# Linear Programming, Lecture 4 

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October 1, 2018

## Standard Form

For an LP, "Standard Form" is usually defined as

$$
\text { Maximize } \mathbf{c}^{T} \mathbf{x} \text { subject to } A \mathbf{x} \leq \mathbf{b}
$$

or
Maximize $\mathbf{c}^{T} \mathbf{x}$ subject to $A \mathbf{x} \leq \mathbf{b}, \mathbf{x} \geq \mathbf{0}$.
or
Maximize $\mathbf{c}^{T} \mathbf{x}$ subject to $A \mathbf{x}=\mathbf{b}, \mathbf{x} \geq \mathbf{0}$.

## Simplex Method

To run the simplex method, we start from a Linear Program (LP) in the following standard simplex form.

$$
\begin{aligned}
& \operatorname{Max} \quad z \\
& \text { s.t. }(-z)+a_{01} x_{1}+\cdots+a_{0 n} x_{n}=b_{0} \\
& a_{11} x_{1}+\cdots+a_{1 n} x_{n}=b_{1} \\
& a_{m 1} x_{1}+\cdots+a_{m n} x_{n}=b_{m} \\
& x_{i} \geq 0
\end{aligned}
$$

$$
\begin{aligned}
& \text { Max } \begin{aligned}
z & \\
\text { s.t. } \quad(-z)+a_{01} x_{1}+\cdots+a_{0 n} x_{n} & =b_{0} \\
a_{11} x_{1}+\cdots+a_{1 n} x_{n} & =b_{1} \\
\vdots & \\
& \\
a_{m 1} x_{1}+\cdots+a_{m n} x_{n} & =b_{m} \\
x_{i} & \geq 0
\end{aligned}
\end{aligned}
$$

To be in standard simplex form:

1. All decision variables $x_{i}$ (except $-z$ ) are non-negative.
2. All other constraints are equalities.
3. The RHS (except for the "cost row" or " $z$-row") is non-negative.
4. For each row $i$, there is a column equal to $e_{i}$ (a 1 in row $i$, and 0 in all other rows).

## Remarks

- There are multiple conventions as to what constitutes "Standard Form." They are all different, but more or less equivalent in terms of requirements.
- Given an LP in the form: Max $z$, subject to inequalities all of the form $\sum a_{i} x_{i} \leq b$, one only needs to introduce slack variables to obtain starting standard form for Simplex Method.
- Today, we will learn techniques for more complicated LPs.


## Example 1

Is the following LP in standard simplex form?
Maximize $z$, subject to $x_{1}, x_{2}, x_{3}, s_{1}, s_{2} \geq 0$ and the equalities:

| $-z$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $s_{1}$ | $s_{2}$ | $=$ | $R H S$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 1 | 9 | 1 | 0 | 0 | 0 |  |
| 0 | 1 | 2 | 3 | 1 | 0 | 9 |  |
| 0 | 3 | 2 | 2 | 0 | 1 |  | 15 |

- Non-negative decision variables?
- Equalities for constrains?
- Non-negative RHS entries?
- Columns $e_{i}$ ? $\checkmark-z, s_{1}, s_{2}$


## Example 2

Is the following LP in standard simplex form?
Maximize $z$, subject to $x_{1}, x_{2}, x_{3}, s_{1}, s_{2} \geq 0$ and the equalities:

| $-z$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $s_{1}$ | $s_{2}$ | $=$ | $R H S$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 1 | 9 | 1 | 0 | 0 | 0 |  |
| 0 | 1 | 2 | 3 | -1 | 0 |  | 9 |
| 0 | 3 | 2 | 2 | 0 | 1 |  | 15 |

- Non-negative decision variables?
- Equalities for constrains?
- Non-negative RHS entries?
- Columns $e_{i}$ ? X No column $\left[\begin{array}{lll}0 & 1 & 0\end{array}\right]^{T}$


## Potential complications

1. Minimizing instead of maximizing.
2. Decision variables allowed to take negative values.
3. Inequalities of form $\sum a_{i} x_{i} \geq b$.

## Maximize/Minimize

Maximizing $f\left(x_{1}, \ldots, x_{n}\right) \Leftrightarrow$ Minimizing $-f\left(x_{1}, \ldots, x_{n}\right)$

Notes:

- To change between max/min, just multiply all coefficients by -1 .
- In practice, some implementations of simplex method assume you are maximizing, and some assume you are minimizing. Max vs Min is a minor detail.
- When doing simplex method by hand, you may simply keep cost row the same as when maximizing, but perform pivots on columns with a negative entry in the cost row.


## Decision variables taking negative values

Bounded below: Suppose we have a variable $x \geq-20$.
Then: Substitute for new variable $x=\hat{x}-20$, or $\hat{x}=x+20$, with $\hat{x} \geq 0$.
Note: For constraint $x \geq 20$, we may introduce surplus variable, or we may use substitution $x=\hat{x}+20$, with $\hat{x} \geq 0$.

Unbounded: Suppose we have unbounded variable $w$.
Then: Use substitution $w=w^{+}-w^{-}$, with $w^{+}, w^{-} \geq 0$.
Note: When more than one decision variable is unrestricted, a single variable $x^{-}$can be used for all of them, with the interpretation that it is the most negative of all the decision variables.

## Put LP in Simplex Form to start Simplex Method

Min

$$
x+y
$$

s.t.

$$
\begin{aligned}
& x+y \leq 20 \\
& x+y \geq-20 \\
& x-y \leq 20 \\
& x-y \geq-20
\end{aligned}
$$

$x, y$ unrestricted

Question: Will the LP have a unique solution?

## Inequalities:

$$
\begin{aligned}
& -20 \leq x+y \leq 20 \\
& -20 \leq x-y \leq 20
\end{aligned}
$$

Substitutions:

$$
\begin{gathered}
x=x^{+}-x^{-} \text {and } y=y^{+}-y^{-} \\
x^{+}, x^{-}, y^{+}, y^{-} \geq 0
\end{gathered}
$$

New inequalities:

$$
\begin{aligned}
& -20 \leq x^{+}-x^{-}+y^{+}-y^{-} \leq 20 \\
& -20 \leq x^{+}-x^{-}-y^{+}+y^{-} \leq 20
\end{aligned}
$$

$\operatorname{Min} x^{+}-x^{-}+y^{+}-y^{-}$
s.t. $\quad x^{+}-x^{-}+y^{+}-y^{-} \leq 20$

$$
x^{+}-x^{-}+y^{+}-y^{-} \geq-20
$$

$$
x^{+}-x^{-}-y^{+}+y^{-} \leq 20
$$

$$
x^{+}-x^{-}-y^{+}+y^{-} \geq-20
$$

$$
x^{+}, x^{-}, y^{+}, y^{-} \geq 0
$$

$\operatorname{Max} z=-x^{+}+x^{-}-y^{-}+y^{-}$
s.t.

$$
x^{+}-x^{-}+y^{+}-y^{-} \leq 20
$$

$$
x^{+}-x^{-}+y^{+}-y^{-} \geq-20
$$

$$
x^{+}-x^{-}-y^{+}+y^{-} \leq 20
$$

$$
x^{+}-x^{-}-y^{+}+y^{-} \geq-20
$$

$$
x^{+}, x^{-}, y^{+}, y^{-} \geq 0
$$

$\operatorname{Max} z=-x^{+}+x^{-}-y^{-}+y^{-}$
s.t. $\quad x^{+}-x^{-}+y^{+}-y^{-}+s_{1}=20$
$x^{+}-x^{-}+y^{+}-y^{-}-s_{2}=-20$
$x^{+}-x^{-}-y^{+}+y^{-}+s_{3}=20$
$x^{+}-x^{-}-y^{+}+y^{-}-s_{4}=-20$
$x^{+}, x^{-}, y^{+}, y^{-} \geq 0$

Maximize z subject to non-negativity constraints and:

$$
\begin{array}{cccccccccc}
(-z) & x^{+} & x^{-} & y^{+} & y^{-} & s_{1} & s_{2} & s_{3} & s_{4} & R H S \\
\hline 1 & -1 & 1 & -1 & 1 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & -1 & 1 & -1 & 1 & 0 & 0 & 0 & 20 \\
0 & 1 & -1 & 1 & -1 & 0 & -1 & 0 & 0 & -20 \\
0 & 1 & -1 & -1 & 1 & 0 & 0 & 1 & 0 & 20 \\
0 & 1 & -1 & -1 & 1 & 0 & 0 & 0 & -1 & -20 \\
& & & & & & & & & \\
(-z) & x^{+} & x^{-} & y^{+} & y^{-} & s_{1} & s_{2} & s_{3} & s_{4} & R H S \\
\hline 1 & -1 & 1 & -1 & 1 & 0 & 0 & 0 & 0 & 0 \\
0 & 1 & -1 & 1 & -1 & 1 & 0 & 0 & 0 & 20 \\
0 & -1 & 1 & -1 & 1 & 0 & 1 & 0 & 0 & 20 \\
0 & 1 & -1 & -1 & 1 & 0 & 0 & 1 & 0 & 20 \\
0 & -1 & 1 & 1 & -1 & 0 & 0 & 0 & 1 & 20
\end{array}
$$

## Alternate Options:

- Since two variables are unbounded, we could substitute $x=x^{+}-t$ and $y=y^{+}-t$, where $x^{+}, y^{+}, t \geq 0$. Here, $t=\max \{-x,-y, 0\}$.
- Inspecting the inequalities, we observe that $x, y \geq-20$. Hence, we could use substitutions

$$
x=\hat{x}-20, \quad y=\hat{y}-20, \quad \hat{x}, \hat{y} \geq 0
$$

## $\geq$ Inequalities

Given: $\sum_{j} a_{i j} x_{j} \geq b_{i}$, with $b_{i}>0$
Introduce: surplus variable $s \geq 0$ to form

$$
\sum_{j} a_{i j} x_{j}-s_{i}=b_{i}
$$

Problem: Neither $\sum_{j} a_{i j} x_{j}-s_{i}=b_{i}$ nor $\sum_{j}-a_{i j} x_{j}+s_{i}=-b_{i}$ give something in simplex form.
Solution: Introduce artificial variable(s) $\alpha_{i} \geq 0$, $\sum_{j} a_{i j} x_{j}-s_{i}+\alpha_{i}=b_{i}$. Then, solve the related LP with objective function $\sum \alpha_{i}$, and same constraints. This produces an initial Basic Feasible Solution and immediately translates to the LP in simplex form.

## Example

$$
\begin{array}{lrl}
\text { Max } & 2 x_{1}+x_{2} & \\
\text { s.t. } & x_{1}+x_{2} & \leq 3 \\
-x_{1}+x_{2} & \geq 1 \\
& x_{1}, x_{2} & \geq 0
\end{array}
$$



## Phase I auxiliary LP:

Min $\quad x_{5}$

$$
\text { s.t. } \begin{aligned}
x_{1}+x_{2}+x_{3} & =3 \\
-x_{1}+x_{2}-x_{4}+x_{5} & =1 \\
x_{1}, \ldots x_{5} & \geq 0
\end{aligned}
$$

| $-w$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $x_{5}$ | $R H S$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 0 | 0 | 0 | 0 | -1 | 0 |
| 0 | 1 | 1 | 1 | 0 | 0 | 3 |
| 0 | -1 | 1 | 0 | -1 | 1 | 1 |


| $-w$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $x_{5}$ | $R H S$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | -1 | 1 | 0 | -1 | 0 | 1 |
| 0 | 1 | 1 | 1 | 0 | 0 | 3 |
| 0 | -1 | 1 | 0 | -1 | 1 | 1 |

Solution to Phase I auxiliary LP:

| $-w$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $x_{5}$ | $R H S$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 0 | 0 | 0 | 0 | -1 | 0 |
| 0 | 2 | 0 | 1 | 1 | -1 | 2 |
| 0 | -1 | 1 | 0 | -1 | 1 | 1 |

Solution of $x_{1}=0, x_{2}=1, x_{3}=2, x_{4}=0$ gives $x_{5}=0$. These values $\left(x_{1}, \ldots, x_{4}\right)$ satisfy the original LPs constraints, and they form our initial Basic Feasible Solution!

Phase II, Solving initial LP:

| $-z$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $R H S$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 2 | 1 | 0 | 0 | 0 |
| 0 | 2 | 0 | 1 | 1 | 2 |
| 0 | -1 | 1 | 0 | -1 | 1 |

ERO's to isolate $x_{2}, x_{3}$, then in simplex form.

Suppose that when performing the simplex method, you obtain column with positive number in objective row, and non-positive numbers in rest of column. Then, the feasible region is unbounded, and a solution does not exist.

Example:

| $-z$ | $x_{1}$ | $x_{2}$ | $x_{3}$ | $x_{4}$ | $R H S$ |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 2 | 0 | 0 | 0 | 0 |
| 0 | -2 | 0 | 1 | 1 | 2 |
| 0 | -1 | 1 | 0 | -1 | 1 |

For basic solution, we let $x_{1}=0$. But, we can let $x_{1}$ be any positive number, and we obtain a better feasible solution.

## Cycling

When several iterations of the simplex method do not improve the current objective value, this is called stalling.

When, after several iterations, the simplex method returns a previous tableau, this is called cycling.

In general, stalling and cycling can occur. Some implementations of the simplex method include special provisions to prevent cycling; other implementations do not try to prevent cycling, and instead rely on small rounding errors to eventually move off the cycle.

## Bland's Rule:

1. Select the first column with positive coefficient in $Z$-row.
2. If there is a tie in Min-Ratio test, choose the first row within the tie.

Theorem: Following Bland's Rule, the simplex method will always terminate in a finite number of steps.

