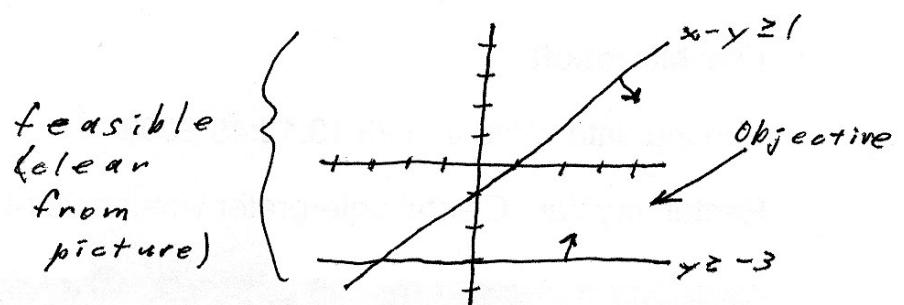


Recitation 1/31/08 : LP Duality & QPs ①

Recall definition of an LP:

linear objective with linear
inequality (+equality) constraints

$$\text{e.g. } \left\{ \begin{array}{l} \min_{x,y} 3x+y \\ \text{s.t. } x-y \geq 1 \\ \cancel{y \geq -3} \end{array} \right\} \rightarrow \left\{ \begin{array}{l} \min_{x,y} 3x+y \\ \text{s.t. } x-y-1 \geq 0 \\ y+3 \geq 0 \end{array} \right\}$$



Now, change to dual:

Take general linear combination of constraints:

$$\begin{aligned} & \alpha(x-y-1) \geq 0 \\ & b(y+3) \geq 0 \\ \hline & \alpha(x-y-1) + b(y+3) \geq 0 \end{aligned} \quad (\text{where } \alpha, b \geq 0)$$

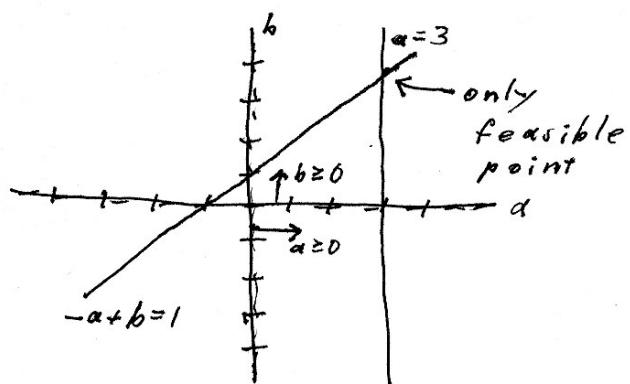
$$\alpha x + (-\alpha + b)y \geq \alpha - 3b$$

→ If $\alpha = 3$ and $-\alpha + b = 1$, then

$$\alpha x + (-\alpha + b)y = \underbrace{3x+y}_{\text{primal objective}} \geq \underbrace{\alpha - 3b}_{\text{lower bound}}$$

So dual problem is to maximize this lower bound.

$$\left\{ \begin{array}{l} \max_{\alpha, b} \alpha - 3b \\ \text{s.t. } \alpha = 3 \\ -\alpha + b = 1 \\ \alpha, b \geq 0 \end{array} \right.$$



(2)

Now, let's go back to primal
by taking the dual of the dual:

Take general linear combination of dual constraints

Note: Change dual objective to $\min -\alpha + 3b$
instead of $\max \alpha - 3b$ b/c
the method we're using to get dual
relies on minimization problem.

$$x(\alpha - 3) + y(-\alpha + b - 1) + z\alpha + wb \geq 0, \quad x, y \in \mathbb{R}$$

$z, w \geq 0$

Note: (Related to above note, we have
 $x, y \in \mathbb{R}$ since equality constraints
and $z, w \geq 0$ since inequality constraints.
If we were deriving dual of a
maximization problem, we would
require $z, w \leq 0$.

$$(\alpha - y + z)\alpha + (y + w)b \geq 3x + y$$

If $\alpha - y + z = 1$ and $y + w = +3$, then

$$(\alpha - y + z)\alpha + (y + w)b = -\alpha + 3b \geq \underbrace{3x + y}_{\substack{\text{dual} \\ \text{objective}}} \geq \underbrace{3x + y}_{\text{lower bound}}$$

Also, we can eliminate w, z :

$$\begin{aligned} x - y + z &= 1 & y + w &= 3 \\ y &= -1 - x + y \geq 0 & w &= 3 - y \geq 0 \\ -x + y &= 1 & -y &\geq -3 \end{aligned}$$

So dual of dual is:

$$\max_{x, y} 3x + y \text{ s.t. } -x + y \geq 1, -y \geq -3$$

Set $x = -x$ and $y = -y$ to get

$$\min_{x, y} 3x + y \text{ s.t. } x - y \geq 1, y \geq 3$$

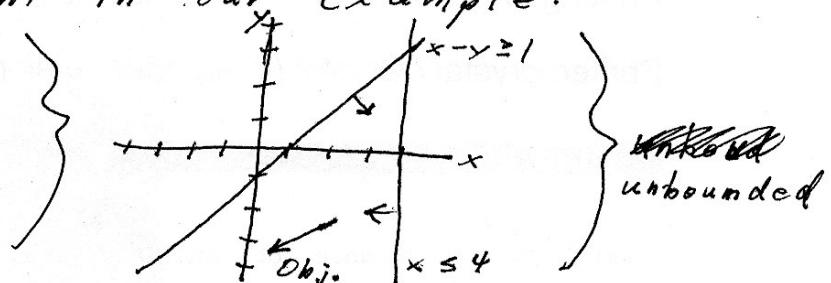
This is our original (primal) problem! (3)

- * For LPs, the dual of the dual is the primal. *
- ↳ b/c strong duality holds (more later...)
- For LPs,
- ① If primal is feasible + finite, dual is too.
 - ② If primal is unbounded, dual is infeasible.
 - ③ If dual " ", primal " " .

To illustrate ②:

Change 1 constraint in our example:

$$\begin{array}{ll} \min_{x,y} & 3x + y \\ \text{s.t.} & x - y \geq 1 \\ & x \leq 4 \end{array}$$



Look at dual:

$$\alpha(x-y-1) + b(-x+4) \geq 0$$

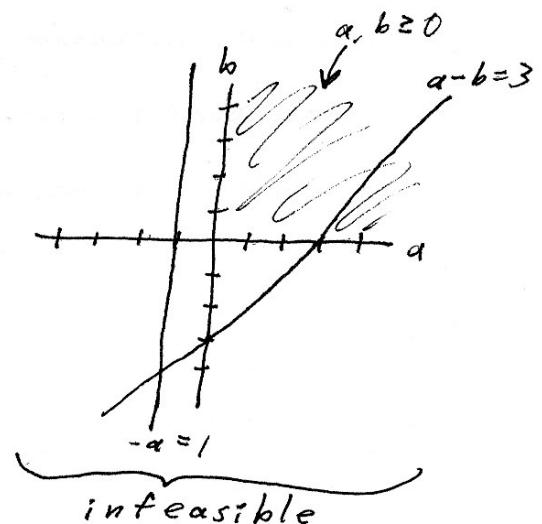
$$(\alpha-b)x - \alpha y \geq \alpha - 4b$$

$$\text{dual} = \max_{\alpha, b} \alpha - 4b$$

$$\text{s.t. } \alpha - b = 3$$

$$-\alpha = 1$$

$$\alpha, b \geq 0$$



(4)

LP Duality = Lagrangians

For LPs, this is an equivalent way of looking at duality.

Say problem is:

$$\begin{cases} \min_{x,y} & 3x+y \\ \text{s.t.} & x-y \geq 1 \\ & y \geq -3 \end{cases}$$

Then write out Lagrangian as:

$$L(\alpha, b, x, y) =$$

$$[3x+y] - [\underbrace{\alpha(x-y-1)}_{\substack{\text{objective} \\ \text{of primal} \\ \text{problem}}} + \underbrace{b(y+3)}_{\text{constraints}}]$$

(same linear combination of " $\dots \geq 0$ " as in previous explanation of duality)

[Note: α, b are called Lagrange multipliers.]

We now want to solve:

$$\min_{x,y} \max_{\alpha, b \geq 0} L(\alpha, b, x, y)$$

↑ ↙
minimize enforce
objective constraints

To understand this, think of this as a game:
 MIN player wants to choose x, y to minimize $L()$, and
 MAX player wants to choose $\alpha, b \geq 0$ to maximize $L()$.

If multiplier $\alpha = 0$, then constraint is not binding.

Suppose MIN chooses $y < -3$ (so constraint is violated). Then MAX can choose b to make $-b(y+3)$ arbitrarily large; so, MIN can't violate constraint b/c multiplier b is enforcing it.

This is complementary slackness: if Lagrange multiplier is > 0 , then the corresponding constraint is binding.

(5)

To derive dual from Lagrangian, write

$$L(a, b, x, y) = \underbrace{(3-a)x + (1+a-b)y + a - 3b}_{\text{dual constraints}} + \underbrace{a}_{\text{dual objective}}$$

$$\left. \begin{array}{l} \frac{\partial}{\partial x} L(a, b, x, y) = 3-a = 0 \\ \frac{\partial}{\partial y} L(a, b, x, y) = 1+a-b = 0 \end{array} \right\} \begin{array}{l} \text{Set } = 0 \text{ since there} \\ \text{must be no incentive} \\ \text{for MIN player} \\ \text{to deviate.} \end{array}$$

*same dual constraints
as before*

QPs

(6)

QP definition: quadratic objective,
linear inequality + equality constraints

$$\min_x \frac{1}{2} x^T H x + c^T x$$

$$\text{s.t. } Ax + b \geq 0$$

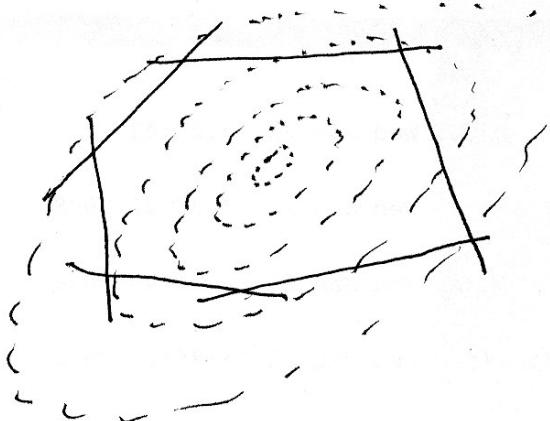
$$Ex + d = 0$$

Convex QP definition: same as QP except
positive semidefinite objective
(H is pos. semidef.)

QPs are hard (NP-complete), but

convex QPs are not (about same as LPs).

↳ Geometric intuition:



If ~~general~~ QP, then
optimal solution
might lie at any
corner of polytope;
e.g. say objective is
largest at center of
ellipses. There could
be a lot of corners
to check....

But if convex QP, solution will be
in interior, or if solution is at
boundary of feasible region, then
duality will make problem easy
to solve.