# **Constraint Qualifications**

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(Calculus 4, 19.5)

# What Happens when NDCQ Fails?

• Critical Points where NDCQ Fails are also candidates for the solution!

- Need to check those points (as well as FOC)

- Can we incorporate this into our Lagrangian as well (just like binding/non-binding cases)?
   – Yes! Add multiplier μ<sub>0</sub> to the objective function!
- Actually, there are other CQ's we can use
   NDCQ is only one of them...

# Theorem 19.10 (CQ: Equality Constraint)

- Suppose f, h are  $C^1$  functions on  $\mathbf{R}^2$
- $\vec{x}^* = (x_1^*, x_2^*)$  solves  $\max \left\{ f(x_1, x_2) \middle| h(x_1, x_2) = c \right\}$
- Form the Lagrangian (with multiplier  $\mu_0$  for f)
  - $\mathcal{L} = \mu_0 f(x_1, x_2) \mu_1 [h(x_1, x_2) c]$
- There exists μ<sup>\*</sup> = (μ<sub>0</sub><sup>\*</sup>, μ<sub>1</sub><sup>\*</sup>) such that

  a) (μ<sub>0</sub><sup>\*</sup>, μ<sub>1</sub><sup>\*</sup>) ≠ (0,0)
  b) μ<sub>0</sub><sup>\*</sup> = 0 or 1
  c) FOC satisfied at (x<sub>1</sub><sup>\*</sup>, x<sub>2</sub><sup>\*</sup>, μ<sub>0</sub><sup>\*</sup>, μ<sub>1</sub><sup>\*</sup>)

Theorem 19.10 (CQ: Equality Constraint)  $\mathcal{L} = \mu_0 f(x_1, x_2) - \mu_1 [h(x_1, x_2) - c]$ 

- There exists  $(\mu_0^*, \mu_1^*) \neq (0, 0)$  s.t.  $\mu_0^* = 0$  or 1
- FOC satisfied at  $(x_1^*, x_2^*, \mu_0^*, \mu_1^*)$

$$\frac{\partial \mathcal{L}}{\partial x_1} = \underbrace{\mu_0} \frac{\partial f}{\partial x_1}(x_1, x_2) - \mu_1 \frac{\partial h}{\partial x_1}(x_1, x_2) = 0$$
$$\frac{\partial \mathcal{L}}{\partial x_2} = \underbrace{\mu_0} \frac{\partial f}{\partial x_2}(x_1, x_2) - \mu_1 \frac{\partial h}{\partial x_2}(x_1, x_2) = 0$$
$$\frac{\partial \mathcal{L}}{\partial \mu_1} = c - h(x_1, x_2) = 0$$

# Example 19.9 (CQ: Equality Constraint)

- max  $\{f(x,y) = x | h(x,y) = x^3 + y^2 = 0\}$
- Form Lagrangian  $\mathcal{L} = \mu_0 x \mu_1 [x^3 + y^2 0]$
- There exists  $(\mu_0^*,\mu_1^*) \neq (0,0)$  s.t.  $\mu_0^* = 0$  or 1
- FOC:  $\frac{\partial \mathcal{L}}{\partial x} = \mu_0 \mu_1 [3x^2] = 0$   $\frac{\partial \mathcal{L}}{\partial y} = -\mu_1 [2y] = 0$ (No solution w/o multiplier  $\mu_0$ )  $\frac{\partial \mathcal{L}}{\partial \mu_1} = -[x^3 + y^2] = 0$

# Theorem 19.11 (Fritz John Theorem)

- Suppose  $f, g_1, \ldots, g_k$  are  $C^1$  functions on  $\mathbf{R}^n$
- Let  $\vec{x}^* = (x_1^*, \cdots, x_n^*)$  be a local maximizer of  $\max \left\{ f(x_1, \cdots, x_n) \middle| g_1(x_1, \cdots, x_n) \le b_1, \\ \cdots, g_k(x_1, \cdots, x_n) \le b_k \right\}$
- Form the Lagrangian (with multiplier  $\lambda_0$  for f)  $\mathcal{L} = \lambda_0 f(x_1, \cdots, x_n) - \lambda_1 [g_1(x_1, \cdots, x_n) - b_1]$   $- \cdots - \lambda_k [g_k(x_1, \cdots, x_n) - b_k]$ • There exists  $\vec{\lambda}^* = (\lambda_0^*, \lambda_1^*, \cdots, \lambda_k^*)$  such that

Theorem 19.11 (Fritz John Theorem)  $\mathcal{L} = \lambda_0 f(x_1, \cdots, x_n) - \lambda_1 [g_1(x_1, \cdots, x_n) - b_1]$  $-\cdots -\lambda_k[g_k(x_1,\cdots,x_n)-b_k]$ a)  $\frac{\partial \mathcal{L}}{\partial x_1}(\vec{x}^*, \vec{\lambda}^*) = 0, \cdots, \frac{\partial \mathcal{L}}{\partial x_n}(\vec{x}^*, \vec{\lambda}^*) = 0$ b)  $\lambda_1^*[g_1(\vec{x}^*) - b_1] = 0, \cdots, \lambda_k^*[g_k(\vec{x}^*) - b_k] = 0$ c)  $\lambda_1^* \ge 0, \cdots, \lambda_k^* \ge 0$ d)  $g_1(\vec{x}^*) - b_1 \leq 0, \cdots, g_k(\vec{x}^*) - b_k \leq 0$ e)  $\lambda_0^* = 0$  or 1 (If NDCQ holds, set  $\lambda_0^* = 1$ !) f)  $(\lambda_0^*, \lambda_1^*, \cdots, \lambda_k^*) \neq (0, 0, \cdots, 0)$ 

# Theorem 19.11 (Fritz John Theorem)

• If NDCQ fails, row vectors linearly dependent:  $Dg_i = \nabla g_i = \left(\frac{\partial g_i}{\partial x_1}(\vec{x}^*), \cdots, \frac{\partial g_i}{\partial x_n}(\vec{x}^*)\right)$ 

• So, for binding constraints, exists  $(a_1, \cdots, a_{k_0}) \neq (0, \cdots, 0)$ 

$$a_{1}\begin{pmatrix}\frac{\partial g_{1}}{\partial x_{1}}(\vec{x}^{*})\\\vdots\\\frac{\partial g_{1}}{\partial x_{n}}(\vec{x}^{*})\end{pmatrix}+\dots+a_{k_{0}}\begin{pmatrix}\frac{\partial g_{k_{0}}}{\partial x_{1}}(\vec{x}^{*})\\\vdots\\\frac{\partial g_{k_{0}}}{\partial x_{n}}(\vec{x}^{*})\end{pmatrix}=\vec{0}$$
  
Set  $(\lambda_{0}^{*},\lambda_{1}^{*},\dots,\lambda_{k}^{*})=(0,a_{1},\dots,a_{k_{0}},0,\dots,0)!$ 

### Generalized Example 18.9

$$\max f(x, y) = xy$$
  
s.t.  $(x + y - I)^3 \le 0$   
 $x \ge 0, y \ge 0$ 

- Same as constraint  $x + y I \leq 0$
- But does NDCQ hold at  $(x^*, y^*) = \left(\frac{I}{2}, \frac{I}{2}\right)$ ? For  $g(x, y) = (x + y I)^3$   $\frac{\partial g}{\partial x} = 3(x + y I)^2$   $\frac{\partial g}{\partial y} = 3(x + y I)^2$ For NDCQ fails!  $\vec{\nabla}g\left(\frac{I}{2},\frac{I}{2}\right) = (0,0)$ **Constraint Qualifications**

# Fritz John Thm on Generalized Example 18.9 • $\tilde{\mathcal{L}} = \lambda_0 xy - \lambda_1 [x + y - I]^3$ has FOC: $\partial ilde{\mathcal{L}}$ $\frac{\partial \mathcal{L}}{\partial x} = \lambda_0 y - 3\lambda_1 [x + y - I]^2 \le 0, x \cdot \frac{\partial \mathcal{L}}{\partial x} = 0$ $\partial \mathcal{ ilde{L}}$ $\frac{\partial \mathcal{L}}{\partial y} = \lambda_0 x - 3\lambda_1 [x + y - I]^2 \le 0, y \cdot \frac{\partial \tilde{\mathcal{L}}}{\partial u} = 0$ $\partial \tilde{\mathcal{L}}$ $\frac{\partial \mathcal{L}}{\partial \lambda_1} = -[x+y-I]^3 \ge 0, \lambda_1 \cdot \left(-[x+y-I]^3\right) = 0$ • Hence, $(x^*, y^*) = (0, 0)$ or $\lambda_0^* = 0 \implies \lambda_1^* > 0)$ $\lambda_1^* = 0 \implies \lambda_0^* = 1$ or $[x^* + y^* - I]^3 = 0$

### Generalize Example 18.9

• Need to check  $(x^*, y^*, \lambda_0^*, \lambda_1^*) = (0, 0, 0, \lambda_1^*)$ - But this yields zero utility! Not Max! • Or  $(x^*, y^*, \lambda_0^*, \lambda_1^*) = (x^*, I - x^*, 1, 0)$ that solves max f(x,y) = xy = x(I-x)s.t. x + y - I = 0x > 0, y > 0( - - )

• I.e. 
$$(x^*, y^*) = \left(\frac{1}{2}, \frac{1}{2}\right)$$

Implicit Function Theorem (Th'm 15.7)  
• 
$$F_1, \dots, F_m : \mathbf{R}^{\mathbf{m}+\mathbf{n}} \to \mathbf{R}^1$$
 are  $C^l$  functions  
•  $\text{Let}(\vec{y}^*, \vec{x}^*) = (y_1, \dots, y_m, x_1, \dots, x_n)$  solve  

$$\begin{cases}
F_1(y_1, \dots, y_m, x_1, \dots, x_n) = & 0 \\
\vdots & \vdots \\
F_m(y_1, \dots, y_m, x_1, \dots, x_n) = & 0
\end{cases}$$
If  $\det \begin{pmatrix}
\frac{\partial F_1}{\partial y_1}(\vec{y}^*, \vec{x}^*) & \dots & \frac{\partial F_1}{\partial y_m}(\vec{y}^*, \vec{x}^*) \\
\vdots & \ddots & \vdots \\
\frac{\partial F_m}{\partial y_1}(\vec{y}^*, \vec{x}^*) & \dots & \frac{\partial F_m}{\partial y_m}(\vec{y}^*, \vec{x}^*)
\end{pmatrix} \neq 0$ 

### Implicit Function Theorem (Th'm 15.7) Then, exists $C^1$ functions $y_1 = f_1(x_1, \dots, x_n)$

$$y_m = f_m(x_1, \cdots, x_n)$$
  
defined on a ball  $B$  around  $\vec{x}^*$  such that  
$$F_1(f_1(\vec{x}), \cdots, f_m(\vec{x}), \vec{x}) = 0, \quad y_1^* = f_1(\vec{x}^*)$$
  
 $\vdots \qquad \vdots$   
$$F_m(f_1(\vec{x}), \cdots, f_m(\vec{x}), \vec{x}) = 0, \quad y_m^* = f_m(\vec{x}^*)$$
  
For all  $\vec{x} = (x_1, \cdots, x_n) \in B$ 

# Implicit Function Theorem (Th'm 15.7)

• And 
$$\frac{\partial f_k}{\partial x_h}(\vec{y}^*, \vec{x}^*) = \frac{\partial y_k}{\partial x_h}(\vec{y}^*, \vec{x}^*)$$
 is computed

by



# Theorem 19.12 (Alternative CQ's)

- Suppose  $f, g_1, \ldots, g_k$  are  $C^1$  functions on  $\mathbf{R}^n$
- Let  $\vec{x}^* = (x_1^*, \cdots, x_n^*)$  be a local maximizer of  $\max \left\{ f(x_1, \cdots, x_n) \middle| g_1(x_1, \cdots, x_n) \le b_1, \\ \cdots, g_k(x_1, \cdots, x_n) \le b_k \right\}$
- Notation: Constraints  $g_1, \ldots, g_{k_0}$  binds
  - $g_1(x_1^*, \cdots, x_n^*) = b_1, \cdots, g_{k_0}(x_1^*, \cdots, x_n^*) = b_{k_0}$
- Constraints  $g_{k_0+1}, \ldots, g_k$  do not binds

$$g_{k_0+1}(x_1^*, \cdots, x_n^*) < b_{k_0+1}, \cdots, g_k(x_1^*, \cdots, x_n^*) < b_k$$

# Theorem 19.12 (Alternative CQ's)

• If constraints  $g_1, \ldots, g_{k_0}$  satisfies one of these: a) NDCQ: Jacobian matrix has maximum rank  $k_0$ 



• Or, row vectors

$$Dg_i = \nabla g_i = \left(\frac{\partial g_i}{\partial x_1}(\vec{x}^*), \cdots, \frac{\partial g_i}{\partial x_n}(\vec{x}^*)\right)$$

are linearly independent

# Theorem 19.12 (Alternative CQ's)

- b) Karush-Kuhn-Tucker CQ: For  $\vec{v} \in \mathbf{R}^n$  such that  $Dg_i(\vec{x}^*)(\vec{v}) \leq 0$ , Exists  $\epsilon > 0$  and  $\alpha : [0, \epsilon) \rightarrow \mathbf{R}^n$ 1.  $\alpha(0) = \vec{x}^*$  (*C*<sup>1</sup> curve) 2.  $\alpha'(0) = \vec{v}$ 
  - 3.  $g_i(\alpha(t)) \le b_i, t \in [0, \epsilon), i = 1, \cdots, k$
- c) Slater CQ: Exists a ball U about  $\vec{x}^*$  in  $\mathbb{R}^n$  such that  $g_1, \ldots, g_{k_0}$  are convex functions on U and there exists  $\vec{z} \in U$  so that each  $g_i(\vec{z}) < b_i$
- d) Constraints  $g_1, \ldots, g_{k_0}$  are concave functions e) Constraints  $g_1, \ldots, g_{k_0}$  are linear functions

# **Definition:** Concave/Convex Functions

- A set U is a convex set if any line segment between two points in the set is also in the set.  $\vec{x}, \vec{y} \in U \Rightarrow l(\vec{x}, \vec{y}) = \{t\vec{x} + (1-t)\vec{y} | 0 \le t \le 1\}$
- f is a concave function on U if  $\forall \vec{x}, \vec{y} \in U, t \in [0, 1]$  $f(t\vec{x} + (1-t)\vec{y}) \ge tf(\vec{x}) + (1-t)f(\vec{y})$
- g is a convex function on U if  $\forall \vec{x}, \vec{y} \in U, t \in [0, 1]$  $g(t\vec{x} + (1-t)\vec{y}) \leq tg(\vec{x}) + (1-t)g(\vec{y})$
- A linear function h is both concave and convex