*) - \sum_{i=1}^m \mu_i \nabla g_i(\mathbf{x}^*) - \sum_{i=1}^\ell \lambda_j \nabla h_j(\mathbf{x}^*) = 0,$
    - To maximize  $f(\mathbf{x})$ :  $-\nabla f(\mathbf{x}^*) - \sum_{i=1}^m \mu_i \nabla g_i(\mathbf{x}^*) - \sum_{i=1}^\ell \lambda_j \nabla h_j(\mathbf{x}^*) = 0,$
  - Primary Feasibility Condition:

$$g_i(\mathbf{x}^*) \le 0, i = 1, ..., m$$
  
 $h_i(\mathbf{x}^*) = 0, j = 1, ..., \ell$ 

Dual Feasibility Condition:

$$\mu_i \geq 0, i = 1, ..., m$$

Complementary Slackness Condition:

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



#### **KKT Conditions**

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

then, these conditions are also sufficient conditions.



# Examples

# Example 1

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

min 
$$f(x_1, x_2) = 4x_1^2 + 2x_2^2$$
  
subject to  $3x_1 + x_2 = 8$   
 $2x_1 + 4x_2 \le 15$ 

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# Example 1

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

min 
$$f(x_1, x_2) = 4x_1^2 + 2x_2^2$$
  
subject to  $3x_1 + x_2 = 8$   $\rightarrow h(x_1, x_2) = 3x_1 + x_2 - 8 = 0$ ,  
 $2x_1 + 4x_2 \le 15 \rightarrow g(x_1, x_2) = 2x_1 + 4x_2 - 15 \le 0$ .

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:

$$3x_1 + x_2 = 8 2x_1 + 4x_2 \le 15$$

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Dual Feasibility Condition:

$$\mu \geq \mathbf{0}$$

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Primal Feasibility Condition:

$$3x_1 + x_2 = 8$$
  
$$2x_1 + 4x_2 \le 15$$

- Dual Feasibility Condition:μ > 0
- Complementary Slackness Condition:  $\mu(2x_1 + 4x_2 15) = 0$ .

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

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$$2x_1 + 4x_2 = 15$$

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

$$\begin{vmatrix}
8x_1 - 2\mu - 3\lambda &= 0 \\
4x_2 - 3\mu - \lambda &= 0 \\
3x_1 + x_2 &= 8 \\
2x_1 + 4x_2 &= 15
\end{vmatrix}
\rightarrow
\begin{bmatrix}
8 & 0 & -2 & -3 \\
0 & 4 & -4 & -1 \\
3 & 1 & 0 & 0 \\
2 & 4 & 0 & 0
\end{bmatrix}
=
\begin{bmatrix}
x_1 \\
x_2 \\
\mu \\
\lambda
\end{bmatrix}
=
\begin{bmatrix}
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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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- $f(17/10,29/10) = \frac{1419}{50}$



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► The solution will be  $x_1 = \frac{24}{11}$ ,  $x_2 = \frac{16}{11}$ ,  $\mu = 0$  and  $\lambda = \frac{64}{11}$ .

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0 \\
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\end{bmatrix}$$

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



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- $f(x_1, x_2) = 4x_1^2 + 2x_2^2$  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.

### Examples

# 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  $(G_{\min} \leq R(Q) - C(Q))$ .

$$\min f(Q) = -R(Q)$$
  
subject to  $G_{\min} \le R(Q) - C(Q)$   
 $Q \ge 0$ 

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$$\begin{aligned} & \text{min } f(Q) = -R(Q) \\ & \text{subject to } G_{\text{min}} \leq R(Q) - C(Q) \rightarrow g_1(Q) = G_{\text{min}} - R(Q) + C(Q) \leq 0, \\ & Q \geq 0 \qquad \qquad \rightarrow g_2(Q) = -Q \leq 0. \end{aligned}$$

Stationarity Condition:

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

Stationarity Condition:

$$\begin{aligned}
&-\frac{dR}{dQ} - \mu_1 \left[ -\frac{dR}{dQ} + \frac{dC}{dQ} \right] - \mu_2 (-1) = 0 \\
&\to (\mu_1 - 1) \frac{dR}{dQ} - \mu_1 \frac{dC}{dQ} + \mu_2 = 0,
\end{aligned}$$

Stationarity Condition:

$$\begin{split} &-\frac{dR}{dQ} - \mu_1 \bigg[ -\frac{dR}{dQ} + \frac{dC}{dQ} \bigg] - \mu_2 (-1) = 0 \\ &\rightarrow (\mu_1 - 1) \frac{dR}{dQ} - \mu_1 \frac{dC}{dQ} + \mu_2 = 0, \end{split}$$

Primal Feasibility Condition:

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

Stationarity Condition:

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Primal Feasibility Condition:

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

Dual Feasibility Condition:

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

Stationarity Condition:

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Primal Feasibility Condition:

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

- Dual Feasibility Condition:  $\mu_1, \mu_2 > 0$ ,
- ► Complementary Slackness Condition:  $\mu_1(G_{\min} R(Q) + C(Q)) = 0$ ,  $\mu_2 Q = 0$ .



Stationarity Condition:

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Primal Feasibility Condition:

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- ▶ Dual Feasibility Condition:  $\mu_1, \mu_2 > 0$ ,
- Complementary Slackness Condition:

$$\mu_1(G_{\min} - R(Q) + C(Q)) = 0,$$
 $\mu_2 Q = 0. \rightarrow \mu_2 = 0.$ 



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

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

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- which indicates that marginal revenue of the company is greater than its marginal costs.