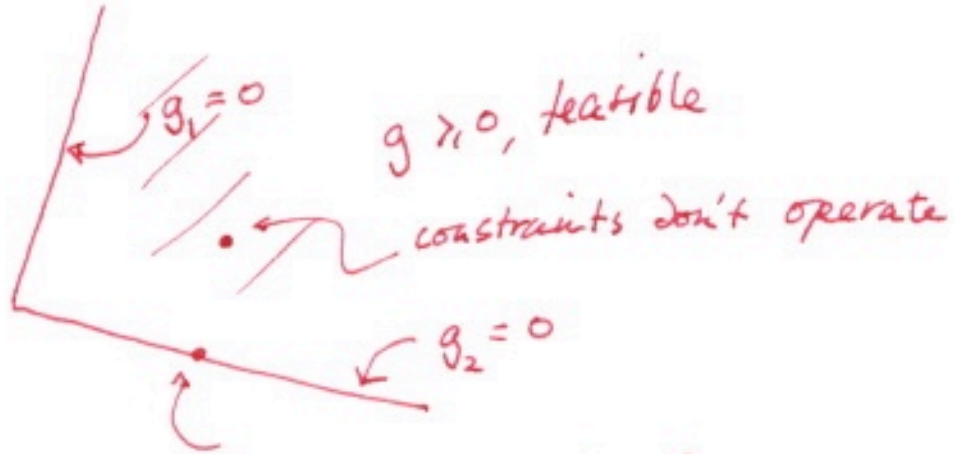


# Lagrangians and inequalities

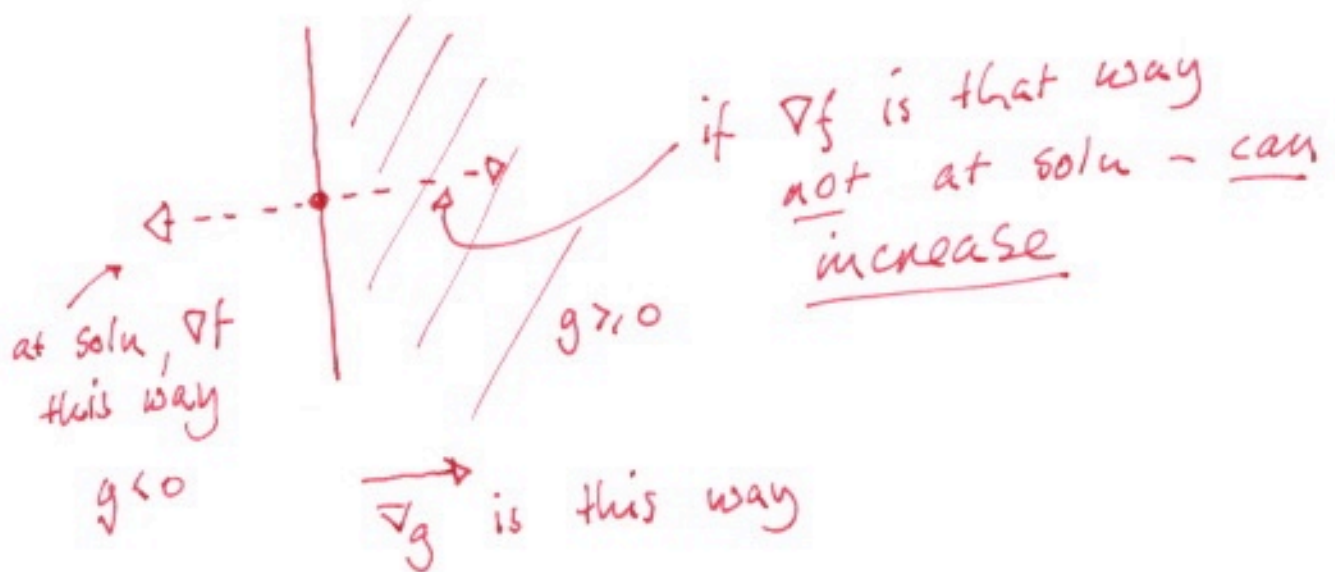
①

[use !p. to visualize]

$$\begin{aligned} \max f(x) \\ \text{st } g(x) \geq 0 \end{aligned}$$



can move along boundary or inside, but not across



all this means that a soln. of (2)

$$\begin{array}{ll} \max & f(x) \\ \text{st} & g(x) \geq 0 \end{array}$$

$$\nabla f + \lambda^T J_g = 0$$

Jacobian; matrix of derivatives.

AND  $\lambda \geq 0$

AND  $g \geq 0$

AND if  $g_i = 0$ ,  $\lambda_i > 0$   
 $g_i > 0$ ,  $\lambda_i = 0$

Sidenote: a great deal of ~~infinite~~ confusion can result caused by different conventions.

various signs will change if:

- min  $f$  instead of max  $f$
- $g \leq 0$  instead of  $g \geq 0$

you are likely to see all possibilities. BUT the signs are there for a reason. You can work it all out as above.

Now bundle in equality constraints

$$\begin{aligned} \max f \quad & \text{st} \quad g \geq 0 \\ & h = 0 \end{aligned}$$

$$L: f - \lambda^T g - \mu^T h$$

Conditions for a solution :

$$\begin{aligned} \nabla_x L &= 0 \\ g &\geq 0 \\ h &= 0 \\ h_i g_i &= 0, \text{ for each } i \\ \lambda &\geq 0 \end{aligned}$$

↖ Karush-Kuhn-Tucker or KKT conditions

Notice that

$$\lambda_i g_i = 0 \quad \text{is } \equiv \text{ to } \begin{cases} g_i = 0, \lambda \geq 0 \\ g_i > 0, \lambda = 0 \end{cases}$$

(4)

this is known as a complementarity condition

and is **VICIOUSLY NON LINEAR**  
and a source of much suffering

We can still construct duals.

$$Q(\lambda, \mu) = \sup_x J(x, \lambda, \mu)$$

(notice: if you want to min, mt)

$$Q(\lambda, \mu) \geq f(x^*)$$

(by previous argument)

Under fairly weak conditions  
(eg: convex problem, feasible set has interior point) (5)

Strong duality holds.

$$\hookrightarrow \inf_{\lambda, \mu} Q(\lambda, \mu) = f(x^*).$$

This has important applications, which we will explore.

Constructing duals:

Linear program

$$\begin{array}{ll} \max & c^T x \\ \text{st.} & Ax = b \\ & x \geq 0 \end{array}$$

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$$L(x, \lambda, \mu) = c^T x - \lambda^T x - \mu^T (Ax - b)$$

$$Q(\lambda, \mu) = \sup_x L(x, \lambda, \mu)$$

notice that this =  $\infty$  unless the  
coeff of  $x$  in  $L$  is 0

so we must have

$$c - A^T \mu - \lambda = 0$$

for the dual to be finite

IN THIS SET,  $L(x, \lambda, \mu) = \mu^T b$

AND  $\lambda \geq 0$   $\leftarrow$  from KKT

so dual is

$$\min \mu^T b$$

$$\text{st } c - A^T \mu - \lambda = 0$$

$$\lambda \geq 0$$

# KKT for linear program

(7)

$$\nabla_x L = 0 = c - A^T \mu - \lambda$$

$$x \geq 0$$

$$Ax - b = 0$$

$$\lambda \geq 0$$

[complementarity]

these are primal feasibility conditions

these are dual feasible conds.

so we have

$$(x, \lambda, \mu)$$

is

- (a) primal feas
- (b) dual feas
- (c) complementarity

|||  
 $(x, \lambda, \mu)$

satisfy KKT

|||  
 $(x, \lambda, \mu)$

are soln

This opens a new world of algorithms 8

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Was: Simplex work in primal.

New

- work with primal feasible  $x$   
dual infeasible  $\lambda, \mu$   
complementarity; improve  $\lambda, \mu$ .
- work with primal infeasible  $x$   
dual feasible  $\lambda, \mu$   
complementarity; ~~fix~~ improve  $x$
- work w/ feasible  $x$   
dual feasible  $\lambda, \mu$   
not comp.; improve comp