Enumerating all basic feasible solutions (BFS), in order to find the optimum is slow.

Simplex Algorithm [George Dantzig 1947] Move from BFS to adjacent BFS, without decreasing objective function.

Two BFSs are called adjacent if the bases just differ in one variable.



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 $\begin{array}{l} \max \ 13a + 23b \\ \text{s.t.} \ 5a + 15b + s_c &= 480 \\ 4a + 4b &+ s_h &= 160 \\ 35a + 20b &+ s_m = 1190 \\ a , b , s_c , s_h , s_m \ge 0 \end{array}$





4 Simplex Algorithm

 $\begin{array}{ll} \max & 13a + 23b \\ \text{s.t.} & 5a + 15b + s_c & = 480 \\ & 4a + 4b & + s_h & = 160 \\ & 35a + 20b & + s_m = 1190 \\ & a & , & b & , s_c & , s_h & , s_m \ge 0 \end{array}$

max Z		basis = { s_c, s_h, s_m }
13a + 23b –	Z = 0	A = B = 0
$5a + 15b + s_c$	= 480	Z = 0
$4a + 4b + s_h$	= 160	$s_c = 480$
$35a + 20b + s_m$	= 1190	$s_h = 160$
a, b, s _c , s _h , s _m	≥ 0	$s_m = 1190$



4 Simplex Algorithm

max Z	
$13a + 23b \qquad -Z = 0$	
$5a + 15b + s_c = 480$	
$4a + 4b + s_h = 160$	
$35a + 20b + s_m = 1190$	
a , b , s_c , s_h , $s_m \ge 0$	JL

basis =
$$\{s_c, s_h, s_m\}$$

 $a = b = 0$
 $Z = 0$
 $s_c = 480$
 $s_h = 160$
 $s_m = 1190$

- choose variable to bring into the basis
- chosen variable should have positive coefficient in objective function
- apply devices test to find out by how much the variable can be increased
- pivot on row found by min-ratio test
- the existing basis variable in this row leaves the basis

max Z		
13a + 23b	-Z = 0	basis = $\{s_c, s_h, s_m\}$ a = b = 0
	-	$\begin{array}{c} u = b = 0 \\ Z = 0 \end{array}$
$5a + 15b + s_c$	= 480	$\Sigma = 0$
$4a + 4b + s_h$	= 160	$s_c = 480$
$35a + 20b + s_m$	= 1190	$s_h = 160$
a, b, s_c, s_h, s_m	≥ 0	$s_m = 1190$

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max Z		
13a + 23b	-Z = 0	basis = $\{s_c, s_h, s_m\}$ a = b = 0
		$\begin{array}{c} a &= b = 0 \\ Z &= 0 \end{array}$
$5a + 15b + s_c$	= 480	
$4a + 4b + s_h$	= 160	$s_c = 480$
35a + 20b + s	m = 1190	$s_h = 160$ $s_m = 1190$
a, b, s_c, s_h, s_c	$m \geq 0$	3m-1190

- choose variable to bring into the basis
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max Z		
13a + 23b	-Z = 0	basis = $\{s_c, s_h, s_m\}$ a = b = 0
		$\begin{array}{c} a = b = 0 \\ Z = 0 \end{array}$
$5a + 15b + s_c$	= 480	
$4a + 4b + s_h$	= 160	$s_c = 480$
35a + 20b + s	m = 1190	$s_h = 160$ $s_m = 1190$
a, b, s_c, s_h, s_c	$m \geq 0$	3m-1190

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- pivot on row found by min-ratio test
- the existing basis variable in this row leaves the basis

max Z		basis = { s_c, s_h, s_m }
13a + 23b	-Z = 0	a = b = 0
$5a + 15b + s_c$	= 480	Z = 0
		$s_{c} = 480$
$4a + 4b + s_h$	= 160	$s_c = 480$ $s_h = 160$
$35a + 20b + s_m$	a = 1190	$s_m = 100$ $s_m = 1190$
$[a, b, s_c, s_h, s_m]$	$_{i} \geq 0$	0 1100

- choose variable to bring into the basis
- chosen variable should have positive coefficient in objective function
- apply min-ratio test to find out by how much the variable can be increased
- pivot on row found by min-ratio test
- the existing basis variable in this row leaves the basis

max Z		basis = { s_c, s_h, s_m }
13a + 23b	-Z = 0	a = b = 0
$5a + 15b + s_c$	= 480	Z = 0
		$s_{c} = 480$
$4a + 4b + s_h$	= 160	$s_c = 480$ $s_h = 160$
$35a + 20b + s_m$	a = 1190	$s_m = 100$ $s_m = 1190$
$[a, b, s_c, s_h, s_m]$	$_{i} \geq 0$	0 1100

- choose variable to bring into the basis
- chosen variable should have positive coefficient in objective function
- apply min-ratio test to find out by how much the variable can be increased
- pivot on row found by min-ratio test
- the existing basis variable in this row leaves the basis

max Z		
13a + 23b	-Z=0	
$5a + 15b + s_c$	= 480	
$4a + 4b + s_h$	= 160	
$35a + 20b + s_m$	= 1190	
a, b, s _c , s _h , s _m	≥ 0	

$basis = \{s_c, s_h, s_m\}$
a = b = 0
Z = 0
$s_c = 480$
$s_h = 160$
$s_m = 1190$

max Z		basis = { s_c, s_h, s_m }
13a + 23 b –	Z = 0	a = b = 0
$5a + 15b + s_c$	= 480	Z = 0
$4a + 4b + s_h$	= 160	$s_c = 480$
$35a + 20b + s_m$	= 1190	$s_h = 160$
a , b , s_c , s_h , s_m	≥ 0	$s_m = 1190$

• Choose variable with coefficient ≥ 0 as entering variable.

max Z		basis = { s_c, s_h, s_m }
13a + 23b –	Z = 0	a = b = 0
$5a + 15b + s_c$	= 480	Z = 0
$4a + 4b + s_h$	= 160	$s_c = 480$
$35a + 20b + s_m$	= 1190	$s_h = 160$
a , b , s_c , s_h , s_m	≥ 0	$s_m = 1190$

- Choose variable with coefficient ≥ 0 as entering variable.
- ▶ If we keep a = 0 and increase b from 0 to $\theta > 0$ s.t. all constraints ($Ax = b, x \ge 0$) are still fulfilled the objective value Z will strictly increase.

max Z		basis = { s_c, s_h, s_m }
13a + 23 b –	Z = 0	a = b = 0
$5a + 15b + s_c$	= 480	Z = 0
$4a + 4b + s_h$	= 160	$s_c = 480$
$35a + 20b + s_m$	= 1190	$s_h = 160$
a , b , s_c , s_h , s_m	≥ 0	$s_m = 1190$

- Choose variable with coefficient ≥ 0 as entering variable.
- If we keep a = 0 and increase b from 0 to θ > 0 s.t. all constraints (Ax = b, x ≥ 0) are still fulfilled the objective value Z will strictly increase.
- For maintaining Ax = b we need e.g. to set $s_c = 480 15\theta$.

max Z		basis = { s_c, s_h, s_m }
13a + 23b –	Z = 0	a = b = 0
$5a + 15b + s_c$	= 480	Z = 0
$4a + 4b + s_h$	= 160	$s_c = 480$
$35a + 20b + s_m$	= 1190	$s_h = 160$
a , b , s_c , s_h , s_m	≥ 0	$s_m = 1190$

- Choose variable with coefficient ≥ 0 as entering variable.
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- For maintaining Ax = b we need e.g. to set $s_c = 480 15\theta$.
- Choosing \(\theta\) = min{480/15, 160/4, 1190/20}\) ensures that in the new solution one current basic variable becomes 0, and no variable goes negative.

max Z		basis = { s_c, s_h, s_m }
13a + 23b –	Z = 0	a = b = 0
$5a + 15b + s_c$	= 480	Z = 0
$4a + 4b + s_h$	= 160	$s_c = 480$
$35a + 20b + s_m$	= 1190	$s_h = 160$
a, b, s _c , s _h , s _m	≥ 0	$s_m = 1190$

- Choose variable with coefficient ≥ 0 as entering variable.
- ▶ If we keep a = 0 and increase b from 0 to $\theta > 0$ s.t. all constraints ($Ax = b, x \ge 0$) are still fulfilled the objective value Z will strictly increase.
- For maintaining Ax = b we need e.g. to set $s_c = 480 15\theta$.
- Choosing \(\theta\) = min{480/15, 160/4, 1190/20}\) ensures that in the new solution one current basic variable becomes 0, and no variable goes negative.
- The basic variable in the row that gives min{480/15, 160/4, 1190/20} becomes the leaving variable.

max Z	
13a + 23b	-Z = 0
$5a + 15b + s_c$	= 480
$4a + 4b + s_h$	= 160
$35a + 20b + s_m$	= 1190
a, b, s_c, s_h, s_m	≥ 0

basis =
$$\{s_c, s_h, s_m\}$$

 $a = b = 0$
 $Z = 0$
 $s_c = 480$
 $s_h = 160$
 $s_m = 1190$

Substitute $b = \frac{1}{15}(480 - 5a - s_c)$.

max Z	
13a + 23b –	-Z=0
$5a + 15b + s_c$	= 480
$4a + 4b + s_h$	= 160
$35a + 20b + s_m$	= 1190
a, b, s _c , s _h , s _m	≥ 0

basis =
$$\{s_c, s_h, s_m\}$$

 $a = b = 0$
 $Z = 0$
 $s_c = 480$
 $s_h = 160$
 $s_m = 1190$

Substitute $b = \frac{1}{15}(480 - 5a - s_c)$.

max Z	
13a + 23b –	-Z=0
$5a + 15b + s_c$	= 480
$4a + 4b + s_h$	= 160
$35a + 20b + s_m$	= 1190
a, b, s _c , s _h , s _m	≥ 0

basis =
$$\{s_c, s_h, s_m\}$$

 $a = b = 0$
 $Z = 0$
 $s_c = 480$
 $s_h = 160$
 $s_m = 1190$

Substitute
$$b = \frac{1}{15}(480 - 5a - s_c)$$
.

 $\max Z$ $\frac{\frac{16}{3}a}{\frac{1}{3}a} - \frac{23}{15}s_c & -Z = -736 \\ \frac{1}{3}a + b + \frac{1}{15}s_c & = 32 \\ \frac{8}{3}a & -\frac{4}{15}s_c + s_h & = 32 \\ \frac{85}{3}a & -\frac{4}{3}s_c & +s_m & = 550 \\ a, b, s_c, s_h, s_m & \ge 0$

basis = {
$$b, s_h, s_m$$
}
 $a = s_c = 0$
 $Z = 736$
 $b = 32$
 $s_h = 32$
 $s_m = 550$

max Z	
$\frac{16}{3}a + \frac{23}{15}s_c$	-Z = -736
$\frac{1}{3}a + b + \frac{1}{15}s_c$	= 32
$\frac{8}{3}a + -\frac{4}{15}s_c + s_h$	= 32
$\frac{85}{3}a + - \frac{4}{3}s_c + s_m$	= 550
a,b, s _c ,s _h ,s _m	≥ 0

basis =
$$\{b, s_h, s_m\}$$

 $a = s_c = 0$
 $Z = 736$
 $b = 32$
 $s_h = 32$
 $s_m = 550$

max Z		
$\frac{16}{3}a + \frac{23}{15}s_c$	-Z = -736	basis = $\{b, s_h, s_m\}$
5 15	2.2	$a = s_c = 0$
$\frac{1}{3}a + b + \frac{1}{15}s_c$	= 32	Z = 736
$\frac{8}{3}a + -\frac{4}{15}s_c + s_h$	= 32	b = 32
$\frac{85}{3}a + - \frac{4}{3}s_c + s_m$	= 550	$s_h = 32$
3° 3° 3° 7° 3°	- 550	$s_m = 550$
a, b, s_c, s_h, s_m	≥ 0	

Choose variable *a* to bring into basis.

max Z		
$\frac{16}{3}a + \frac{23}{15}s_c$	-Z = -736	basis = $\{b, s_h, s_m\}$
5 15		$a = s_c = 0$
$\frac{1}{3}a + b + \frac{1}{15}s_c$	= 32	Z = 736
$\frac{8}{3}a + -\frac{4}{15}s_c + s_h$	= 32	b = 32
$\frac{85}{3}a + - \frac{4}{3}s_c + s_m$	= 550	$s_h = 32$
$3^{\alpha} + 3^{\beta} - 3^{\beta$	- 550	$s_m = 550$
a, b, s_c, s_h, s_m	≥ 0	

Choose variable *a* to bring into basis.

Computing min{ $3 \cdot 32$, $3 \cdot 32/8$, $3 \cdot 550/85$ } means pivot on line 2.

max Z		
$\frac{16}{3}a + \frac{23}{15}s_c$	-Z = -736	basis = $\{b, s_h, s_m\}$
5 15		$a = s_c = 0$
$\frac{1}{3}a + b + \frac{1}{15}s_c$	= 32	Z = 736
$\frac{8}{3}a + -\frac{4}{15}s_c + s_h$	= 32	<i>b</i> = 32
$\frac{85}{3}a + - \frac{4}{3}s_c + s_m$	= 550	$s_h = 32$
5 5		$s_m = 550$
a, b, s_c, s_h, s_m	≥ 0	

Choose variable *a* to bring into basis.

Computing min{3 · 32, 3 · 32/8, 3 · 550/85} means pivot on line 2. Substitute $a = \frac{3}{8}(32 + \frac{4}{15}s_c - s_h)$.

max Z		
$\frac{16}{3}a + \frac{23}{15}s_c$ -	-Z = -736	basis = { b, s_h, s_m }
5 15		$a = s_c = 0$
$\frac{1}{3}a + b + \frac{1}{15}s_c$	= 32	Z = 736
$\frac{8}{3}a + -\frac{4}{15}s_c + s_h$	= 32	<i>b</i> = 32
$\frac{85}{3}a + -\frac{4}{3}s_c + s_m$	= 550	$s_h = 32$
5 5		$s_m = 550$
a, b, s_c, s_h, s_m	≥ 0	

Choose variable *a* to bring into basis.
Computing min{
$$3 \cdot 32, 3 \cdot 32/8, 3 \cdot 550/85$$
} means pivot on line 2.
Substitute $a = \frac{3}{8}(32 + \frac{4}{15}s_c - s_h)$.

max Z

	$-s_c - 2s_h - Z$	= -800
	$b + \frac{1}{10}s_c - \frac{1}{8}s_h$	= 28 = 12
а	$-\frac{1}{10}s_{c}+\frac{3}{8}s_{h}$	= 12
	$\frac{3}{2}s_c - \frac{85}{8}s_h + s_m$	= 210
а,	b , s_c , s_h , s_m	≥ 0

basis = {
$$a, b, s_m$$
}
 $s_c = s_h = 0$
 $Z = 800$
 $b = 28$
 $a = 12$
 $s_m = 210$

Pivoting stops when all coefficients in the objective function are non-positive.

- any feasible solution satisfies all equations in the tableaux
- in particular: $Z = 800 s_c 2s_h$, $s_c \ge 0$, $s_h \ge 0$
- hence optimum solution value is at most 800.
- the current solution has value 800



Pivoting stops when all coefficients in the objective function are non-positive.

Solution is optimal:

any feasible solution satisfies all equations in the tableaux in particular: $Z = 800 - s_c - 2s_b$, $s_c \ge 0$, $s_b \ge 0$ hence optimum solution value is at most 800 the current solution has value 800



Pivoting stops when all coefficients in the objective function are non-positive.

- any feasible solution satisfies all equations in the tableaux
- in particular: $Z = 800 s_c 2s_h, s_c \ge 0, s_h \ge 0$
- hence optimum solution value is at most 800
- the current solution has value 800



Pivoting stops when all coefficients in the objective function are non-positive.

- any feasible solution satisfies all equations in the tableaux
- in particular: $Z = 800 s_c 2s_h$, $s_c \ge 0$, $s_h \ge 0$
- hence optimum solution value is at most 800
- the current solution has value 800



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Pivoting stops when all coefficients in the objective function are non-positive.

- any feasible solution satisfies all equations in the tableaux
- in particular: $Z = 800 s_c 2s_h, s_c \ge 0, s_h \ge 0$
- hence optimum solution value is at most 800
- the current solution has value 800



Let our linear program be

$$\begin{array}{rclcrcrc} c_B^t x_B &+& c_N^t x_N &=& Z\\ A_B x_B &+& A_N x_N &=& b\\ x_B &, & x_N &\geq& 0 \end{array}$$

The simplex tableaux for basis B is

$$(c_N^t - c_B^t A_B^{-1} A_N) x_N = Z - c_B^t A_B^{-1} b$$

$$Ix_B + A_B^{-1} A_N x_N = A_B^{-1} b$$

$$x_B , \qquad x_N \ge 0$$

The BFS is given by $x_N = 0, x_B = A_B^{-1}b$.

If $(c_N^t - c_B^t A_B^{-1} A_N) \le 0$ we know that we have an optimum solution.

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Let our linear program be

$$c_B^t x_B + c_N^t x_N = Z$$

$$A_B x_B + A_N x_N = b$$

$$x_B , \quad x_N \ge 0$$

The simplex tableaux for basis B is

$$(c_N^t - c_B^t A_B^{-1} A_N) x_N = Z - c_B^t A_B^{-1} b$$

$$Ix_B + A_B^{-1} A_N x_N = A_B^{-1} b$$

$$x_B , \qquad x_N \ge 0$$

The BFS is given by $x_N = 0, x_B = A_B^{-1}b$.

If $(c_N^t - c_B^t A_B^{-1} A_N) \le 0$ we know that we have an optimum solution.



4 Simplex Algorithm

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Let our linear program be

$$c_B^t x_B + c_N^t x_N = Z$$

$$A_B x_B + A_N x_N = b$$

$$x_B , \quad x_N \ge 0$$

The simplex tableaux for basis B is

$$(c_N^t - c_B^t A_B^{-1} A_N) x_N = Z - c_B^t A_B^{-1} b$$

$$Ix_B + A_B^{-1} A_N x_N = A_B^{-1} b$$

$$x_B , x_N \ge 0$$

The BFS is given by $x_N = 0, x_B = A_B^{-1}b$.

If $(c_N^t - c_B^t A_B^{-1} A_N) \le 0$ we know that we have an optimum solution.



4 Simplex Algorithm

Let our linear program be

$$c_B^t x_B + c_N^t x_N = Z$$

$$A_B x_B + A_N x_N = b$$

$$x_B , \quad x_N \ge 0$$

The simplex tableaux for basis B is

$$(c_N^t - c_B^t A_B^{-1} A_N) x_N = Z - c_B^t A_B^{-1} b$$

$$Ix_B + A_B^{-1} A_N x_N = A_B^{-1} b$$

$$x_B , \qquad x_N \ge 0$$

The BFS is given by $x_N = 0, x_B = A_B^{-1}b$.

If $(c_N^t - c_B^t A_B^{-1} A_N) \le 0$ we know that we have an optimum solution.



Geometric View of Pivoting



Geometric View of Pivoting














• Given basis *B* with BFS x^* .

- Choose index $j \notin B$ in order to increase x_j^* from 0 to $\theta > 0$. Other non-basis variables should star at 0. Hasis variables change to maintain feasibility.
- Go from x^* to $x^* + \theta \cdot d$.

- $d_j = 1$ (normalization)
- $\ell = 0, \, \ell \in B, \, \ell \neq j$
- $A(x^* + \partial d) = b$ must hold. Hence Ad = 0.
- Altogether: $A_n d_n + A_{n,j} = Ad = 0$, which gives $d_n = -A_n^{-1}A_{n,j}$.



- Given basis *B* with BFS x^* .
- Choose index $j \notin B$ in order to increase x_i^* from 0 to $\theta > 0$.
 - Other non-basis variables should stay at 0.
 - Basis variables change to maintain feasibility.
- Go from x^* to $x^* + \theta \cdot d$.

- $d_f=1$ (normalization)
- $d_l = 0, l \in B, l \neq j$
- $A(x^* + \partial d) = b$ must hold. Hence Ad = 0.
- Altogether: $A_n d_n + A_{n,j} = Ad = 0$, which gives $d_n = -A_n^{-1}A_{n,j}$.



- Given basis B with BFS x^* .
- Choose index $j \notin B$ in order to increase x_i^* from 0 to $\theta > 0$.
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Definition 2 (*j***-th basis direction)**

Let *B* be a basis, and let $j \notin B$. The vector *d* with $d_j = 1$ and $d_{\ell} = 0, \ell \notin B, \ell \neq j$ and $d_B = -A_B^{-1}A_{*j}$ is called the *j*-th basis direction for *B*.

Going from x^* to $x^* + \theta \cdot d$ the objective function changes by

$$\theta \cdot c^t d = \theta (c_j - c_B^t A_B^{-1} A_{*j})$$



4 Simplex Algorithm

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4 Simplex Algorithm

Definition 3 (Reduced Cost)

For a basis *B* the value

$$\tilde{c}_j = c_j - c_B^t A_B^{-1} A_{*j}$$

is called the reduced cost for variable x_j .

Note that this is defined for every j. If $j \in B$ then the above term is 0.



Let our linear program be

$$\begin{array}{rclcrcrc} c_B^t x_B &+& c_N^t x_N &=& Z\\ A_B x_B &+& A_N x_N &=& b\\ x_B & , & x_N &\geq & 0 \end{array}$$

The simplex tableaux for basis *B* is

$$\begin{array}{rcl} (c_{N}^{t}-c_{B}^{t}A_{B}^{-1}A_{N})x_{N} &=& Z-c_{B}^{t}A_{B}^{-1}b\\ Ix_{B} &+& A_{B}^{-1}A_{N}x_{N} &=& A_{B}^{-1}b\\ x_{B} &, & x_{N} &\geq& 0 \end{array}$$

The BFS is given by $x_N = 0, x_B = A_B^{-1}b$.

If $(c_N^t - c_B^t A_B^{-1} A_N) \le 0$ we know that we have an optimum solution.

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 How do we find the initial basic feasible solution?
- Is there always a basis B such that

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- Then we can terminate because we know that the solution is optimal.
- If yes how do we make sure that we reach such a basis?



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For this one computes b_i/A_{ie} for all constraints i and calculates the minimum positive value.

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Because a variable x_{ℓ} with $\ell \in B$ is already 0.

The set of inequalities is degenerate (also the basis is degenerate).

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A BFS x^* is called degenerate if the set $J = \{j \mid x_j^* > 0\}$ fulfills |J| < m.



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Non Degenerate Example





















- ► We can choose a column *e* as an entering variable if *c̃_e* > 0 (*c̃_e* is reduced cost for *x_e*).
- The standard choice is the column that maximizes \tilde{c}_e .
- If $A_{ie} \leq 0$ for all $i \in \{1, ..., m\}$ then the maximum is not bounded.
- ► Otw. choose a leaving variable *l* such that b_l/A_{le} is minimal among all variables *i* with A_{ie} > 0.
- ► If several variables have minimum b_ℓ/A_{ℓe} you reach a degenerate basis.
- Depending on the choice of *l* it may happen that the algorithm runs into a cycle where it does not escape from a degenerate vertex.



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What do we have so far?

Suppose we are given an initial feasible solution to an LP. If the LP is non-degenerate then Simplex will terminate.

Note that we either terminate because the min-ratio test fails and we can conclude that the LP is <u>unbounded</u>, or we terminate because the vector of reduced cost is non-positive. In the latter case we have an <u>optimum solution</u>.



• $Ax \le b, x \ge 0$, and $b \ge 0$.

- ► The standard slack from for this problem is $Ax + Is = b, x \ge 0, s \ge 0$, where *s* denotes the vector of slack variables.
- Then s = b, x = 0 is a basic feasible solution (how?).
- We directly can start the simplex algorithm.



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- Multiply all rows with $b_f < 0$ by -1.
- $(z_1, maximize \rightarrow \sum_i v_i \text{ s.t. } Ax + I = b_i, x > 0, v > 0$ using Simplex. x = 0, v = b is initial feasible.
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- Otw. you have $x \ge 0$ with Ax = b.
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- **1.** Multiply all rows with $b_i < 0$ by -1.
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- **3.** If $\sum_i v_i > 0$ then the original problem is infeasible.
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Optimality

Lemma 5

Let B be a basis and x^* a BFS corresponding to basis B. $\tilde{c} \le 0$ implies that x^* is an optimum solution to the LP.

