

C&O 355

Lecture 3

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Outline

- Review Local-Search Algorithm
- Pitfall #1: Defining corner points
 - Polyhedra that don't contain a line have corner points
- Pitfall #2: No corner points?
 - Equational form of LPs

Local-Search Algorithm: Pitfalls & Details

Algorithm

Let x be any corner point

For each corner point y that is a neighbor of x

 If $c^T y > c^T x$ then set $x = y$

Halt

Local-Search Algorithm: Pitfalls & Details

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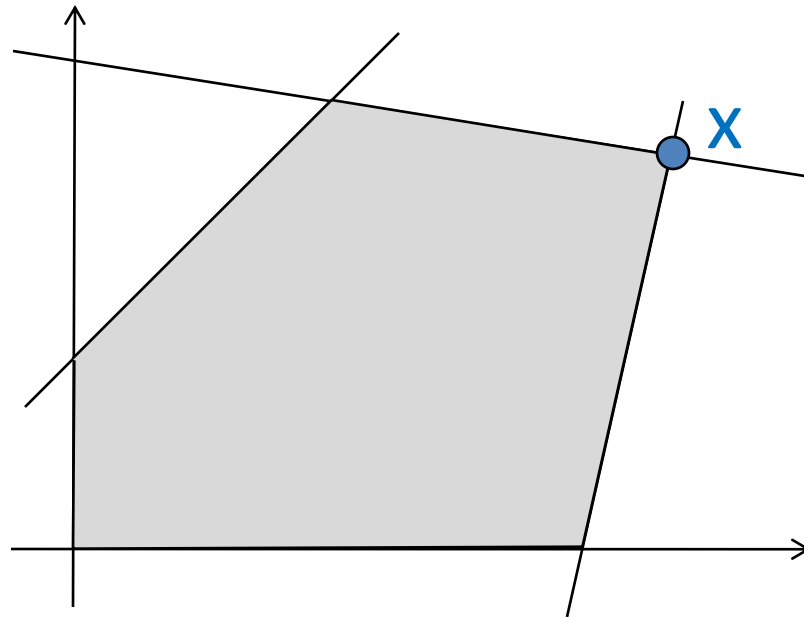
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1. What is a corner point?
2. What if there are no corner points?
3. What are the “neighboring” corner points?
4. How to choose a neighboring point?
5. How can I find a starting corner point?
6. Does the algorithm terminate?
7. Does it produce the right answer?

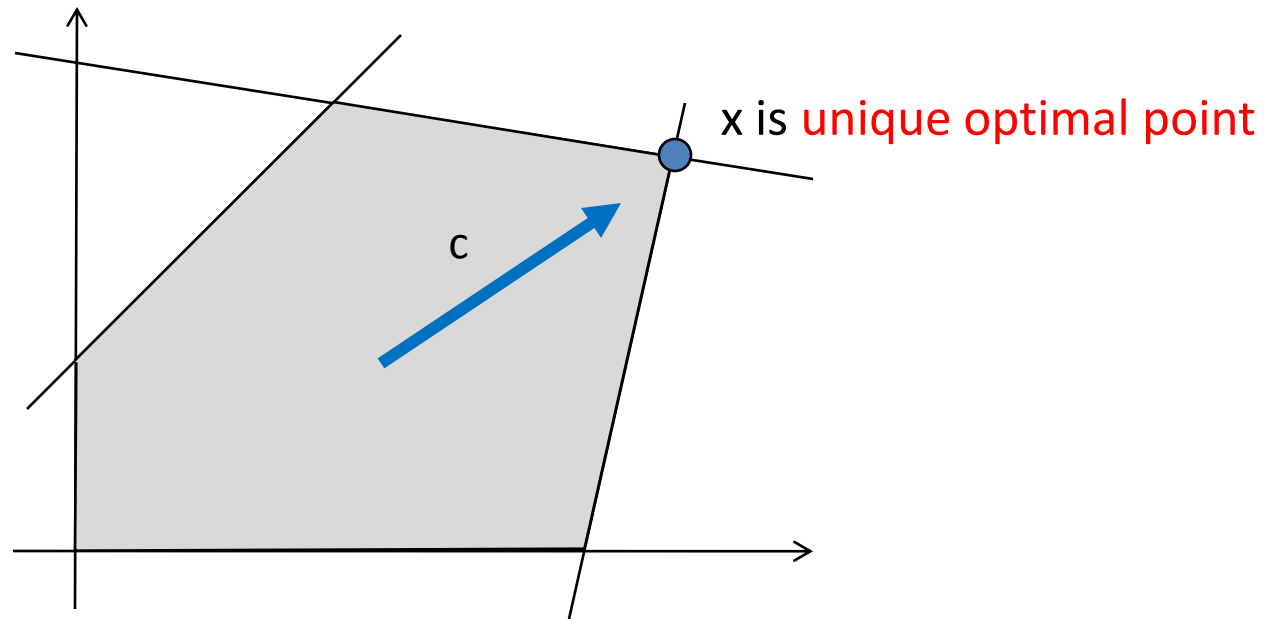
Pitfall #1: What is a corner point?

- How should we define corner points?
- Under any reasonable definition, point x should be considered a corner point



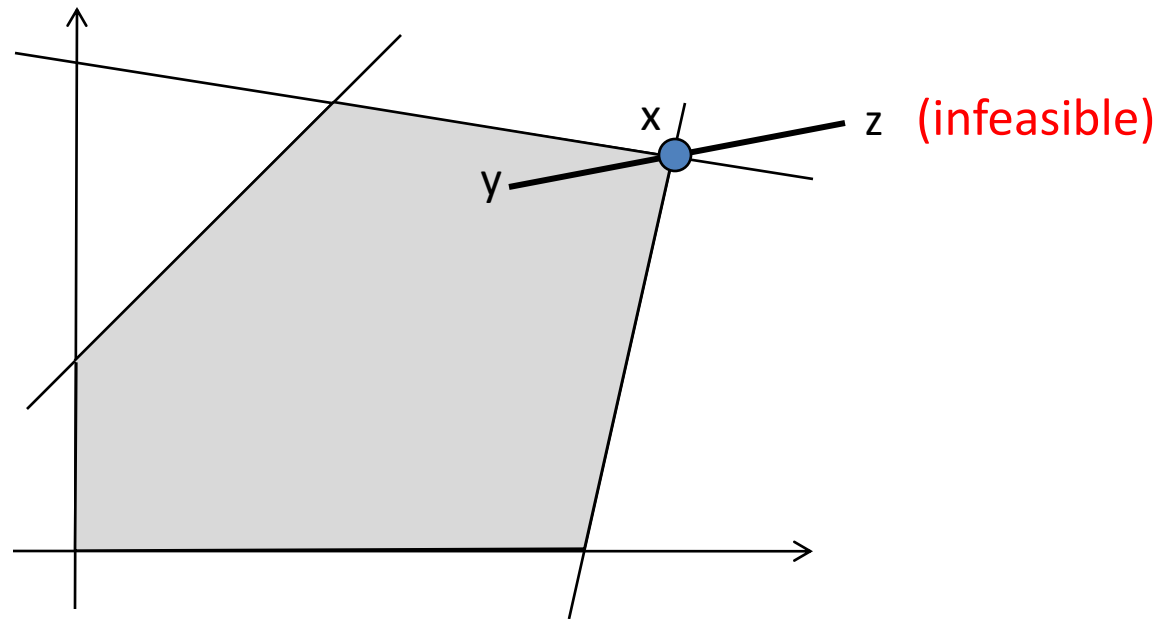
Pitfall #1: What is a corner point?

- Attempt #1: “x is the ‘farthest point’ in some direction”
- Let $P = \{ \text{feasible region} \}$
- There exists $c \in \mathbb{R}^n$ s.t. $c^T x > c^T y$ for all $y \in P \setminus \{x\}$
- “For some objective function, x is the unique optimal point when maximizing over P”
- Such a point x is called a “**vertex**”



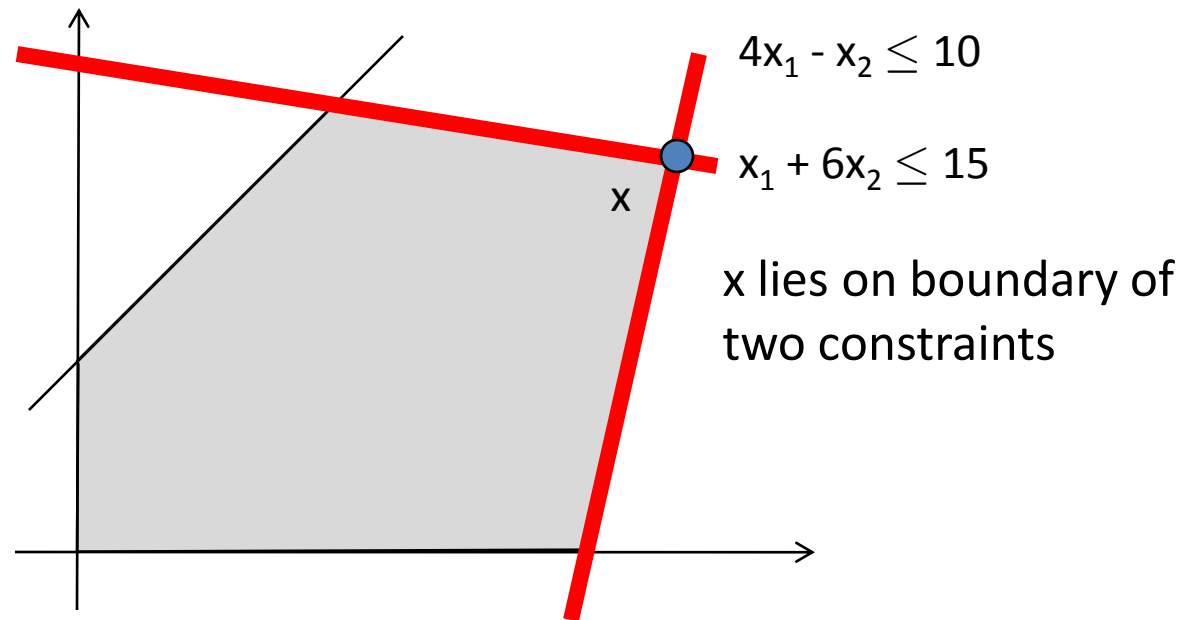
Pitfall #1: What is a corner point?

- Attempt #2: “There is no feasible line-segment that goes through x in both directions”
- Whenever $x = \alpha y + (1 - \alpha)z$ with $y, z \neq x$ and $\alpha \in (0, 1)$, then either y or z must be infeasible.
- “If you write x as a convex combination of two feasible points y and z , the only possibility is $x = y = z$ ”
- Such a point x is called an “**extreme point**”



Pitfall #1: What is a corner point?

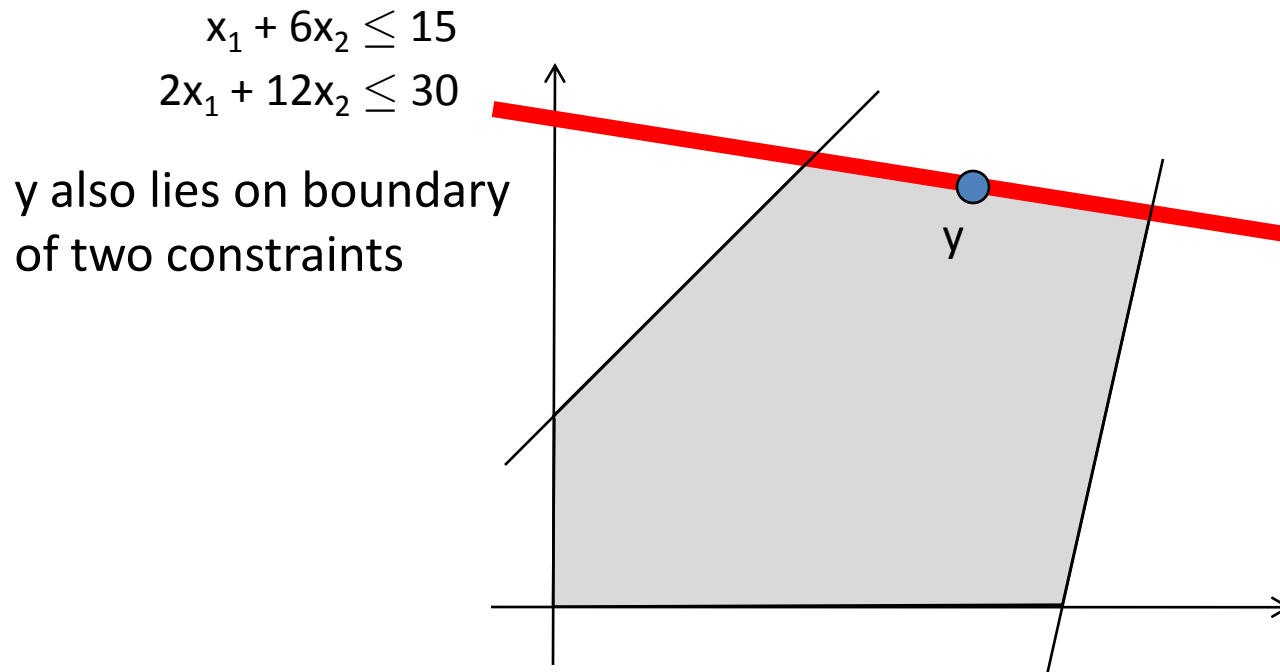
- Attempt #3: “x lies on the boundary of many constraints”
- Note: This discussion differs from textbook



Pitfall #1: What is a corner point?

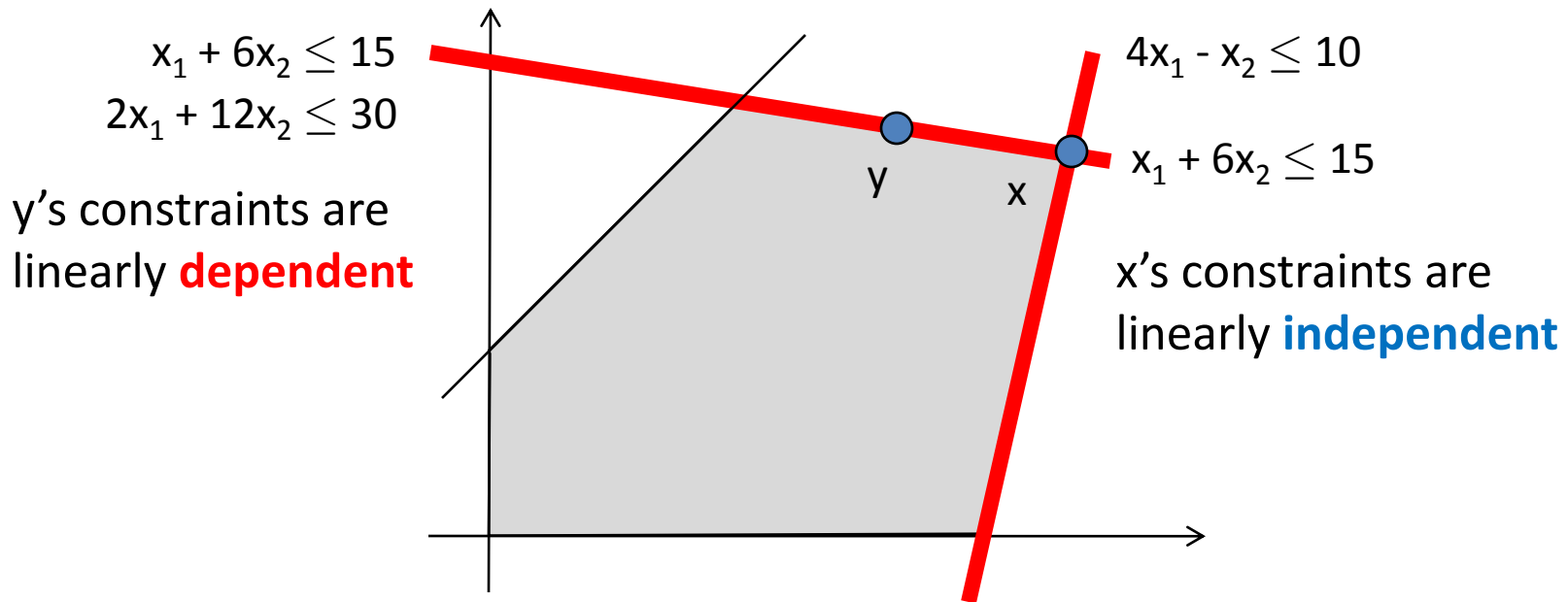
- Attempt #3: “x lies on the boundary of **many** constraints”
- **Note:** This discussion differs from textbook
- What if I introduce **redundant** constraints?

Not the right
condition



Pitfall #1: What is a corner point?

- Revised Attempt #3: “ x lies on the boundary of many **linearly independent constraints**”
- Feasible region: $P = \{ x : a_i^T x \leq b_i \ \forall i \} \subset \mathbb{R}^n$
- Let $\mathcal{I}_x = \{ i : a_i^T x = b_i \}$ and $\mathcal{A}_x = \{ a_i : i \in \mathcal{I}_x \}$. (“**Tight constraints**”)
- x is a “**basic feasible solution (BFS)**” if $\text{rank } \mathcal{A}_x = n$



Lemma: Let P be a polyhedron. The following are equivalent.

- i. x is a vertex
- ii. x is an extreme point
- iii. x is a basic feasible solution (BFS)

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Proof of (i) \Rightarrow (ii):

x is a vertex $\Rightarrow \exists c$ s.t. x is unique maximizer of $c^T x$ over P

Suppose $x = \alpha y + (1-\alpha)z$ where $y, z \in P$ and $\alpha \in (0,1)$.

Suppose $y \neq x$. Then

$$c^T x = \underbrace{\alpha c^T y}_{< c^T x} + (1-\alpha) \underbrace{c^T z}_{\leq c^T x} \quad \begin{array}{l} \text{(since } c^T x \text{ is optimal value)} \\ \text{(since } x \text{ is unique optimizer)} \end{array}$$

$$\Rightarrow c^T x < \alpha c^T x + (1-\alpha) c^T x = c^T x \quad \textbf{Contradiction!}$$

So $y=x$. Symmetrically, $z=x$.

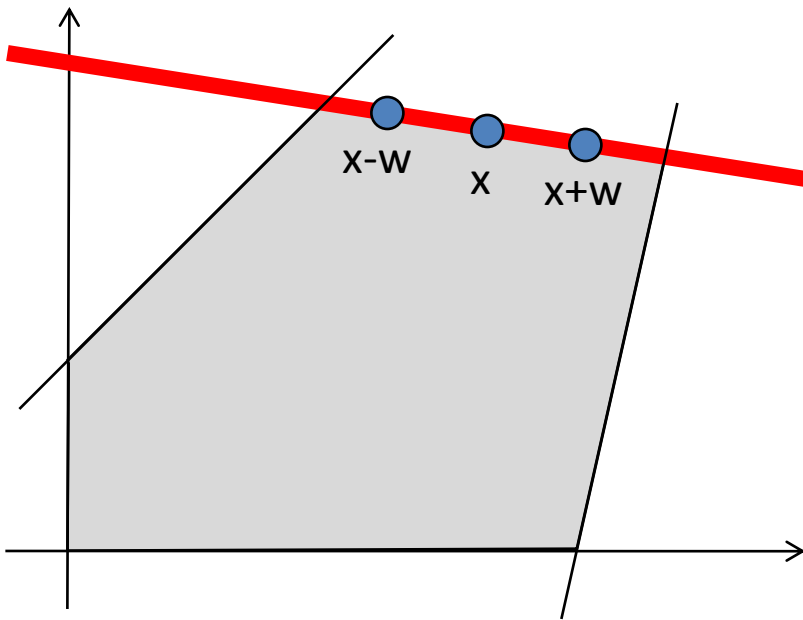
So x is an extreme point of P . ■

Lemma: Let $P = \{x : a_i^T x \leq b_i \ \forall i\} \subset \mathbb{R}^n$. The following are equivalent.

- i. x is a vertex
- ii. x is an extreme point
- iii. x is a basic feasible solution (BFS)

Proof Idea of (ii) \Rightarrow (iii):

x **not** a BFS $\Rightarrow \text{rank } \mathcal{A}_x \leq n-1$



- Each tight constraint removes one degree of freedom
- At least one degree of freedom remains
- So x can “wobble” while staying on all the tight constraints
- Then x is a convex combination of two points obtained by “wiggling”.
- So x is not an extreme point.

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \} \subset \mathbb{R}^n$. The following are equivalent.

- i. x is a vertex
- ii. x is an extreme point
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Proof of (ii) \Rightarrow (iii): We'll show contrapositive.

x **not** a BFS $\Rightarrow \text{rank } \mathcal{A}_x < n$ (Recall $\mathcal{A}_x = \{ a_i : a_i^T x = b_i \}$)

Claim: $\exists w \in \mathbb{R}^n, w \neq 0$, s.t. $a_i^T w = 0 \ \forall a_i \in \mathcal{A}_x$ (w orthogonal to all of \mathcal{A}_x)

Proof: Let M be matrix whose rows are the a_i 's in \mathcal{A}_x .

$\dim \text{row-space}(M) + \dim \text{null-space}(M) = n$

But $\dim \text{row-space}(M) < n \Rightarrow \exists w \neq 0$ in the null space. \square

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \} \subset \mathbb{R}^n$. The following are equivalent.

- i. x is a vertex
- ii. x is an extreme point
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Claim: $\exists w \in \mathbb{R}^n, w \neq 0$, s.t. $a_i^T w = 0 \ \forall a_i \in \mathcal{A}_x$

(w orthogonal to all of \mathcal{A}_x)

Let $y = x + \epsilon w$ and $z = x - \epsilon w$, where $\epsilon > 0$.

Claim: If ϵ very small then $y, z \in P$.

Proof: First consider tight constraints at x . (i.e., those in \mathcal{I}_x)

$$a_i^T y = a_i^T x + \epsilon a_i^T w = b_i + 0$$

So y satisfies this constraint. Similarly for z .

Next consider the loose constraints at x . (i.e., those not in \mathcal{I}_x)

$$b_i - a_i^T y = \underbrace{b_i - a_i^T x}_{\text{Positive}} - \underbrace{\epsilon a_i^T w}_{\text{As small as we like}} \geq 0$$

Positive **As small as we like**

So y satisfies these constraints. Similarly for z . \square

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \} \subset \mathbb{R}^n$. The following are equivalent.

- i. x is a vertex
- ii. x is an extreme point
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Claim: $\exists w \in \mathbb{R}^n, w \neq 0$, s.t. $a_i^T w = 0 \ \forall a_i \in \mathcal{A}_x$ (w orthogonal to all of \mathcal{A}_x)

Let $y = x + \epsilon w$ and $z = x - \epsilon w$, where $\epsilon > 0$.

Claim: If ϵ very small then $y, z \in P$.

Then $x = \alpha y + (1 - \alpha)z$, where $y, z \in P$, $y, z \neq x$, and $\alpha = 1/2$.

So x is **not** an extreme point. ■

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \} \subset \mathbb{R}^n$. The following are equivalent.

- i. x is a vertex
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- iii. x is a basic feasible solution (BFS)

Proof of (iii) \Rightarrow (i):

Let x be a BFS $\Rightarrow \text{rank } \mathcal{A}_x = n$

(Recall $\mathcal{A}_x = \{ a_i : a_i^T x = b_i \}$)

Let $c = \sum_{i \in \mathcal{I}_x} a_i$.

Claim: $c^T x = \sum_{i \in \mathcal{I}_x} b_i$

Proof: $c^T x = \sum_{i \in \mathcal{I}_x} a_i^T x = \sum_{i \in \mathcal{I}_x} b_i$. \square

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \} \subset \mathbb{R}^n$. The following are equivalent.

- i. x is a vertex
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(Recall $\mathcal{A}_x = \{ a_i : a_i^T x = b_i \}$)

Let $c = \sum_{i \in \mathcal{I}_x} a_i$.

Claim: $c^T x = \sum_{i \in \mathcal{I}_x} b_i$

Claim: x is an optimal point of $\max \{ c^T x : x \in P \}$.

Proof: $y \in P \Rightarrow a_i^T y \leq b_i$ for all i

$\Rightarrow c^T y = \sum_{i \in \mathcal{I}_x} a_i^T y \leq \sum_{i \in \mathcal{I}_x} b_i = c^T x. \quad \square$

If one of these is strict,
then this is strict.

Claim: x is the **unique** optimal point of $\max \{ c^T x : x \in P \}$.

Proof: If for any $i \in \mathcal{I}_x$ we have $a_i^T y < b_i$ then $c^T y < c^T x$.

So every optimal point y has $a_i^T y = b_i$ for all $i \in \mathcal{I}_x$.

Since $\text{rank } \mathcal{A}_x = n$, there is only one solution: $y = x! \quad \square$

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \} \subset \mathbb{R}^n$. The following are equivalent.

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Proof of (iii) \Rightarrow (i):

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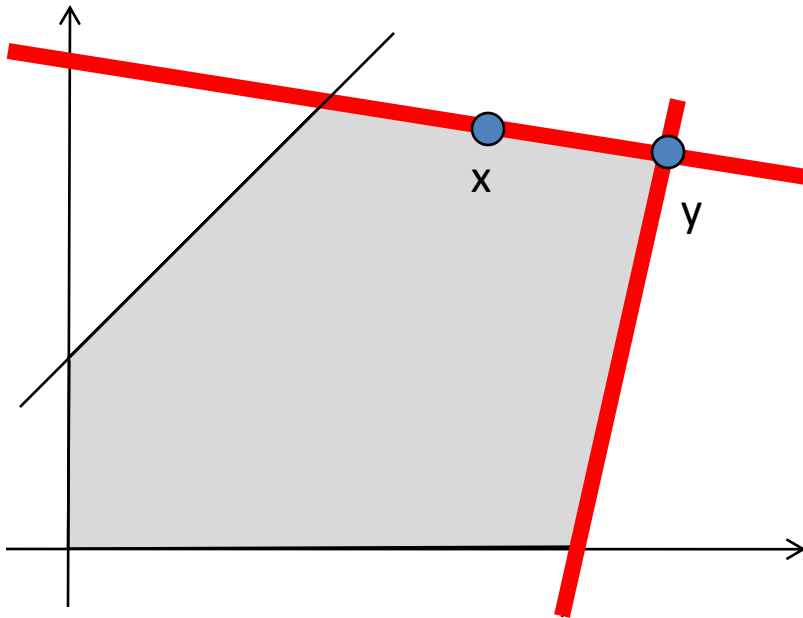
So x is a vertex. ■

More on corner points

Definition: A **line** is a set $L = \{ r + \lambda s : \lambda \in \mathbb{R} \}$ where $r, s \in \mathbb{R}^n$ and $s \neq 0$.

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \}$. Suppose P is non-empty and P does not contain any line. Then P has a corner point.

Proof Idea: Pick any $x \in P$. Suppose x not a BFS.



- At least one degree of freedom remains at x
- So x can “wiggle” while staying on all the tight constraints
- x cannot wiggle off to infinity in both directions because P contains no line
- So when x wiggles, it hits a constraint
- When it hits first constraint, it is still feasible.
- So we have found a point y which has a new tight constraint.

Definition: A **line** is a set $L = \{ r + \lambda s : \lambda \in \mathbb{R} \}$ where $r, s \in \mathbb{R}^n$ and $s \neq 0$.

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \}$. Suppose P is non-empty and P does not contain any line. Then P has a corner point.

Proof: Pick $x \in P$. Suppose x not a BFS.

Claim: $\exists w \in \mathbb{R}^n, w \neq 0$, s.t. $a_i^T w = 0 \ \forall i \in \mathcal{I}_x$ (We saw this before)

Let $y(\epsilon) = x + \epsilon w$. Note $y(0) = x \in P$.

Claim: $\exists \epsilon$ s.t. $y(\epsilon) \notin P$. WLOG $\epsilon > 0$. (Otherwise P contains a line)

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Claim: $\exists \epsilon \text{ s.t. } y(\epsilon) \notin P$. WLOG $\epsilon > 0$. (Otherwise P contains a line)

So set $\delta = 0$ and gradually increase δ . What is largest δ s.t. $x \in P$?

$$\begin{aligned} y(\delta) \in P &\Leftrightarrow a_i^T y(\delta) \leq b_i \ \forall i \\ &\Leftrightarrow a_i^T x + \delta a_i^T w \leq b_i \ \forall i && \text{(Always satisfied if } a_i^T w \leq 0) \\ &\Leftrightarrow \delta \leq \underbrace{(b_i - a_i^T x) / a_i^T w}_{\text{this}} \ \forall i \text{ s.t. } a_i^T w > 0 \end{aligned}$$

Let h be the i that minimizes **this**. Then $\delta = (b_h - a_h^T x) / a_h^T w$.

Claim: $\mathcal{I}_x \subseteq \mathcal{I}_{y(\delta)}$.

Proof: If $i \in \mathcal{I}_x$ then $a_i^T x = b_i$. But $a_i^T w = 0$, so $a_i^T y(\delta) = b_i$ too. \square

Lemma: Let $P = \{x : a_i^T x \leq b_i \ \forall i\}$. Suppose P is non-empty and P does not contain any line. Then P has a corner point.

Proof: Pick $x \in P$. Suppose x not a BFS.

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Claim: $\mathcal{I}_x \subseteq \mathcal{I}_{y(\delta)}$.

Claim: $h \in \mathcal{I}_{y(\delta)} \setminus \mathcal{I}_x$. ($y(\delta)$ has at least one more tight constraint)

Proof: By definition $a_h^T w > 0$, so $h \notin \mathcal{I}_x$.

But $a_h^T y(\delta) = a_h^T x + \delta a_h^T w = a_h^T x + ((b_h - a_h^T x) / a_h^T w) a_h^T w = b_h \Rightarrow h \in \mathcal{I}_{y(\delta)}$. \square

Lemma: Let $P = \{ x : a_i^T x \leq b_i \ \forall i \}$. Suppose P is non-empty and P does not contain any line. Then P has a corner point.

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Let h be the i that minimizes **this**. Then $\delta = (b_h - a_h^T x) / a_h^T w$.

Claim: $\mathcal{I}_x \subseteq \mathcal{I}_{y(\delta)}$.

Claim: $h \in \mathcal{I}_{y(\delta)} \setminus \mathcal{I}_x$. ($y(\delta)$ has at least one more tight constraint)

Claim: $a_h \notin \text{span}(\mathcal{A}_x)$.

Proof: $a_h^T x < b_h$ but $a_h^T y(\delta) = b_h \Rightarrow 0 \neq a_h^T (y(\delta) - x) = \epsilon a_h^T w$.

But, $a_i^T w = 0 \ \forall a_i \in \mathcal{A}_x \Rightarrow a_i^T w = 0 \ \forall a_i \in \text{span}(\mathcal{A}_x) \Rightarrow a_h \notin \text{span}(\mathcal{A}_x)$. \square

Lemma: Let $P = \{x : a_i^T x \leq b_i \ \forall i\}$. Suppose P is non-empty and P does not contain any line. Then P has a corner point.

Proof: Pick $x \in P$. Suppose x not a BFS.

Claim: $\exists w \in \mathbb{R}^n, w \neq 0, \text{ s.t. } a_i^T w = 0 \ \forall i \in \mathcal{I}_x$ (We saw this before)

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Claim: $\mathcal{I}_x \subseteq \mathcal{I}_{y(\delta)}$.

Claim: $h \in \mathcal{I}_{y(\delta)} \setminus \mathcal{I}_x$. ($y(\delta)$ has at least one more tight constraint)

Claim: $a_h \notin \text{span}(\mathcal{A}_x)$.

So $\text{rank } \mathcal{A}_{y(\delta)} > \text{rank } \mathcal{A}_x$. Repeat this argument with $y(\delta)$ instead of x .

Eventually find z with $\text{rank } \mathcal{A}_z = n \Rightarrow z$ is a BFS. ■

Local-Search Algorithm: Pitfalls & Details

Algorithm

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For each corner point y that is a neighbor of x

 If $c^T y > c^T x$ then set $x = y$

Halt



What is a corner point?

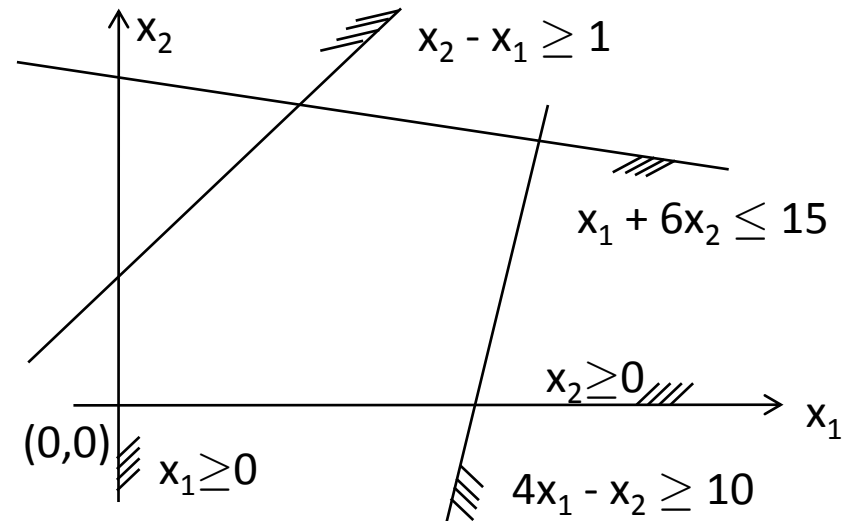
2. What if there are no corner points?
3. What are the “neighboring” corner points?
4. What if there are no neighboring corner points?
5. How can I find a starting corner point?
6. Does the algorithm terminate?
7. Does it produce the right answer?

Pitfall #2: No corner points?

- This is possible
 - Case 1: LP infeasible

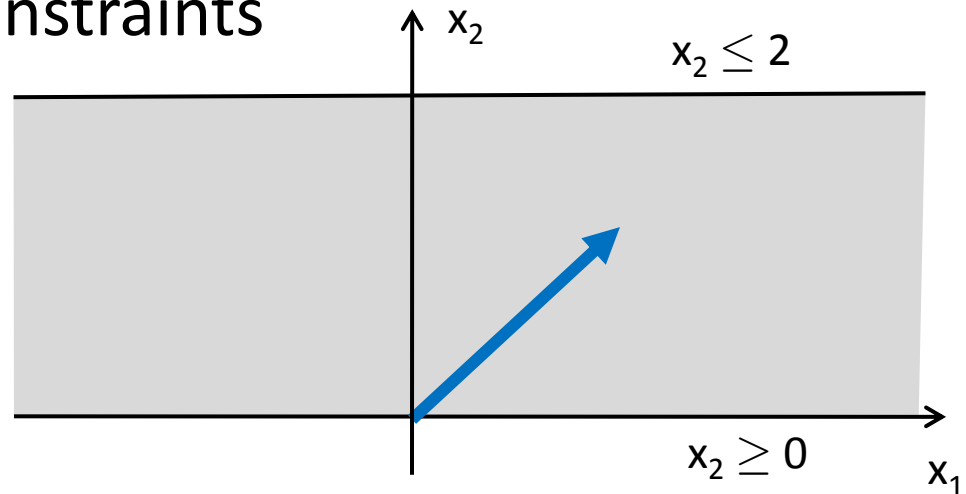
This is unavoidable.

Algorithm must detect this case.



- Case 2: Not enough constraints

$$\begin{array}{ll}\max & x_1 + x_2 \\ \text{s.t.} & x_2 \leq 2 \\ & x_2 \geq 0\end{array}$$



A Fix!

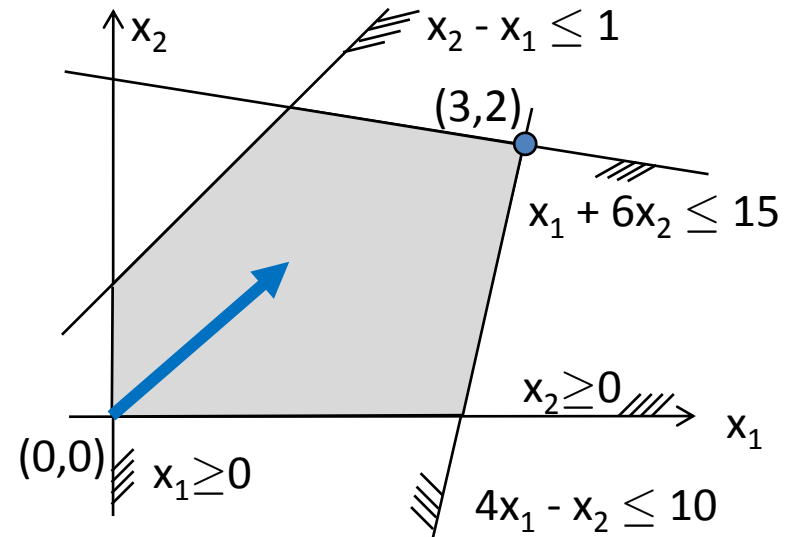
We avoid this case by manipulating the LP a bit...

Converting to Equational Form

- General form of an LP

$$\begin{array}{ll}\max & c^T x \\ \text{s.t.} & Ax \leq b\end{array}$$

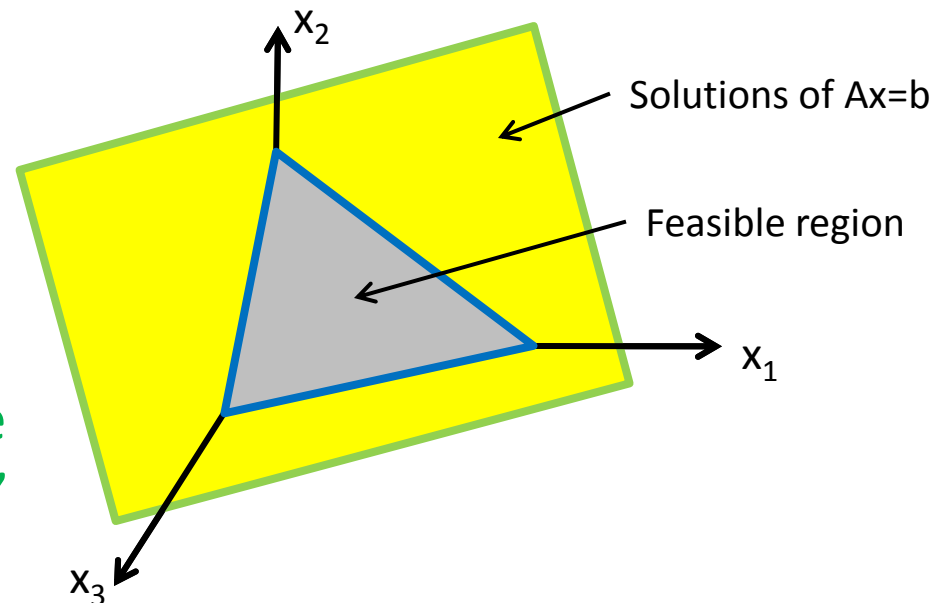
“Intersection of finitely many half-spaces”



- Another form of an LP

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

“Intersection of an affine space with the non-negative orthant”



Converting to Equational Form

- General form of an LP

$$\begin{array}{ll}\max & c^T x \\ \text{s.t.} & Ax \leq b\end{array}$$

“Inequality form”
or “Canonical form”

- Another form of an LP

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

“Equational form”
or “Standard form”

- **Claim:** These two forms of LPs are equivalent.