

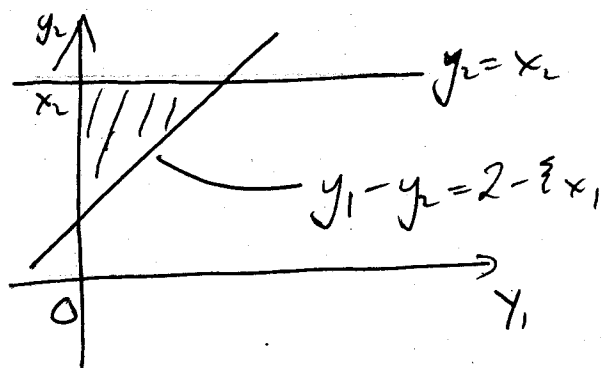
## Birge & Louvarov, Page 101, Question 2

Need to satisfy  $y_1 - y_2 \leq 2 - \xi x_1$        $\xi$  is restricted to  
 $y_2 \leq x_2$       be nonnegative.  
 $y_1, y_2 \geq 0$ .

Note that if  $x_2 < 0$  then no feasible  $y$  exists.

② If  $x_1 \leq 0$  and  $x_2 \geq 0$  then  $y_1 = 0, y_2 = 0$  is feasible.

If  $x_1 \geq 0, x_2 \geq 0$ :



In both distributions,  $\xi \geq 0$ .

As  $\xi$  and/or  $x_1$  increases, the diagonal line moves to the left.

So the problem is feasible if  $y_1 = 0, y_2 = x_2$  is feasible,

that is, if  $-x_2 \leq 2 - \xi x_1$ .

So  $K_2(\xi) = \{x : \xi x_1 - x_2 \leq 2, x_2 \geq 0\}$

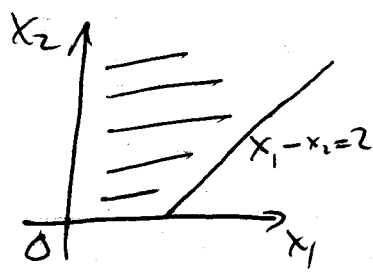
page 101, Q2

(a)  $\xi \in \mathcal{U}[0, 1]$ .

$$K_2 = \bigcap_{\xi \in [0, 1]} K_2(\xi)$$

$$= \bigcap_{\xi \in [0, 1]} \left\{ x : x_1 \leq \frac{2+x_2}{\xi}, x_2 \geq 0 \right\}$$

$$= \underline{\underline{\left\{ x : x_1 - x_2 \leq 2, x_2 \geq 0 \right\}}}$$



(b)  $\xi \sim \text{Poisson}(\lambda), \lambda > 0$

$$K_2 = \bigcap_{\xi} \left\{ x : x_1 \leq \frac{2+x_2}{\xi}, x_2 \geq 0 \right\}$$

$$= \underline{\underline{\left\{ x : x_1 \leq 0, x_2 \geq 0 \right\}}}$$

since  $K_2(\xi)$  restricts  $x_1$  more and more as  $\xi$  grows.

In both cases,  $\Omega$  is fixed and  $\xi$  has finite second moments, so  $K_2 = K_2^*$  and  $K_2$  is closed and convex.

In (a), the support  $\Xi$  is ~~finite~~ polyhedral, so  $K_2$  is polyhedral.

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$$Q(x) = \int_1^{\infty} Q(x, \xi) f(\xi) d\xi$$

$$Q(x, \xi) = \max\{\xi, x\}, \text{ but for all } \xi.$$

$$\text{So } \mathcal{K}_2^p = \{x: x \geq 0\}.$$

$$Q(x) = \int_1^x x f(\xi) d\xi + \int_x^{\infty} \xi f(\xi) d\xi \quad \text{if } x \geq 1$$

$$= \int_1^x \frac{2x}{\xi^3} d\xi + \int_x^{\infty} \frac{2}{\xi^2} d\xi$$

$$= \left[-\frac{x}{\xi^2}\right]_1^x - \left[\frac{2}{\xi}\right]_x^{\infty}$$

$$= x - \frac{1}{x} + \frac{2}{x} = x + \frac{1}{x} \quad \text{if } x \geq 1.$$

if  $x < 1$ :

$$Q(x) = \int_1^{\infty} \xi f(\xi) d\xi = \int_1^{\infty} \frac{2}{\xi^2} d\xi = \left[-\frac{2}{\xi}\right]_1^{\infty}$$

$$= 2.$$

So  $Q(x) \cap$  but for all  $x$ , so  $\mathcal{K}_2 = \mathcal{K}_2^p = \{x: x \geq 0\}$ .

$\xi$  does not have bounded second moment:

$$\int_1^{\infty} \xi^2 f(\xi) d\xi = \int_1^{\infty} \frac{2}{\xi} d\xi, \text{ diverges.}$$

(a) Write constraints as equality constraints:

$$\begin{aligned} \text{min } & 2y_1 + y_2 \\ \text{st. } & y_1 + y_2 - y_3 = 1 - x_1 \\ & y_1 - y_4 = \frac{1}{2} - x_1 - x_2 \\ & y_1, y_2, y_3, y_4 \geq 0. \end{aligned}$$

$$\text{So } W = \begin{bmatrix} 1 & 1 & -1 & 0 \\ 1 & 0 & 0 & -1 \end{bmatrix}$$

To show complete recourse, need to verify that  $Wy = b, y \geq 0$  has a solution for every  $b$ .

$$\text{Let } y_1 = \max\{0, b_1, b_2\}, y_2 = 0, y_3 = y_1 - b_1, y_4 = y_1 - b_2.$$

This satisfies  $Wy = b, y \geq 0$ .

So the problem has complete recourse.

Note that the information about  $\epsilon(\frac{1}{2})$  is irrelevant.

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(b)  $\max z_1, z_2$   
 s.t.  $y_1, y_2 \geq 1 - x_1$   
 $y_1 \geq \xi - x_1 - x_2$   
 $y_1, y_2 \geq 0$  (P)

Dual is:  $\max (1 - x_1)z_1 + (\xi - x_1 - x_2)z_2$   
 s.t.  $z_1 + z_2 \leq 2$  (D)  
 $z_1 \leq 1$   
 $z_1, z_2 \geq 0$ .

If  $\xi \geq x_1 + x_2$ : Given soln  $y$  has value  $2(\xi - x_1 - x_2) + (1 - \xi + x_1)^+$   
 If  $(1 - \xi + x_1) \leq 0$ :

Let  $z_1 = 0, z_2 = 2$ .

Then primal & dual feasible, and optimal values are both  $2(\xi - x_1 - x_2)$ .

If  $(1 - \xi + x_1) > 0$ :

Let  $z_1 = 1, z_2 = 1$ .

Then primal & dual feasible, and optimal values are both  $1 + \xi - 2x_1 - x_2$ .

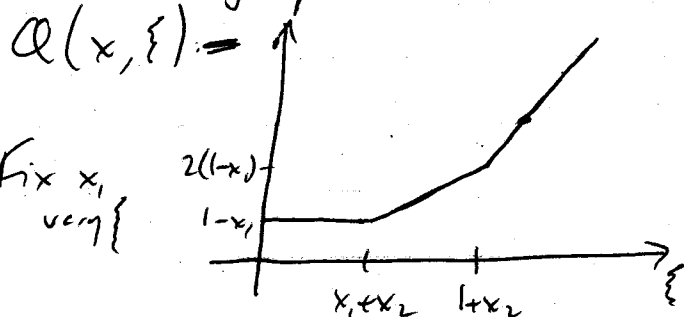
If  $\xi \leq x_1 + x_2$ :

Let  $z_1 = 1, z_2 = 0$

Then primal and dual feasible, and optimal values are both  $1 - x_1$ .

From the above, we immediately get  $Q(x, \xi) \leq$  stated.

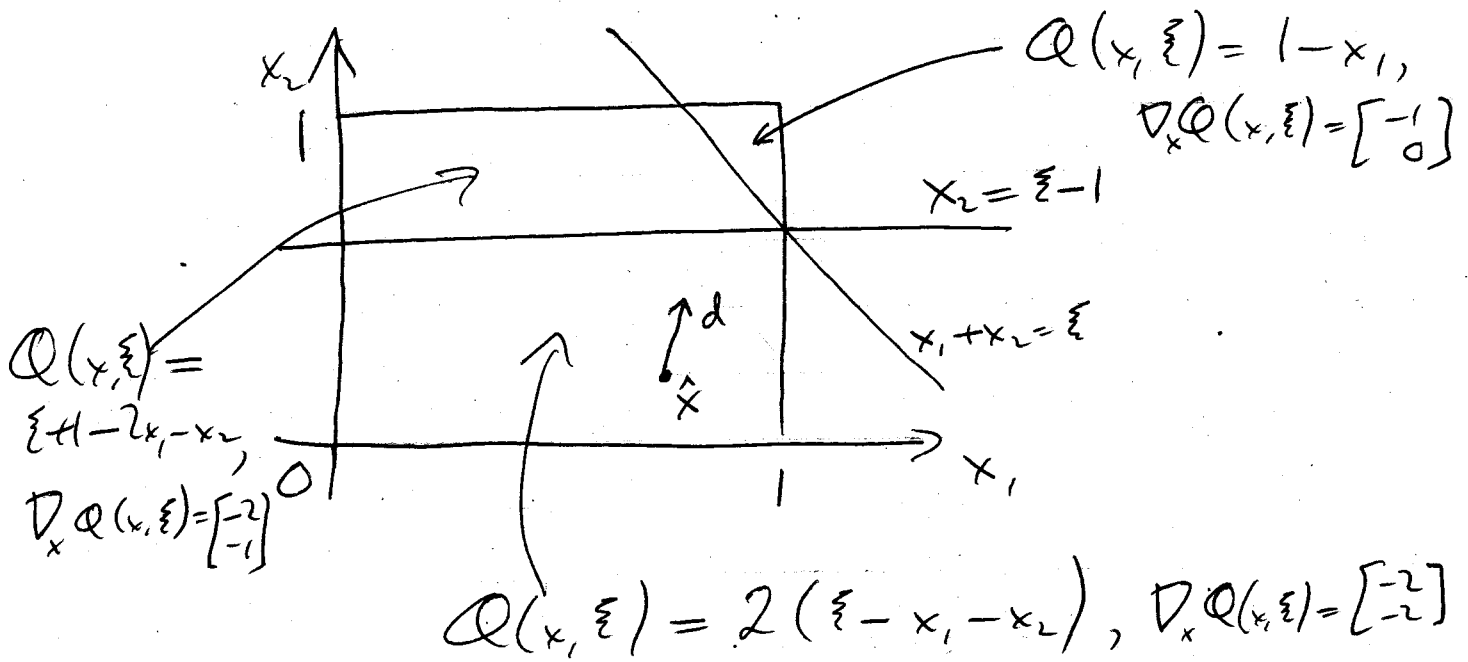
$Q(x, \xi)$  is clearly piecewise linear in either  $\xi$  or  $x_1$  and  $x_2$



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If we fix  $\xi$  and vary  $x_1, x_2$ , show  $Q(x, \xi)$

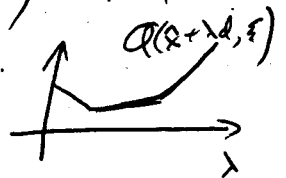
is convex and piecewise linear:



$Q(x, \xi)$  is linear on each segment.

It is continuous across the boundaries of the segments.

For any  $\hat{x}$  and any direction  $d$ , the directional derivative ~~changes as  $\lambda$  increases~~ ~~increases~~ nonotonically in  $\lambda$  for points  $\hat{x} + \lambda d$ . So function is convex.



Eg:  $\xi = 1.1, \hat{x} = (1, 0), d = \text{---}(-2, 5)$ :

$$d^T \nabla_x Q(\hat{x} + \lambda d, \xi) = \begin{cases} -6 & \text{if } 0 \leq \lambda < \frac{1}{50} \\ -1 & \text{if } \frac{1}{50} < \lambda < \frac{1}{30} \\ 2 & \text{if } \frac{1}{30} < \lambda < \frac{1}{5} \end{cases}$$

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(c)  $\xi \sim U[0, 2]$ .

$$Q(x) = \frac{1}{4}(x_1^2 + 2x_2^2 + 2x_1x_2 - 8x_1 - 6x_2 + 9)$$

We have fixed variance and finite second moments for  $\xi$ .

Note that this expression for  $Q(x)$  does not apply

for all  $x$ . It is valid within  $0 \leq x_1 \leq 1, 0 \leq x_2 \leq 1$ .

As written,  $Q(x)$  is not Lipschitzian everywhere.

If  $x_1 \geq 1$  and  $x_1 + x_2 \geq 2$ , then  $Q(x, \xi) = 0$  for all  $\xi \in [0, 2]$ ,  
so  $Q(x) = 0$ , for example.

So we only check Thm 6 for  $0 \leq x_1 \leq 1, 0 \leq x_2 \leq 1$ .

$$Q(x) \text{ is convex: } \nabla Q(x) = \begin{bmatrix} \frac{1}{2}x_1 + \frac{1}{2}x_2 - 8 \\ \frac{1}{2}x_1 + x_2 - 6 \end{bmatrix}$$

$$\text{Hessian: } \nabla^2 Q(x) = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 1 \end{bmatrix}$$

$\nabla^2 Q(x)$  is positive definite, so  $Q(x)$  is convex.

$Q(x)$  is differentiable.

Since we're only checking a ~~finite~~<sup>bounded</sup> domain, and since  $Q$  is differentiable, it is Lipschitzian, from the Mean Value Theorem.

BIRGE & LOUVAUX, PAGE 102, QUESTION 5

If  $\xi < 0$ :  $\Delta y = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$  is a ray that drives the objective to  $-\infty$ , so  $Q(x, \xi) = -\infty$

If  $0 \leq \xi \leq 1$ : Optimal  $y$  is  $y_1 = 1 - x_1, y_2 = 0$ .

$$Q(x, \xi) = \xi(1 - x_1)$$

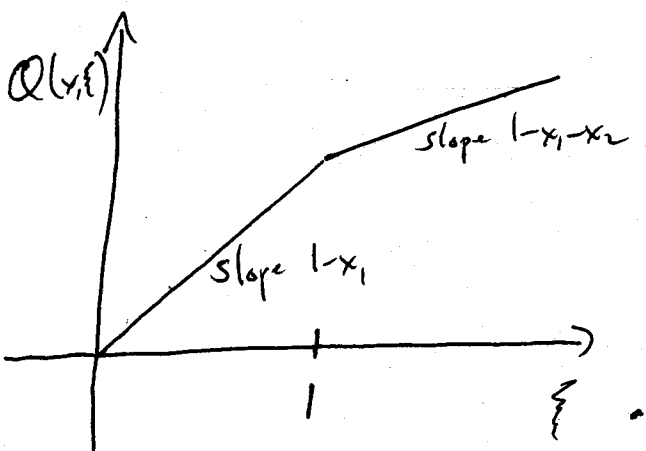
If  $\xi \geq 1$ : (A) If  $x_1 + x_2 \geq 1$ : Optimal  $y$  is  $y_1 = 0, y_2 = 1 - x_1$

$$\text{So } Q(x, \xi) = 1 - x_1$$

(B) If  $x_1 + x_2 \leq 1$ : Optimal  $y$  is  $y_1 = 1 - x_1 - x_2, y_2 = x_2$

$$\text{So } Q(x, \xi) = \xi(1 - x_1) + (1 - \xi)x_2$$

$x_1 + x_2 \leq 1$ :



$x_1 + x_2 \geq 1$ :

