

## KARUSH-KUHN-TUCKER CONDITIONS

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### INTRODUCTION

IN THIS SECTION WE INTRODUCE THE FAMOUS KARUSH-KUHN-TUCKER (K.K.T.) CONDITIONS FOR THE CONSTRAINED NON-LINEAR OPTIMIZATION PROBLEM. WE WILL SEE:

1. UNDER SOME ASSUMPTIONS, THE KKT CONDITIONS IS NECESSARY FOR THE (LOCAL) OPTIMALITY.
2. IF OUR PROBLEM IS CONVEX, THE KKT CONDITIONS IS SUFFICIENT AND NECESSARY FOR THE GLOBAL OPTIMALITY.

### PROBLEMS WITH ONLY INEQUALITY CONSTRAINTS

CONSIDER PROBLEM

$$\begin{aligned} & \min f(\underline{x}) \\ \text{SPACE OF CONTINUOUSLY} \\ \text{DIFFERENTIABLE FUNCTIONS} \\ \text{s.t. } g_i(\underline{x}) \leq 0 & \quad i \in I = \{1, \dots, m\} \end{aligned}$$

WHERE  $f, g_i \in C^1(\mathbb{R}^n)$ ,  $\underline{x} \in \mathbb{R}^n$ , AND WITH THE ASSUMPTION: THE FEASIBLE REGION IS NOT EMPTY, BUT ITS INTERIOR CAN BE EMPTY, MATHEMATICALLY SPEAKING,

$$\text{FEASIBLE REGION} \rightarrow S = \{\underline{x} \in \mathbb{R}^n : g_i(\underline{x}) \leq 0, \forall i \in I\} \neq \emptyset$$

DEF 1 FOR EACH  $\underline{x} \in S$ , WE DEFINE:

- CONE OF THE FEASIBLE DIRECTIONS AS

$$\Phi(\underline{x}) := \left\{ \underline{d} \in \mathbb{R}^n : \exists \tilde{\alpha} > 0 \text{ s.t. } \underline{x} + \alpha \underline{d} \in S, \forall \alpha \in [0, \tilde{\alpha}] \right\}$$

- SET OF INDICES OF ACTIVE CONSTRAINTS

WHY DO WE DEFINE  $\Phi(\underline{x})$ ?  
BECAUSE GRADIENT IS  
THE DIRECTION WITH  
WHICH THE FUNCTION

- SET OF INDICES OF ACTIVE CONSTRAINTS

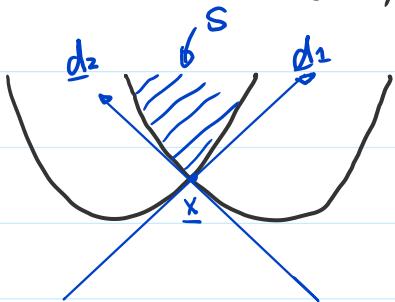
$$I(\underline{x}) := \{ i \in I : g_i(\underline{x}) = 0 \} \subseteq I$$

DEFINITE GRADIENT IN THE DIRECTION WITH WHICH THE FUNCTION GROWS MOST FASTLY, SO WE WANT TO CHOOSE THE DIREC ST. THE FUNCTION POSSIBLY DECREASES, IN THIS WAY  $g_i(\underline{x})$  SO STILL HOLDS

- CONE OF THE DIRECTIONS CONSTRAINED BY THE GRADIENTS OF THE ACTIVE CONSTRAINTS

$$D(\underline{x}) := \{ \underline{d} \in \mathbb{R}^n : \nabla^T g_i(\underline{x}) \cdot \underline{d} \leq 0, \forall i \in I(\underline{x}) \}$$

**REMARK** THE CONE  $\Phi(\underline{x})$  CAN BE(TOPOLOGICALLY) OPEN, FOR EXAMPLE



OBVIOUSLY  $d_1$  AND  $d_2$  ARE NEITHER FEASIBLE DIRECTIONS FOR  $\underline{x}$ .

VECTOR IN  $\mathbb{R}^n$ , THEN WE DISCUSS OPEN AND CLOSED WITH RESPECT TO CANONICAL TOPOLOGY OF  $\mathbb{R}^n$ .

**Prop 2**  $\overline{\Phi(\underline{x})} \subseteq D(\underline{x})$ .

**Proof** GIVEN  $\underline{d} \in \Phi(\underline{x})$ , FOR SUFFICIENTLY SMALL VALUES OF  $\alpha$  WE HAVE

$$0 \geq g_i(\underline{x} + \alpha \underline{d}) = g_i(\underline{x}) + \alpha \nabla^T g_i(\underline{x}) \cdot \underline{d} + o(\alpha)$$

FOR ALL  $i \in I(\underline{x})$  WITH  $g_i(\underline{x}) = 0$ . THEREFORE

$$\nabla^T g_i(\underline{x}) \cdot \underline{d} \leq 0 \leftarrow \begin{array}{l} \text{LET } \alpha \rightarrow 0^+ : \\ 0 \geq \lim_{\alpha \rightarrow 0^+} \frac{g_i(\underline{x} + \alpha \underline{d})}{\alpha} = \nabla^T g_i(\underline{x}) \cdot \underline{d} \end{array}$$

FOR ALL  $i \in I(\underline{x})$ , THAT IS  $\underline{d} \in D(\underline{x})$ . SO  $\overline{\Phi(\underline{x})} \subseteq D(\underline{x})$ .

MOREOVER, ACCORDING TO THE DEFINITION OF  $D(\underline{x})$ ,  $D(\underline{x})$  MUST BE CLOSED SET. HENCE  $\overline{\Phi(\underline{x})} \subseteq D(\underline{x})$ . ■

**THEOREM 3** [EXTENSION OF THE FIRST ORDER NECESSARY OPT. CONDS.]

IF  $f \in C^1(S)$  AND  $\tilde{\underline{x}} \in S$  IS A LOCAL MINIMUM OF  $f$  ON  $S$ ,

$$\therefore \nabla^T f(\tilde{\underline{x}}) \cdot \underline{d} \geq 0 \quad \forall \underline{d} \in \overline{D(\tilde{\underline{x}})}$$

IF  $f \in \mathcal{C}^1(S)$  AND  $\underline{x} \in S$  IS A LOCAL MINIMUM OF  $f$  ON  $S$ ,  
 THEN  $\nabla f(\underline{x}) \underline{d} \geq 0$ , FOR ALL  $\underline{d} \in \overline{\mathcal{D}(\underline{x})}$ , THAT IS, ALL THE  
 FEASIBLE DIRECTIONS ARE ASCENT DIRECTIONS.

**Proof** WE ALREADY KNOW THE RESULT HOLDS FOR ALL  $\underline{d} \in \mathcal{D}(\underline{x})$ ,  
 IT SUFFICES TO EXTEND IT TO THE CLOSURE. FOR ALL  $\underline{d} \in \overline{\mathcal{D}(\underline{x})}$ ,  
 THERE EXISTS  $\{\underline{d}_k\}_{k \in \mathbb{N}}$  SUCH THAT  $\lim_{k \rightarrow \infty} \underline{d}_k = \underline{d}$ . SINCE  
 $\nabla f^T(\underline{x}) \underline{d}_k \geq 0$  FOR ALL  $k \in \mathbb{N}$ , THEN

$$\lim_{k \rightarrow \infty} \nabla f^T(\underline{x}) \underline{d}_k = \nabla f^T(\underline{x}) \underline{d} \geq 0$$

BECAUSE  $\nabla f \in \mathcal{C}^0(S)$ , RECALL THAT  $f \in \mathcal{C}^1(S)$ . ■

NOTICE THAT THESE NECESSARY CONDITIONS CAN BE HARDLY USED  
 BECAUSE  $\overline{\mathcal{D}(\underline{x})}$  IS DIFFICULT TO CHARACTERIZE. SINCE  $D(\underline{x})$   
 IS WELL-CRITERIALIZED, WE INTRODUCE FURTHER CONDITIONS  
 ON THE SET OF CONSTRAINTS.

**DEF4** [ZANGWILL CONSTRAINT QUALIFICATION, C.Q.]

THE CONSTRAINT QUALIFICATION ASSUMPTION HOLDS  
 AT  $\underline{x} \in S$  IF  $\overline{\mathcal{D}(\underline{x})} = D(\underline{x})$ .

**THEOREM 5** [KARUSH-KUHN-TUCKER NECESSARY OPTIMALITY CONDS]

SUPPOSE  $f, g_i \in \mathcal{C}^1(\mathbb{R}^n)$  AND THE C.Q. ASSUMPTION HOLDS  
 AT  $\widetilde{x} \in S = \{\underline{x} \in \mathbb{R}^n : g_i(\underline{x}) \leq 0, \forall i \in I\}$ . IF  $\widetilde{x}$  IS A LOCAL  
 MINIMUM OF  $f$  OVER  $S$ , THEN  $\exists u_1, \dots, u_m \geq 0$  (REFERRED  
 AS KKT-MULTIPLIERS) SUCH THAT:

$$-\nabla f(\widetilde{x}) - \sum_{i=1}^m u_i \nabla g_i(\widetilde{x}) = 0$$

AS KKT-MULTIPLIERS) SUCH THAT:

$$\nabla f(\tilde{x}) + \sum_{i \in I(\tilde{x})} u_i \nabla g_i(\tilde{x}) = 0$$

OR EQUIVALENTLY,

$$\begin{cases} \nabla f(\tilde{x}) + \sum_{i=1}^m u_i \nabla g_i(\tilde{x}) = 0 \\ u_i g_i(\tilde{x}) = 0, \quad \forall i \in I \end{cases}$$

**PROOF** ASSUMING Q.Q. ASSUMPTIONS HOLDS AT  $\tilde{x}$ , WE HAVE  $\overline{\Phi}(\tilde{x}) = D(\tilde{x})$ .

A NECESSARY CONDITION FOR  $\tilde{x}$  TO BE A LOCAL MINIMUM

OF  $f$  OVER  $S$  IS: ORIGINALLY WE HAVE  $(x)$  HOLDS FOR  $d \in \overline{\Phi}(\tilde{x})$ ,  
SINCE WE HAVE  $(x)$ , THEN  $\overline{\Phi}(\tilde{x}) = D(\tilde{x})$ , WE GET THIS:

$$(*) \quad \nabla f^T(\tilde{x}) d \geq 0, \quad \forall d \text{ s.t. } \nabla g_i^T(\tilde{x}) d \leq 0, \quad \forall i \in I(\tilde{x})$$

NOW RECALL THE FARKAS LEMMA:

$$\begin{cases} A\bar{u} = \bar{b} \\ \bar{u} \geq 0 \end{cases} \text{ HAS A SOLUTION} \Leftrightarrow \begin{cases} \bar{b}^T d \geq 0 \\ \forall d \text{ s.t. } d^T A \geq 0 \end{cases} \text{ HAS A SOLUTION}$$

TAKING  $\bar{b} = \nabla f(\tilde{x})$  AND  $A = (-\nabla g_{i_1}(\tilde{x}), \dots, -\nabla g_{i_p}(\tilde{x}))$  MATRIX  
WITH  $n$  ROWS AND  $p := |I(\tilde{x})|$  COLUMNS, AND ACCORDING  
TO FARKAS LEMMA,  $(*)$  IS EQUIVALENT TO

$$\begin{cases} A\bar{u} = \bar{b} \\ \bar{u} \geq 0 \end{cases} \Leftrightarrow \begin{cases} \sum_{i \in I(\tilde{x})} u_i (-\nabla g_i(\tilde{x})) = \nabla f(\tilde{x}) \\ u \geq 0 \end{cases}$$

WHICH COMPLETES THE PROOF.

■

**REMARK** [GEOMETRIC INTERPRETATION]

FOR A LOCAL MINIMUM  $\tilde{x}$ ,  $-\nabla f(\tilde{x})$  MUST BE EXPRESSIBLE

AS A LINEAR COMBINATION WITH COEFFICIENTS  $u_i \geq 0$

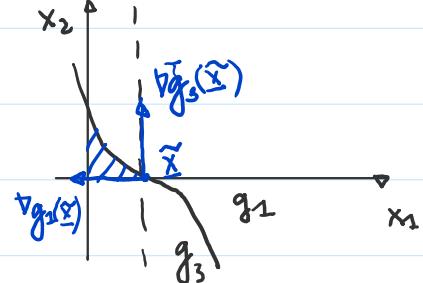
OF THE  $\nabla g_i(\tilde{x})$ , FOR ALL  $i \in I(\tilde{x})$ , I.E., MUST FORM AN

OBTUSE ANGLE WITH ALL THE FEASIBLE DIRECTIONS.

OF THE  $\nabla g_i(\underline{x})$ , FOR ALL  $i \in I(\underline{x})$ , I.E., MUST FORM AN OBTUSE ANGLE WITH ALL THE FEASIBLE DIRECTIONS.

**REMARK** IF C.Q. ASSUMPTION DOES NOT HOLD AT  $\tilde{\underline{x}}$ , THEN THE KKT CONDITIONS NEED NOT BE NECESSARY FOR  $\tilde{\underline{x}}$  TO BE A LOCAL OPTIMAL SOLUTION. FOR EXAMPLE, CONSIDER

$$\begin{aligned} \min f(\underline{x}) &= -x_1 \\ \text{s.t. } g_1(\underline{x}) &= -x_1 \leq 0 \\ g_2(\underline{x}) &= -x_2 \leq 0 \\ g_3(\underline{x}) &= -(1-x_1)^3 + x_2 \leq 0. \end{aligned}$$



WE CAN OBSERVE THAT GLOBAL MINIMAL IS OBTAINED AT  $\tilde{\underline{x}}$ , THEN  $\overline{\Phi(\tilde{\underline{x}})} = \{(2,0) : \alpha \leq 0\}$ . ON THE OTHER HAND

$$\nabla^T g_1(\tilde{\underline{x}}) = (-1, 0), \quad \nabla^T g_3(\tilde{\underline{x}}) = (0, 1)$$

THEN OBVIOUSLY  $D(\tilde{\underline{x}}) \neq \overline{\Phi(\tilde{\underline{x}})}$ .

NOW,  $f$  OBTAINS THE GLOBAL MINIMUM AT  $\tilde{\underline{x}} = (1,0)$  BUT NO MULTIPLIERS  $u_1, u_2, u_3 \geq 0$  SUCH THAT  $\nabla f(\tilde{\underline{x}}) + \sum_{i \in I(\tilde{\underline{x}})} u_i \nabla g_i(\tilde{\underline{x}}) = 0$ . INDEED  $u_1 = 0$ , AND

$$\begin{pmatrix} -1 \\ 0 \end{pmatrix} + u_2 \begin{pmatrix} 0 \\ -1 \end{pmatrix} + u_3 \begin{pmatrix} 0 \\ 1 \end{pmatrix} \neq \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \quad \forall u_2, u_3 \geq 0$$

THAT IS,  $\tilde{\underline{x}}$  DOES NOT SATISFY THE KKT CONDITIONS.

SINCE C.Q. ASSUMPTION IS DIFFICULT TO VERIFY IN PRACTICE, THERE EXIST SUFFICIENT CONDITIONS FOR C.Q. TO HOLD.

### PROP 6 [SUFFICIENT CONDITIONS FOR CONSTRAINT QUALIFICATION]

IF 1) ALL  $g_i$  ARE LINEAR FUNCTIONS. OR

2) ALL  $g_i$ 's ARE CONVEX AND THERE EXISTS A  $\underline{x} \in \mathbb{R}^n$

- IF 1) ALL  $g_i$  ARE LINEAR FUNCTIONS. OR  
 2) ALL  $g_i$ 'S ARE CONVEX AND THERE EXISTS  $\underline{a} \in \mathbb{R}^n$   
 SUCH THAT  $g_i(\underline{a}) < 0$  FOR ALL  $i \in I$ .

THEN THE C.Q. ASSUMPTION HOLDS AT EVERY FEASIBLE POINT  $\underline{x} \in S$ .

IF 3)  $\nabla g_i(\underline{x})$  ARE LINEARLY INDEPENDENT FOR ALL  $i \in I(\underline{x})$ .

THEN THE C.Q. ASSUMPTION HOLDS AT  $\underline{x} \in S$ .

**REMARK** IN 3), WHEN THE GRADIENTS OF THE ACTIVE CONSTRAINTS ARE LINEAR INDEPENDENT, THE KKT MULTIPLIER VECTOR IS UNIQUE.

THEN WE REVEAL THE IMPORTANCE OF KKT CONDITIONS IN CONVEX PROBLEMS:

**THEOREM 7** [NECESSARY AND SUFFICIENT CONDITIONS FOR GLOBAL OPTIMALITY IN CONVEX PROBLEMS]

IF  $f \in \mathcal{C}^1(\mathbb{R}^n)$  IS CONVEX,  $g_i \in \mathcal{C}^1(\mathbb{R}^n)$  IS CONVEX FOR ALL  $i \in I$ , AND THERE EXISTS  $\underline{a}$  SUCH THAT  $g_i(\underline{a}) < 0$  FOR ALL  $i \in I$ , THEN  $\underline{x}^* \in S$  IS A GLOBAL MINIMUM IF AND ONLY IF THERE EXIST  $u_1, \dots, u_m \geq 0$  SUCH THAT

$$\begin{cases} \nabla f(\underline{x}^*) + \sum_{i=1}^m u_i \nabla g_i(\underline{x}^*) = \underline{0} \\ u_i g_i(\underline{x}^*) = 0, \quad \forall i \in I. \end{cases}$$

**Proof** " $\Rightarrow$ ":  $\underline{x}^*$  (LOCAL) MINIMUM  $\Rightarrow \underline{x}^*$  SATISFIES KKT CONDITIONS

(NOTICE WE HAVE C.Q. CONDITION, THEN KKT ARE NECESSARY)

" $\Leftarrow$ ": KKT CONDITIONS  $\Rightarrow \nabla f^T(\underline{x}^*) \underline{d} \geq 0, \forall \underline{d} \in \overline{\text{ef}(x^*)} = D(\underline{x}^*)$

(BY C.Q. CONDITION AND FARKAS LEMMA)

$\Rightarrow \underline{x}^*$  (GLOBAL) MINIMUM (THM FOR CONVEX PROBLEMS)

~ CONVEX FUNCTION AND TAYLOR'S APPROXIMATION

$\Rightarrow \underline{x}^*$  (GLOBAL) MINIMUM (THM FOR CONVEX PROBLEMS)

## PROBLEM IN GENERAL CASE

CONSIDER:

$$\min f(\underline{x})$$

$$\text{s.t. } g_i(\underline{x}) \leq 0 \quad i \in I = \{1, \dots, m\}$$

$$h_l(\underline{x}) = 0 \quad l \in L = \{1, \dots, p\}$$

$$\underline{x} \in X \subseteq \mathbb{R}^n$$

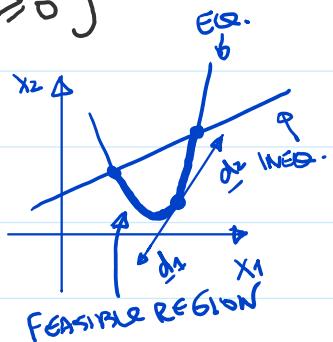
WHERE  $f, g_i, h_l \in C^1$  FOR ALL  $i \in I$ , ALL  $l \in L$ .

OBSERVE THAT: IN THE PRESENCE OF NON-LINEAR EQUALITY CONSTRAINTS, USUALLY  $\emptyset(\underline{x}) = \{0\}$ . TO EXTEND THE PREVIOUS RESULTS, WE DEFINE A CONE OF DIRECTIONS THAT ALSO TAKES INTO ACCOUNT THE EQUALITY CONSTRAINTS.

**DEF 8** CLOSED CONE OF THE TANGENTS AT  $\underline{x}$  IS DEFINED AS:

$$T(\underline{x}) := \left\{ \underline{d} \in \mathbb{R}^n : \underline{d} = \lambda \lim_{k \rightarrow \infty} \frac{\underline{x}_k - \underline{x}}{\|\underline{x}_k - \underline{x}\|}, \lambda \geq 0 \right\}$$

WHERE  $\{\underline{x}_k\}_{k \in \mathbb{N}}$  ARE FEASIBLE SOLUTIONS.



**DEF 9** [ABADIE CONSTRAINT QUALIFICATION]

THE C.Q. ASSUMPTION HOLDS AT  $\underline{x} \in S$  IF

$$T(\underline{x}) = D(\underline{x}) \cap H(\underline{x})$$

WHERE  $D(\underline{x}) = \{ \underline{d} \in \mathbb{R}^n : \nabla g_i^T(\underline{x}) \underline{d} \leq 0, \forall i \in I(\underline{x}) \}$

AND  $H(\underline{x}) = \{ \underline{d} \in \mathbb{R}^n : \nabla h_l^T(\underline{x}) \underline{d} = 0, \forall l \in L \}$ .

AND  $H(\underline{x}) = \{ \underline{d} \in \mathbb{R}^n : \nabla f^T_h(\underline{x}) \underline{d} = 0, \forall h \in L \}$ .

### THEOREM 10 [GENERAL KKT NECESSARY OPTIMALITY CONDITIONS]

SUPPOSE  $f \in \mathcal{C}^1$ ,  $g_i \in \mathcal{C}^1$  FOR ALL  $i \in I$ ,  $h_l \in \mathcal{C}^1$  FOR ALL  $l \in L$  AND THE (ABADIE) C.Q. ASSUMPTION HOLDS AT  $\tilde{\underline{x}} \in S$ .

IF  $\tilde{\underline{x}}$  IS A LOCAL MINIMUM OF  $f$  OVER  $S$ , THEN THERE EXIST  $u_i \geq 0$  FOR ALL  $i \in I(\tilde{\underline{x}})$  AND  $v_l \in \mathbb{R}$  FOR ALL  $l \in L$  SUCH THAT

$$\nabla f(\tilde{\underline{x}}) + \sum_{i \in I(\tilde{\underline{x}})} u_i \nabla g_i(\tilde{\underline{x}}) + \sum_{l \in L} v_l \nabla h_l(\tilde{\underline{x}}) = \underline{0}$$

**Proof** PLEASE REFER TO BOOK LINEAR AND NON-LINEAR PROGRAMMING, D. G. LUENBERGER, Y. YE, SPRINGER, THIRD EDITION, PAGE 342. NOTICE ON THE BOOK THE PROOF IS UNDER A SUFFICIENT ASSUMPTION FOR OBTAINING (ABADIE) C.Q. CONDITION. ■

**REMARK** IN THE CASE THAT ALL CONSTRAINTS ARE EQUALITY CONSTRAINTS THE KKT CONDITIONS COINCIDE WITH THE **CLASSICAL LAGRANGE OPTIMITY CONDITION**.

STILL, (ABADIE) C.Q. CONDITION IS DIFFICULT TO CHARACTERIZE, HERE, AT THE END, WE GIVE A SUFFICIENT CONDITION.

### PROP 11 [SUFFICIENT CONDITION FOR C.Q.]

- 1) C.Q. ASSUMPTION HOLDS AT EVERY  $\tilde{\underline{x}} \in S$  IF  $g_i$  ARE CONVEX, FOR ALL  $i \in I$ ; AND  $h_l$  ARE LINEAR, FOR ALL  $l \in L$ ; AND

FOR ALL  $i \in I$ ; AND  $h_l$  ARE LINEAR, FOR ALL  $l \in L$ ; AND  
THERE EXISTS  $\underline{a} \in X \subset \mathbb{R}^n$  SUCH THAT  $g_i(\underline{a}) < 0$  FOR ALL  
 $i \in I$  AND ALL  $h_l(\underline{a}) = 0$  FOR ALL  $l \in L$ .

- 2) C.Q. ASSUMPTION HOLDS AT  $\tilde{x} \in S$  IF  $\nabla g_i(\tilde{x}), \forall i \in I(\tilde{x})$   
AND  $\nabla h_l(\tilde{x}), \forall l \in L$  ARE LINEAR INDEPENDENT.