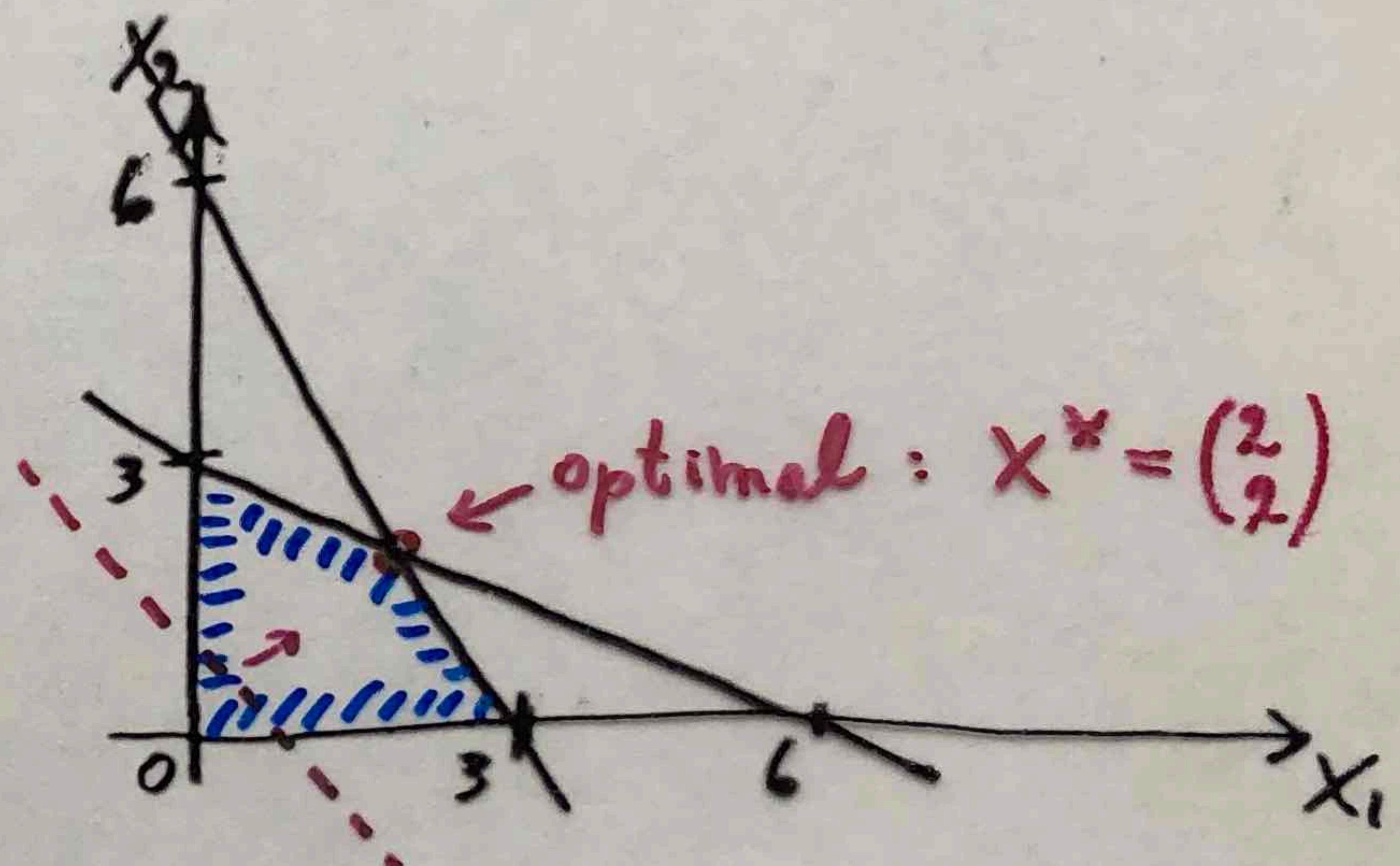


Towards the Simplex Method

- "Simplex" is in top ten algorithms having "the greatest influence on the development and practice of science and engineering in the 20th century".
 (Journal of Computing in Science & Engineering)

- Recall the example:

$$\begin{aligned} \text{max } & X_1 + X_2 \\ \text{s.t. } & 2X_1 + 4X_2 \leq 12 \\ & 2X_1 + X_2 \leq 6 \\ & X_1, X_2 \geq 0 \end{aligned}$$



- Intersection points of constraint boundaries are important; characterize them.

- Definition: Let $Ax \leq b$ be a p by n system of linear inequalities (can include $x \geq 0$). Then a sol-n \bar{x} is called a Corner Point Feasible Solution (CPF) if

- i) $A\bar{x} \leq b$ (\bar{x} is feasible)
- ii) n or more inequality constraints are satisfied at equality.

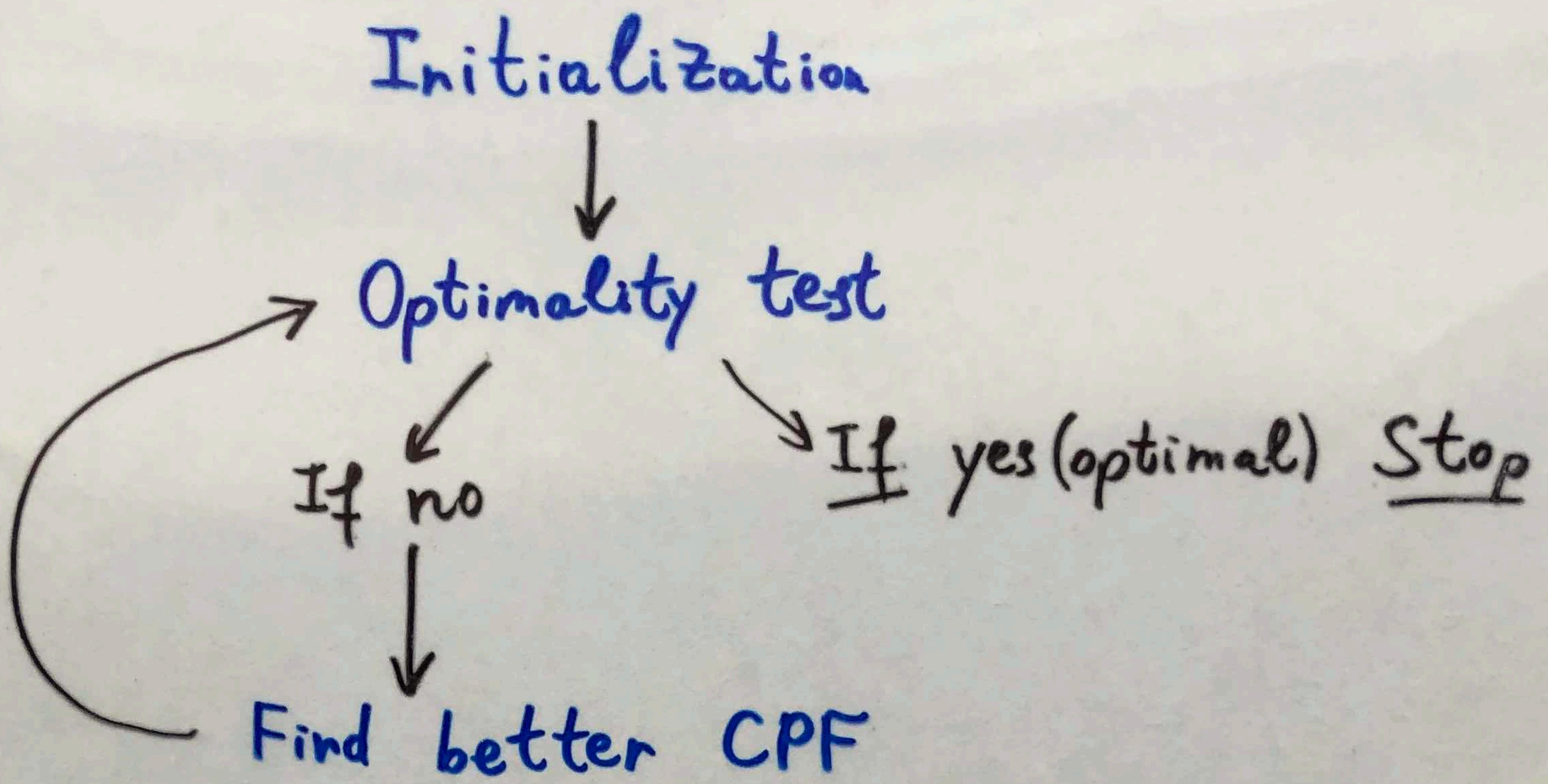
- Examples: $(0,0)$, $(3,0)$, $(0,3)$, $(2,2)$ are CPF sol-n's.

- Def-n: Extreme point of $\{x : Ax \leq b\} = S$ is a point in the set that does not lie on any line segment joining two other points of S .
 - Extreme points and CPF's are the same in LP.
-
- Def-n: Two CPF's are adjacent to each other if they share $n-1$ constraint boundaries.
Ex.: $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$ and $\begin{pmatrix} 0 \\ 3 \end{pmatrix}$ are adjacent in our ex.
 - Def-n: The line segment connecting two adjacent CPF's is called an edge.

Some solution concepts for Simplex

- If an LP has an optimal sol-n then there is a CPF which is optimal.
Thus, look for an optimal CPF.
- If a CPF has no adjacent CPF solutions that are better (in terms of obj. f-n) then it must be an optimal sol-n.
otherwise move to a better adjacent CPF.

The structure of Simplex



How to realize this algebraically?

Idea: Introduce another description of CPF which is easy to manipulate algebraically.

- Add slack variables X_3 :

$$\left\{ \begin{array}{l} Ax \leq b \\ x \geq 0 \end{array} \right\} \rightarrow \left\{ \begin{array}{l} Ax + X_3 = b \\ x \geq 0, X_3 \geq 0 \end{array} \right\} \rightarrow \text{augmented system}$$

Advantage: Equality system, easy to treat

Disadvantage: # of variables increased by m

Example: $\begin{aligned} 2x_1 + x_2 &\leq 6 \\ 2x_1 + 4x_2 &\leq 12 \\ x_1, x_2 &\geq 0 \end{aligned}$

$$\begin{aligned} 2x_1 + x_2 + X_3 &= 6 \\ 2x_1 + 4x_2 + X_4 &= 12 \\ x_1, x_2, X_3, X_4 &\geq 0 \end{aligned}$$

original variables: $\bar{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$

slack variables: $\bar{X} = \begin{pmatrix} X_3 \\ X_4 \end{pmatrix}$

$$\left\{ \begin{array}{l} Ax \leq b \\ x \geq 0 \end{array} \right\} \leftrightarrow \left\{ \begin{array}{l} Ax + x_3 = b \\ x \geq 0, x_3 \geq 0 \end{array} \right\} \leftrightarrow \left\{ \begin{array}{l} [A \ I] \begin{bmatrix} x \\ x_3 \end{bmatrix} = b \\ \begin{bmatrix} x \\ x_3 \end{bmatrix} \geq 0 \end{array} \right\}$$

- Want to solve $[A \ I] \begin{bmatrix} x \\ x_3 \end{bmatrix} = b$.
- A has size $m \times n \rightarrow [A \ I]$ has size $m \times (n+m)$
 \rightarrow we have n degrees of freedom \rightarrow
set n variables to 0 and solve the rest

Example: $2x_1 + x_2 + x_3 = 6$
 $2x_1 + 4x_2 + x_4 = 12$

$m=2, n+m=4 \rightarrow$ 2 degrees of freedom

\rightarrow set 2 variables to 0

▼ set $x_1 = x_2 = 0$; solve $x_3 = 6$,
 $x_4 = 12$;

get solution $\begin{bmatrix} x \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 6 \\ 12 \end{bmatrix}$ or $x = \begin{bmatrix} 0 \\ 0 \\ 6 \\ 12 \end{bmatrix}$

▼ set $x_3 = x_4 = 0$; solve $2x_1 + x_2 = 6$,
 $2x_1 + 4x_2 = 12$;

get solution $\begin{bmatrix} x \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 2 \\ 0 \\ 0 \end{bmatrix}$ or $x = \begin{bmatrix} 2 \\ 2 \\ 0 \\ 0 \end{bmatrix}$

was the optimal
CPF

▼ set $x_1 = x_3 = 0$; solve $\begin{array}{l} x_2 = 6 \\ 4x_2 + x_4 = 12 \end{array}$

get $\begin{bmatrix} x \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 6 \\ 0 \\ -12 \end{bmatrix}$ or $x = \begin{bmatrix} 0 \\ 6 \\ 0 \\ -12 \end{bmatrix}$

was not feasible

- When does this work? What if the resulting system is singular?

Example: $\left\{ \begin{array}{l} x_1 \leq 5 \\ x_1 + x_2 \leq 10 \\ x_1, x_2 \geq 0 \end{array} \right\} \rightarrow \left\{ \begin{array}{l} x_1 + x_3 = 5 \\ x_1 + x_2 + x_4 = 10 \\ x_1, x_2, x_3, x_4 \geq 0 \end{array} \right\}$

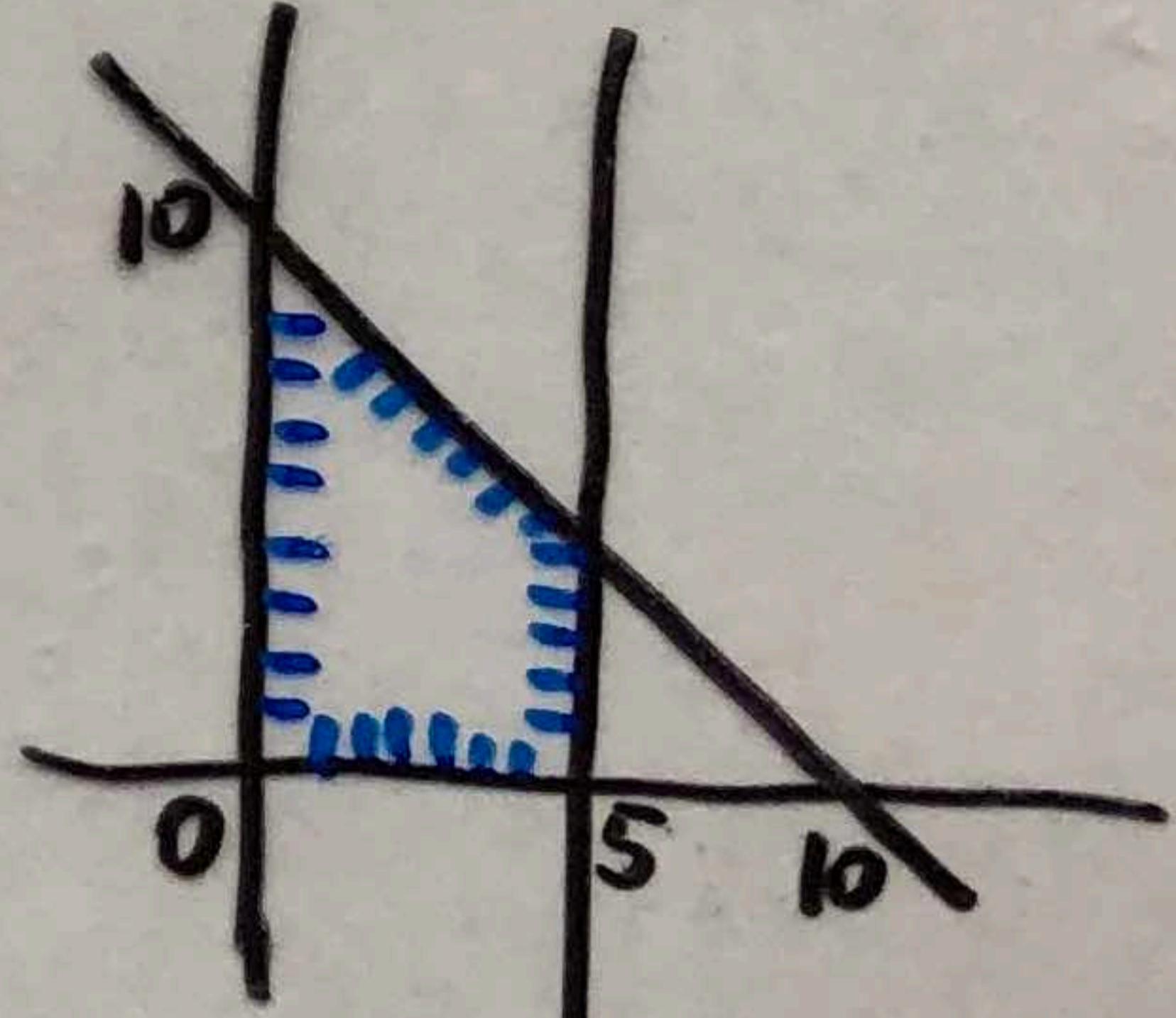
▼ set $x_1 = x_3 = 0$;

Resulting system:

$$\begin{pmatrix} 0 & 0 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_2 \\ x_4 \end{pmatrix} = \begin{pmatrix} 5 \\ 10 \end{pmatrix}$$

singular → no sol-n

(parallel lines don't intersect)



Basic solutions

Def-n: Given a system of linear equalities $A\bar{x} = b$, where $A \in \mathbb{R}^{P \times Q}$, $Q > P$,
 a solution $\bar{x} \in \mathbb{R}^Q$ is called basic solution if

- i) $A\bar{x} = b$
- ii) $Q-P$ variables are set to 0
 (denote these variables \bar{x}_N and call them nonbasic variables)
- iii) the columns corresponding to remaining P variables are linearly independent
 (denote these variables \bar{x}_B and call them basic variables).

- Thus, $\bar{x} = (\bar{x}_B, \bar{x}_N) = (\bar{x}_B, \bar{0})$
- In our case, the system was

$$[A \ I] \begin{bmatrix} x \\ x_N \end{bmatrix} = b$$

where $[A \ I] \in \mathbb{R}^{m \times (n+m)}$.

Thus, the number of nonbasic variables is

$$(n+m)-m=n;$$

the ——— basic ———
 m .