

Integer (linear) programming problem

$$\begin{aligned}
 \text{(IP)} \quad & \min c^T x && (\max) \\
 & \text{s.t. } Ax \leq b && (=, \geq) \\
 & && x \geq 0, x \text{ is integer}
 \end{aligned}$$

$$x, c \in \mathbb{R}^n, b \in \mathbb{R}^m, A \in \mathbb{R}^{m \times n}$$

Special cases

- $x_i \in \{0, 1\}, i = 1, \dots, n$
- mixed IP: 0/1 and integer (continuous and integer)

Feasible region

$$F = \{x : Ax \leq b, x \geq 0, x \text{ is integer}\}$$

Feasible region in LP-relaxation

$$F^{LP} = \{x : Ax \leq b, x \geq 0\}$$

Convex hull of F

$$F^C = \left\{ x : x = \sum_{i=1}^{|F|} \lambda_i x^i, x^i \in F, \sum_{i=1}^{|F|} \lambda_i = 1, \lambda_i \geq 0 \forall i \right\}$$

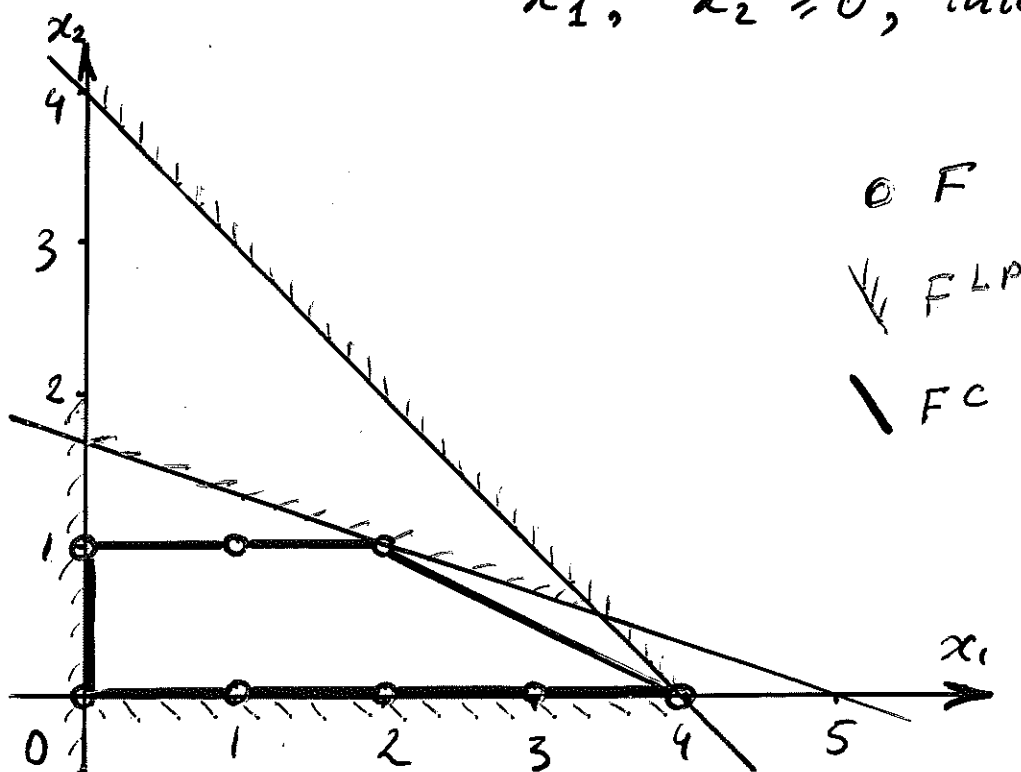
$$F \subset F^C \subset F^{LP}$$

Ex. $\max z = x_1 + 5x_2$

s.t. $x_1 + x_2 \leq 4$

$x_1 + 3x_2 \leq 5$

$x_1, x_2 \geq 0, \text{ integer}$



$F^C = \{x : x_1 + 2x_2 \leq 4, x_2 \leq 1, x_1 \geq 0, x_2 \geq 0\}$

$\max z = x_1 + 5x_2 \text{ s.t. } x \in F \Rightarrow x_{IP}^* = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, z_{IP}^* = 7$

$\max z = x_1 + 5x_2 \text{ s.t. } x \in F^{LP} \Rightarrow x_{LP}^* = \begin{pmatrix} 0 \\ 5/3 \end{pmatrix}, z_{LP}^* = 8\frac{1}{3}$

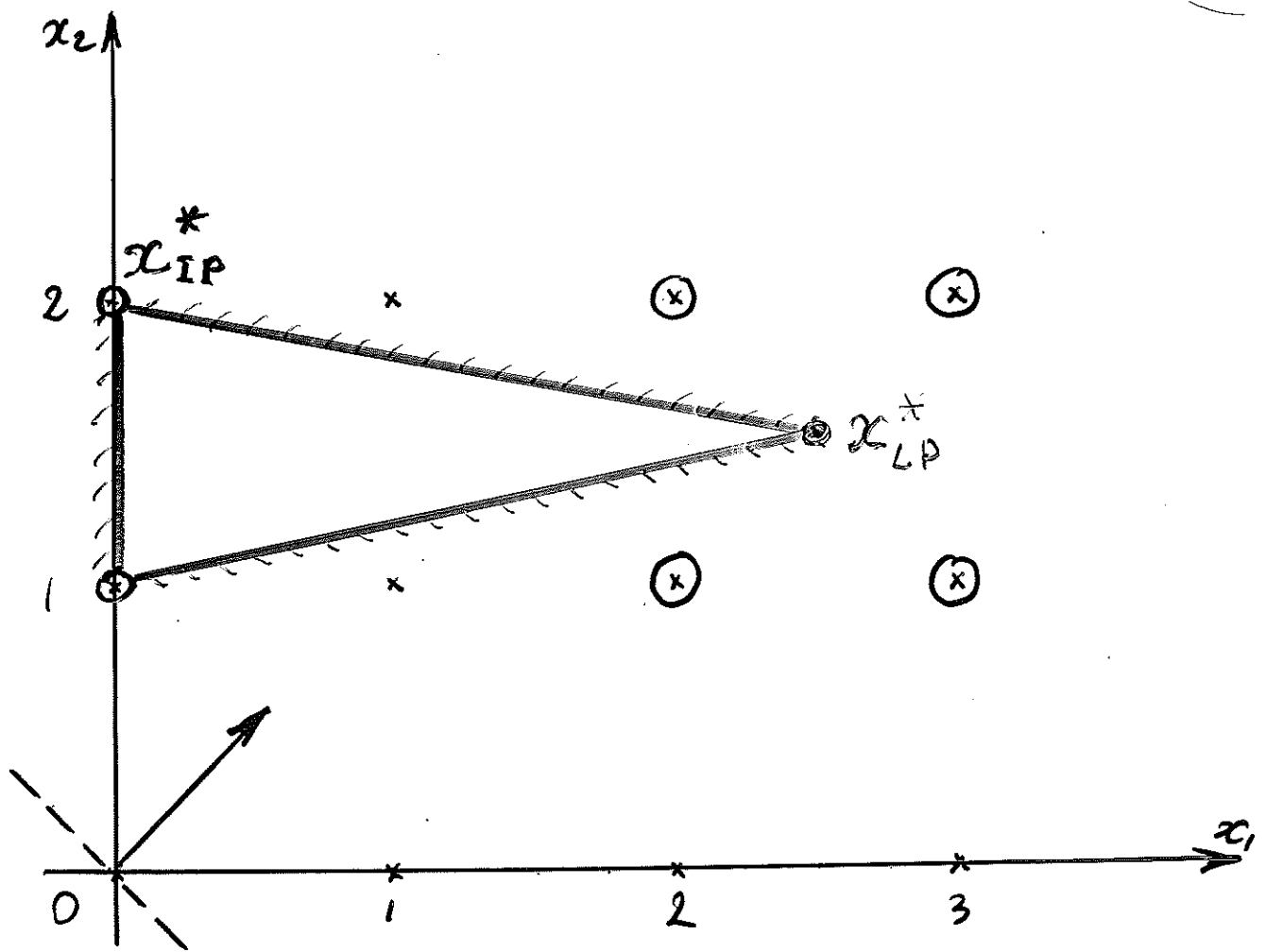
$\max z = x_1 + 5x_2 \text{ s.t. } x \in F^C \Rightarrow x_C^* = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, z_C^* = 7$
 (this is an LP-problem in the λ -variables)

$F \subset F^C \subset F^{LP}$

$\max \Rightarrow z_{IP}^* = z_C^* \leq z_{LP}^*$

$\min \Rightarrow z_{IP}^* = z_C^* \geq z_{LP}^*$

Rounding of x_{LP}^*



⇒ Rounding of x_{LP}^* may give infeasible points

Formulating IP problems

- 1) Given m constraints $a_i^T x \leq b_i$, $i=1, \dots, m$, at least k of which should be satisfied

Introduce:

$$y_i = \begin{cases} 1, & \text{implies that } a_i^T x \leq b_i \text{ holds} \\ 0, & \text{may not imply this} \end{cases}$$

$$\sum y_i \geq k$$

$$a_i^T x \leq b_i + M(1 - y_i), \quad i=1, \dots, m$$

← large enough

Then:

$$y_i = 1 \Rightarrow a_i^T x \leq b_i$$

$$y_i = 0 \Rightarrow a_i^T x \leq b_i + M$$

- 2) Variable $x \in \{\alpha_1, \alpha_2, \alpha_3, \alpha_4\}$.

Introduce:

$$y_1, y_2, y_3, y_4 \in \{0, 1\}$$

$$x = \alpha_1 y_1 + \alpha_2 y_2 + \alpha_3 y_3 + \alpha_4 y_4$$

$$y_1 + y_2 + y_3 + y_4 = 1$$

- 3) Either-or constraints. At least one of the constraints $f(x) \leq 0$ and $g(x) \leq 0$ should be satisfied. Introduce:

$$y = \begin{cases} 0, & f \leq 0 \\ 1, & g \leq 0 \end{cases}$$

$$\begin{aligned} f(x) &\leq M y \\ g(x) &\leq M(1 - y) \end{aligned}$$

← large enough

4) If-then constraints. If $f(x) > 0$, then $g(x) \geq 0$. 11.7

Introduce:

$$y = \begin{cases} 0, & \text{if } f(x) > 0 \\ 1, & \text{if } f(x) \leq 0 \end{cases}$$

$$-g(x) \leq My$$

$$f(x) \leq M(1-y)$$

$$\text{Ex. } \max f(x) = 3x_1 + 5x_2 + 7x_3$$

$$\text{s.t. } g(x) = x_1 + 2x_2 + 3x_3 - 4 \leq 0 \quad \leftarrow u \geq 0$$

$$x \in X = \{0, 1\}^3$$

Integer knapsack problem \Rightarrow not convex

$$x^* = (1, 0, 1), \quad f(x^*) = 10$$

$$\begin{aligned} L(x, u) &= 3x_1 + 5x_2 + 7x_3 \underset{\substack{\uparrow \\ \text{max!}}}{-} u(x_1 + 2x_2 + 3x_3 - 4) = \\ &= 4u + (3-u)x_1 + (5-2u)x_2 + (7-3u)x_3 \end{aligned}$$

$$\begin{aligned} \varphi(u) &= 4u + \max_{x_1=0,1} (3-u)x_1 + \max_{x_2=0,1} (5-2u)x_2 \\ &\quad + \max_{x_3=0,1} (7-3u)x_3 \end{aligned}$$

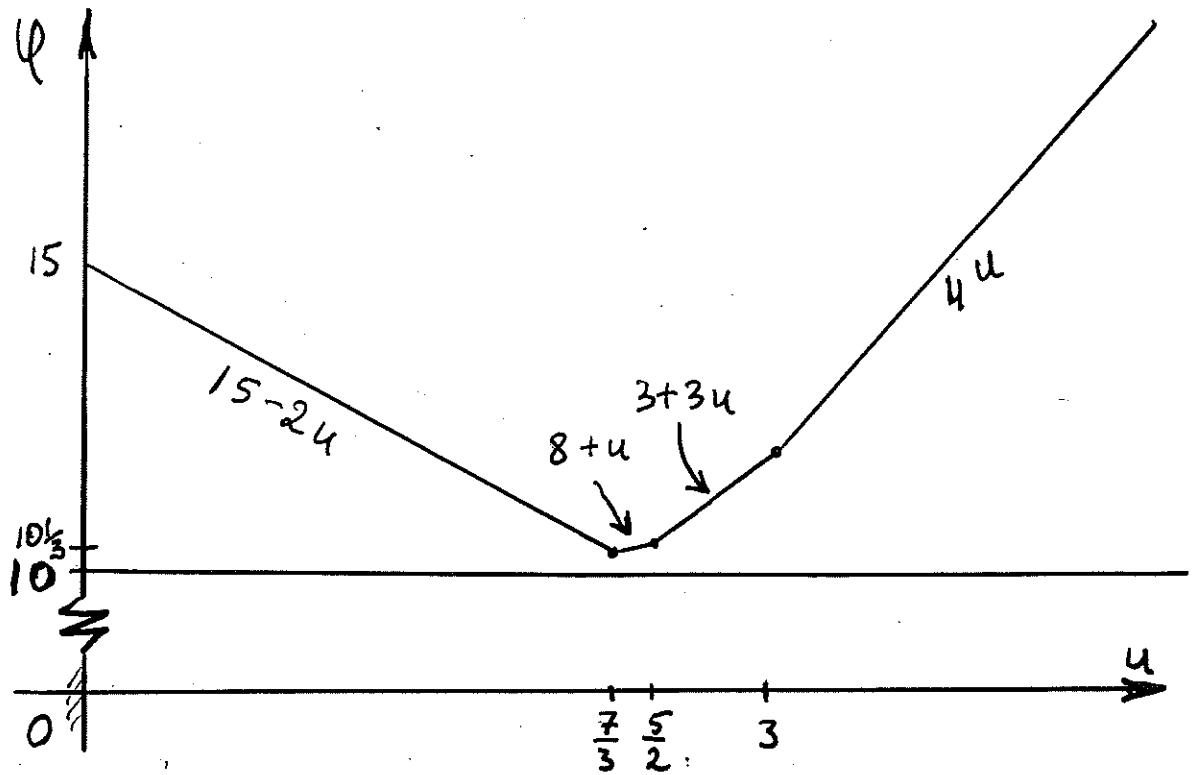
$$\varphi(u) = \begin{cases} 15-2u, & \text{if } 0 \leq u \leq \frac{7}{3} & \leftarrow x(u) = (1, 1, 1) \\ 8+u, & \text{if } \frac{7}{3} \leq u \leq \frac{5}{2} & \leftarrow x(u) = (1, 1, 0) \\ 3+3u, & \text{if } \frac{5}{2} \leq u \leq 3 & \leftarrow x(u) = (1, 0, 0) \\ 4u, & \text{if } 3 \leq u & \leftarrow x(u) = (0, 0, 0) \end{cases}$$

An alternative:

$$\varphi(u) = \max_{x \in \{0,1\}^3} \{4u + (3-u)x_1 + (5-2u)x_2 + (7-3u)x_3\}$$

$$= \max \left\{ \begin{array}{l} 4u, \quad 7+u, \quad 5+2u, \quad 12-u, \quad 3+3u, \\ (0,0,0) \quad (0,0,1) \quad (0,1,0) \quad (0,1,1) \quad (1,0,0) \\ 10, \quad 8+u, \quad 15-2u \\ (1,0,1) \quad (1,1,0) \quad (1,1,1) \end{array} \right\}$$

$$2^3 = 8$$



$$(D) \min_{u \geq 0} \varphi(u) \Rightarrow u^* = \frac{7}{3}, \varphi(u^*) = 10\frac{1}{3}$$

$\varphi(u^*) > f(x^*)$, because (P) is not convex

$$\text{the dual gap} = \varphi(u^*) - f(x^*) = \frac{1}{3} > 0$$

Nondifferentiable Optimization

$$\min_{x \in \mathbb{R}^n} f(x)$$

A. f is convex

If $f \in C^1$ then

$$f(x) \geq f(\bar{x}) + \nabla f(\bar{x})^T (x - \bar{x}), \quad \forall x, \bar{x}$$

Def. A vector $\bar{y} \in \mathbb{R}^n$ such that

$$f(x) \geq f(\bar{x}) + \bar{y}^T (x - \bar{x}), \quad \forall x$$

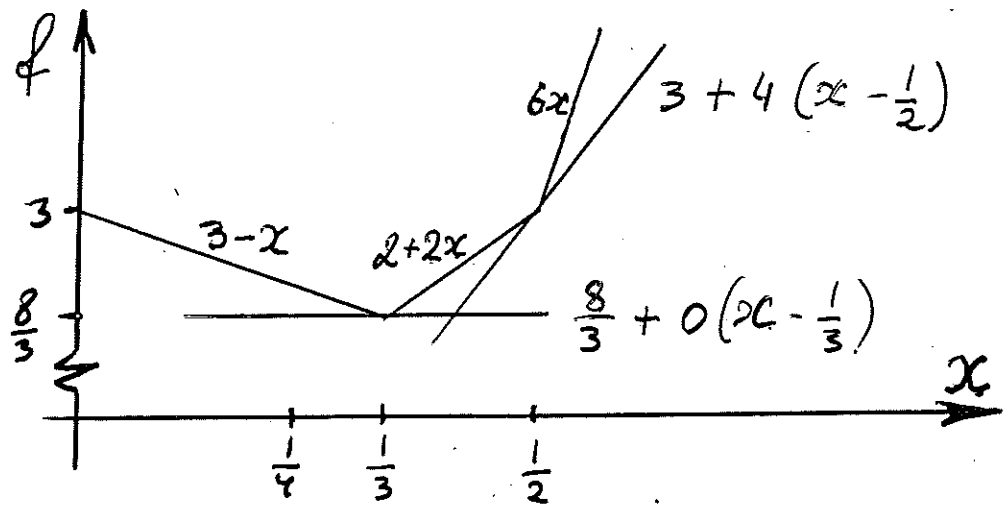
is called subgradient of f at \bar{x} .

Def. The set of subgradients of f at \bar{x} is called subdifferential and is denoted by $\partial f(\bar{x})$.

Rem. If f is concave

$$f(x) \leq f(\bar{x}) + \bar{y}^T (x - \bar{x}) \quad \forall x.$$

$$\text{Ex. } f(x) = \begin{cases} 3-x & \text{if } 0 \leq x \leq \frac{1}{3} \\ 2+2x, & \text{if } \frac{1}{3} \leq x \leq \frac{1}{2} \\ 6x, & \text{if } \frac{1}{2} \leq x \end{cases}$$



$\bar{x} = \frac{1}{2}$, $\bar{y} = 4$ is a subgradient, because
 $f(x) \geq f(\frac{1}{2}) + \bar{y}^T (x - \frac{1}{2}) = 3 + 4(x - \frac{1}{2}), \forall x$

$$\partial f(\frac{1}{2}) = [2, 6]$$

$$\partial f(\frac{1}{3}) = [-1, 2]$$

$$\partial f(\frac{1}{4}) = \{-1\}$$

$$0 \in \partial f(\frac{1}{3}) = [-1, 2]$$

$$\Rightarrow f(x) \geq \frac{8}{3} + 0(x - \frac{1}{3}), \forall x$$

$$\Rightarrow f(x) \geq f(\frac{1}{3}) = \frac{8}{3}, \forall x$$

$$\Rightarrow x^* = \frac{1}{3}$$

Rem.

- f is differentiable in $\bar{x} \Rightarrow \partial f(\bar{x}) = \{\nabla f(\bar{x})\}$
- $\partial f(\bar{x}) = \{\bar{g}\}$ (one element) $\Rightarrow f$ is differentiable in \bar{x} and $\nabla f(\bar{x}) = \bar{g}$

Corollary. f is differentiable in $\bar{x} \Leftrightarrow$
subgradient \bar{g} is unique in \bar{x} ($\bar{g} = \nabla f(\bar{x})$)

Th. x^* is optimal $\Leftrightarrow 0 \in \partial f(x^*)$

Proof. Suppose $0 \in \partial f(x^*) \Rightarrow$

$$f(x) \geq f(x^*) + 0^T(x - x^*), \forall x \Rightarrow$$

$$f(x) \geq f(x^*), \forall x \Rightarrow x^* \text{ is optimal}$$

Suppose x^* is optimal \Rightarrow

$$f(x) \geq f(x^*), \forall x \Rightarrow$$

$$f(x) \geq f(x^*) + 0^T(x - x^*), \forall x \Rightarrow 0 \in \partial f(x^*)$$

Th. If $x(\bar{\lambda})$ solves $\min_{x \in X} f(x) + \bar{\lambda}^T g(x)$

$$\Rightarrow g(x(\bar{\lambda})) \in \partial \psi(\bar{\lambda})$$

(ψ is concave)

Proof. $\psi(\lambda) = \min_{x \in X} f(x) + \lambda^T g(x)$

$$\leq f(x(\bar{\lambda})) + \lambda^T g(x(\bar{\lambda})) =$$

$$= \underbrace{f(x(\bar{\lambda})) + \bar{\lambda}^T g(x(\bar{\lambda}))}_{\psi(\bar{\lambda})} + \lambda^T g(x(\bar{\lambda})) - \underbrace{\bar{\lambda}^T g(x(\bar{\lambda}))}_{\psi(\bar{\lambda})}$$

$$= \psi(\bar{\lambda}) + g^T(x(\bar{\lambda})) \cdot (\lambda - \bar{\lambda}), \quad \forall \lambda$$

$$\Rightarrow g(x(\bar{\lambda})) \in \partial \psi(\bar{\lambda})$$

Rem. $x(\bar{\lambda})$ is unique \Rightarrow

ψ is differentiable in $\bar{\lambda}$ with $\nabla \psi(\bar{\lambda}) = g(x(\bar{\lambda}))$

$\left. \begin{array}{l} X \text{ is convex} \\ f \text{ is strictly convex} \\ g_i \text{ is convex } \forall i \end{array} \right\} \Rightarrow \psi(\lambda) \text{ is differentiable}$

Subgradientoptimering på Lagrange-dual

Primalt problem:

$$\begin{aligned} \min \quad & f(x) \\ \text{då} \quad & g(x) \leq 0 \\ & x \in X. \end{aligned}$$

Lagrange-dualt problem:

$$\max_{u \geq 0} h(u)$$

där

$$h(u) = \min_{x \in X} f(x) + u^T g(x).$$

Subgradientoptimering:

0. Välj $u^0 \geq 0$, sätt $k = 0$ och $LBD_{-1} = -\infty$.
1. Lös Lagrange-relaxationen $\min_{x \in X} f(x) + (u^k)^T g(x)$.
Optimum: $x(u^k) \Rightarrow h(u^k) = f(x(u^k)) + (u^k)^T g(x(u^k))$.
Sätt $LBD_k = \max\{LBD_{k-1}, h(u^k)\}$.
2. Beräkna subgradienten $\gamma^k = g(x(u^k))$ och en steglängd $t_k > 0$.
3. Beräkna nytt iterat som $u^{k+1} = \max\{0, u^k + t_k \gamma^k\}$
(där maximum tas komponentvis).
4. Sätt $k = k + 1$ och gå till steg 1.

Polyak-steglängder:

$$t_k = \lambda_k \frac{UBD_k - h(u^k)}{\|\gamma^k\|_2^2}, \quad k = 0, 1, 2, \dots,$$

där

- värdena på *relaxationsparametern* λ_k väljs så att $0 < \varepsilon_1 \leq \lambda_k \leq 2 - \varepsilon_2 < 2$ gäller för alla k
- $UBD_k \geq h^*$, dvs UBD_k utgör en *optimistisk uppskattning* av h^*
- UBD_k kan vara *iterationsoberoende*, dvs för alla k väljs $UBD_k = UBD$
- vanligen ges UBD_k av den bästa kända primala tillåtna lösningen (som typiskt är heuristiskt genererad)
- relaxationsparametern kan till exempel väljas enligt $\lambda_0 = 2 - \varepsilon_3$ och $\lambda_k = (1 - \varepsilon_4)\lambda_{k-1}$, $k = 1, 2, \dots$, där $\varepsilon_3, \varepsilon_4 \geq 0$ är små.

SATS

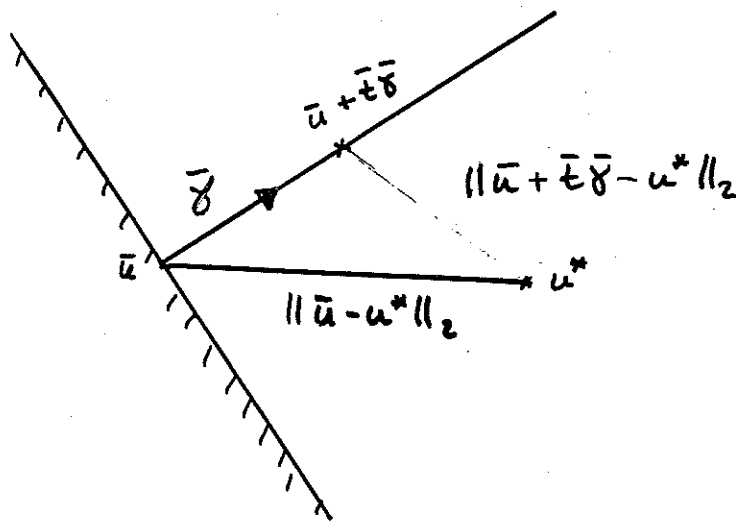
u^* optimal i (D)

$\bar{u} \geq 0$

$\bar{\gamma} \in \partial h(\bar{u})$

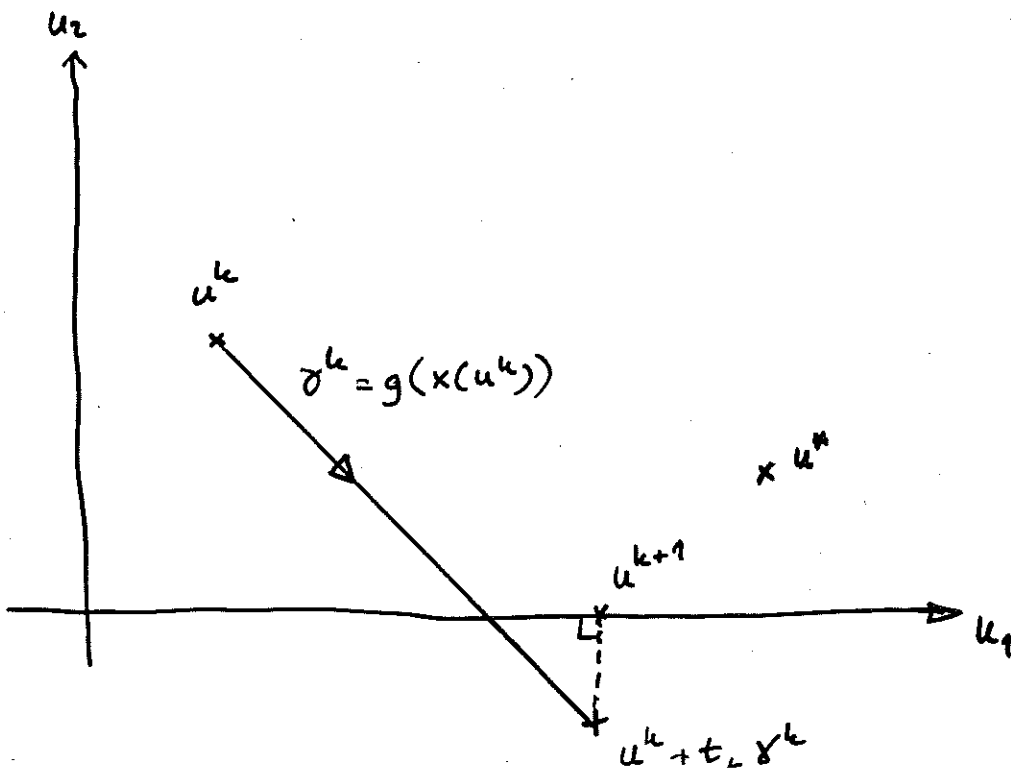
$u^* \in \{u \mid \bar{\gamma}^T(u - \bar{u}) \geq 0\}$

Tolkning: $\bar{\gamma}$ pekar in i ett halvrum där u^* ligger.



\bar{t} = steglängd

Konsekvens: Avståndet till u^* minskar om steget $\bar{t} > 0$ längs $\bar{\gamma}$ är tillräckligt litet.



Ex 1

$$\min f(x) = (x_1 - 7)^2 + (x_2 - 1)^2$$

$$\text{d.h. } g(x) = x_1 + 2x_2 - 4 \leq 0$$

$$x \in \mathbb{R}^2$$

$$x^* = \begin{pmatrix} 6 \\ -1 \end{pmatrix} \quad f^* = 5$$

$$h(u) = -4u + \min_{x_1 \in \mathbb{R}} \{ (x_1 - 7)^2 + ux_1 \} + \min_{x_2 \in \mathbb{R}} \{ (x_2 - 1)^2 + 2ux_2 \}$$

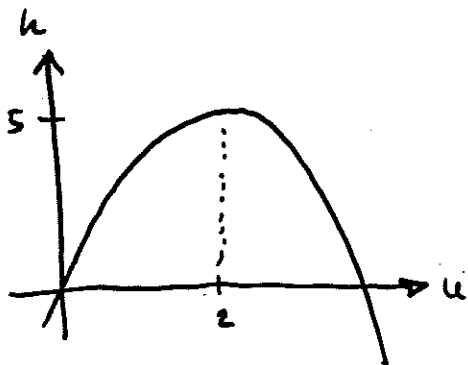
h kann technas explizit.

$$\frac{\partial}{\partial x_1} \{ (x_1 - 7)^2 + ux_1 \} = 2(x_1 - 7) + u = 0 \Rightarrow x_1(u) = 7 - \frac{u}{2}$$

$$\dots \quad x_2(u) = 1 - u$$

$$x(u) = \begin{pmatrix} 7 - \frac{u}{2} \\ 1 - u \end{pmatrix}$$

Insättning ger $h(u) = 5u(1 - \frac{1}{4}u)$



$$\max_{u \geq 0} h(u) \Rightarrow u^* = 2 \quad h^* = 5$$

$$x(u^*) = \begin{pmatrix} 7 - \frac{u^*}{2} \\ 1 - u^* \end{pmatrix} = \begin{pmatrix} 6 \\ -1 \end{pmatrix}$$

$$x^* = x(u^*) = \begin{pmatrix} 6 \\ -1 \end{pmatrix} \text{ l\u00f6ser (P).}$$

$$g(x^*) = x_1^* + 2x_2^* - 4 = 0 \leq 0 \quad \text{OK}$$

$$f(x^*) = (x_1^* - 7)^2 + (x_2^* - 1)^2 = 5 = h^* \quad \text{OK}$$

Ex 2

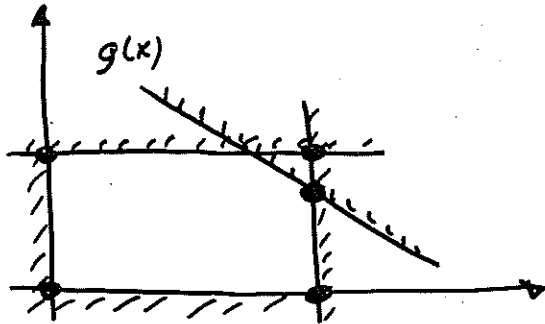
$$\max f(x) = x_1 + x_2$$

$$\text{d} \begin{cases} 2x_1 + 3x_2 \leq 6 & : g(x) \leq 0 \mid u \geq 0 \\ x_1 \leq 2 \\ x_2 \leq 1 \\ x_1, x_2 \geq 0 \end{cases}$$

$$x_1 \leq 2$$

$$x_2 \leq 1$$

$$x_1, x_2 \geq 0$$



$$x^* = \begin{pmatrix} 2 \\ 2/3 \end{pmatrix}$$

$$f^* = 8/3$$

$$L(x, u) = x_1 + x_2 \ominus u (2x_1 + 3x_2 - 6)$$

max problem

$$h(u) = \max_{x \in \mathcal{X}} L(x, u) = \max_{\text{d} \begin{cases} x_1 \leq 2 \\ x_2 \leq 1 \\ x_1, x_2 \geq 0 \end{cases}} 6u + (1-2u)x_1 + (1-3u)x_2$$

$$x_1, x_2 \geq 0$$

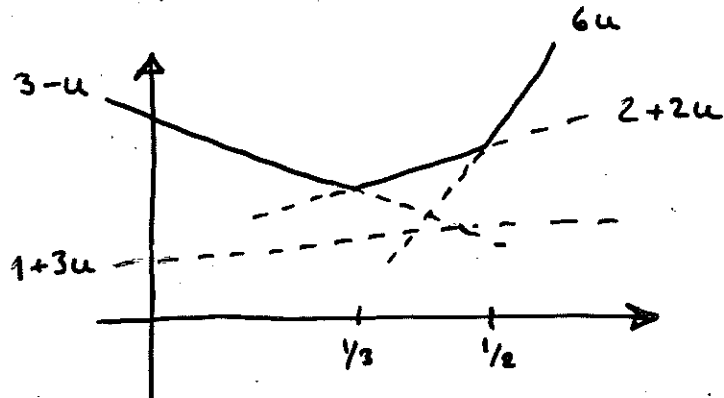
$$(0,0) : 6u$$

$$(2,0) : 2+2u$$

$$(0,1) : 1+3u$$

$$(2,1) : 3-u$$

$$h(u) = \max \{ 6u, 2+2u, 1+3u, 3-u \}$$



$$u^* = \frac{1}{3} \Rightarrow (2,1) \text{ och } (2,0) \text{ optimala} \Rightarrow x_1^* = 2 \quad 0 \leq x_2^* \leq 1$$

$$h^* = \frac{8}{3}$$

$$\bullet \quad u^{*\top} g(x^*) = 0 \Rightarrow \underbrace{\frac{1}{3}}_{>0} (2x_1^* + 3x_2^* - 6) = 0 \Rightarrow$$

$$\Rightarrow 2x_1^* + 3x_2^* = 6 \quad \Rightarrow \quad \begin{matrix} x_2^* = \frac{2}{3} \in [0,1] \\ x_1^* = 2 \end{matrix} \quad \text{ok!}$$

$$\bullet \quad g(x^*) = 0 \leq 0 \quad \text{ok!}$$

$$\therefore x^* = \begin{pmatrix} 2 \\ 2/3 \end{pmatrix} \quad f(x^*) = \frac{8}{3} = u^*$$

Ex 3 $\max f(x) = 3x_1 + 5x_2 + 7x_3$

$$g(x) = x_1 + 2x_2 + 3x_3 - 4 \leq 0 \quad | \quad u \geq 0$$

$$x \in \mathcal{X} = \{0,1\}^3$$

$$x^* = (1,0,1) \quad f^* = 10$$

$$h(u) = 4u + \max_{x_1=0/1} (3-u)x_1 + \max_{x_2=0/1} (5-2u)x_2 + \max_{x_3=0/1} (7-3u)x_3$$

$$h(u) = \begin{cases} 15-2u & 0 \leq u \leq 7/3 & x(u) = (1,1,1) \\ 8+u & 7/3 \leq u \leq 5/2 & x(u) = (1,1,0) \\ 3+3u & 5/2 \leq u \leq 3 & x(u) = (1,0,0) \\ 4u & u \geq 3 & x(u) = (0,0,0) \end{cases}$$

$$\min_{u \geq 0} h(u) \Rightarrow u^* = \frac{7}{3} \quad h^* = 10\frac{1}{3}$$

$$\text{Dualgap} \quad h^* - f^* = 10\frac{1}{3} - 10 = \frac{1}{3}$$

$$\mathcal{I}(u^*) = \{(1,1,1), (1,1,0)\}$$

$$g(1,1,1) = 2 \not\leq 0 \Rightarrow (1,1,1) \text{ tillåten}$$

$$g(1,1,0) = -1 \leq 0 \Rightarrow (1,1,0) \text{ tillåten}$$

$$u^{*\top} g(1,1,0) = \frac{7}{3} \cdot (-1) \neq 0 \quad x^* \text{ kan ej bestämmas så här, pga att } h^* > f.$$