## Randomised Algorithms

Lecture 6: Linear Programming: Introduction

Thomas Sauerwald (tms41@cam.ac.uk)


## Outline

## Introduction

## A Simple Example of a Linear Program

Formulating Problems as Linear Programs

## Standard and Slack Forms




- linear programming is a powerful tool in optimisation
- inspired more sophisticated techniques such as quadratic optimisation, convex optimisation, integer programming and semi-definite programming
- we will later use the connection between linear and integer programming to tackle several problems (Vertex-Cover, Set-Cover, TSP, satisfiability)


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## What are Linear Programs?

Linear Programming (informal definition)

- maximise or minimise an objective, given limited resources (competing constraint)
- constraints are specified as (in)equalities
- objective function and constraints are linear


## A Simple Example of a Linear Optimisation Problem

- Laptop


## A Simple Example of a Linear Optimisation Problem

- Laptop
- selling price to retailer: 1,000 GBP


## A Simple Example of a Linear Optimisation Problem

- Laptop
- selling price to retailer: 1,000 GBP
- glass: 4 units


## A Simple Example of a Linear Optimisation Problem

- Laptop
- selling price to retailer: 1,000 GBP
- glass: 4 units
- copper: 2 units


## A Simple Example of a Linear Optimisation Problem

- Laptop
- selling price to retailer: 1,000 GBP
- glass: 4 units
- copper: 2 units
- rare-earth elements: 1 unit


## A Simple Example of a Linear Optimisation Problem

- Laptop
- selling price to retailer: 1,000 GBP
- glass: 4 units
- copper: 2 units
- rare-earth elements: 1 unit
- Smartphone


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- Smartphone
- selling price to retailer: 1,000 GBP
- glass: 1 unit


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- You have a daily supply of:


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- copper: 1 unit
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- You have a daily supply of:
- glass: 20 units


## A Simple Example of a Linear Optimisation Problem

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- copper: 2 units
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- Smartphone
- selling price to retailer: 1,000 GBP
- glass: 1 unit
- copper: 1 unit
- rare-earth elements: 2 units
- You have a daily supply of:

- glass: 20 units
- copper: 10 units


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- copper: 2 units
- rare-earth elements: 1 unit
- Smartphone
- selling price to retailer: 1,000 GBP
- glass: 1 unit
- copper: 1 unit
- rare-earth elements: 2 units
- You have a daily supply of:
- glass: 20 units
- copper: 10 units
- rare-earth elements: 14 units


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- glass: 20 units
- copper: 10 units
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- (and enough of everything else...)



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- glass: 20 units
- copper: 10 units
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- (and enough of everything else...)


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How to maximise your daily earnings?

The Linear Program
Linear Program for the Production Problem
maximise $\quad x_{1}+x_{2}$
subject to

$$
\begin{aligned}
& 4 x_{1}+x_{2} \leq 20 \\
& 2 x_{1}+ x_{2} \\
& \leq 10 \\
& x_{1}+2 \leq x_{2} \\
& x_{1}, x_{2} \\
& \geq 14 \\
&
\end{aligned}
$$

## The Linear Program

## Linear Program for the Production Problem

maximise $\quad x_{1}+x_{2}$
subject to

| $4 x_{1}+x_{2}$ | $\leq 20$ |
| ---: | :--- |
| $2 x_{1}+$ | $x_{2}$ |
| $\leq$ | 10 |
| $x_{1}+2 x_{2}$ | $\leq 14$ |
| $x_{1}, x_{2}$ |  |
| $\geq$ |  |

The solution of this linear program yields the optimal production schedule.

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Formal Definition of Linear Program

## The Linear Program

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The solution of this linear program yields the optimal production schedule.
Formal Definition of Linear Program

- Given $a_{1}, a_{2}, \ldots, a_{n}$ and a set of variables $x_{1}, x_{2}, \ldots, x_{n}$, a linear function $f$ is defined by

$$
f\left(x_{1}, x_{2}, \ldots, x_{n}\right)=a_{1} x_{1}+a_{2} x_{2}+\cdots+a_{n} x_{n} .
$$

## The Linear Program

## Linear Program for the Production Problem

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- Linear Equality: $f\left(x_{1}, x_{2}, \ldots, x_{n}\right)=b$
- Linear Inequality: $f\left(x_{1}, x_{2}, \ldots, x_{n}\right) \geq b$


## The Linear Program

## Linear Program for the Production Problem

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| ---: | :--- |
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## Linear Constraints

- Linear-Progamming Problem: either minimise or maximise a linear function subject to a set of linear constraints


# Finding the Optimal Production Schedule 

| maximise | $x_{1}$ | + | $x_{2}$ |  |
| :--- | :---: | :--- | :--- | :--- |
| subject to |  |  |  |  |
|  | $4 x_{1}$ | + | $x_{2}$ | $\leq 20$ |
|  | $2 x_{1}+$ | $x_{2}$ | $\leq 10$ |  |
|  | $x_{1}+$ | $2 x_{2}$ | $\leq 14$ |  |
|  | $x_{1}, x_{2}$ |  | $\geq 0$ |  |

## Finding the Optimal Production Schedule

| $\operatorname{maximise}$ | $x_{1}$ | + | $x_{2}$ |
| :--- | :---: | :---: | :---: |
| subject to |  |  |  |$c$

Any setting of $x_{1}$ and $x_{2}$ satisfying all constraints is a feasible solution

Finding the Optimal Production Schedule


Finding the Optimal Production Schedule


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## Finding the Optimal Production Schedule



Question: Which aspect did we ignore in the formulation of the linear program?

## Finding the Optimal Production Schedule



While the same approach also works for higher-dimensions, we need to take a more systematic and algebraic procedure.

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## Shortest Paths

## Single-Pair Shortest Path Problem

- Given: directed graph $G=(V, E)$ with edge weights $w: E \rightarrow \mathbb{R}$, pair of vertices $s, t \in V$



## Shortest Paths

## Single-Pair Shortest Path Problem

- Given: directed graph $G=(V, E)$ with edge weights $w: E \rightarrow \mathbb{R}$, pair of vertices $s, t \in V$
- Goal: Find a path of minimum weight from $s$ to $t$ in $G$



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- Goal: Find a path of minimum weight from $s$ to $t$ in $G$

$$
\begin{aligned}
& p=\left(v_{0}=s, v_{1}, \ldots, v_{k}=t\right) \text { such that } \\
& w(p)=\sum_{i=1}^{k} w\left(v_{k-1}, v_{k}\right) \text { is minimised. }
\end{aligned}
$$



## Shortest Paths

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Exercise: How can we translate the SPSP problem into a linear program?

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Shortest Paths as LP
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Shortest Paths as LP
subject to

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\begin{aligned}
& d_{v} \leq d_{u}+w(u, v) \quad \text { for each edge }(u, v) \in E \\
& d_{s}=0
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Shortest Paths as LP
maximise $\quad d_{t}$
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## Maximum Flow

Maximum Flow Problem

- Given: directed graph $G=(V, E)$ with edge capacities $c: E \rightarrow \mathbb{R}^{+}$ (recall $c(u, v)=0$ if $(u, v) \notin E)$, pair of vertices $s, t \in V$


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Maximum Flow as LP
maximise $\quad \sum_{v \in V} f_{s v}-\sum_{v \in V} f_{v s}$
subject to

$$
\begin{array}{rlrl}
f_{u v} & \leq & c(u, v) & \\
\text { for each } u, v \in V \\
\sum_{v \in V} f_{v u} & = & \sum_{v \in V} f_{u v} & \text { for each } u \in V \backslash\{s, t\} \\
f_{u v} & \geq & 0 & \text { for each } u, v \in V .
\end{array}
$$

## Minimum-Cost Flow

## Extension of the Maximum Flow Problem

Minimum-Cost-Flow Problem

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- Given: directed graph $G=(V, E)$ with capacities $c: E \rightarrow \mathbb{R}^{+}$, pair of vertices $s, t \in V$, cost function $a: E \rightarrow \mathbb{R}^{+}$, flow demand of $d$ units


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- Goal: Find a flow $f: V \times V \rightarrow \mathbb{R}$ from $s$ to $t$ with $|f|=d$ while minimising the total cost $\sum_{(u, v) \in E} a(u, v) f_{u v}$ incurrred by the flow.


## Extension of the Maximum Flow Problem

## Minimum-Cost-Flow Problem

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(a)

(b)

Figure 29.3 (a) An example of a minimum-cost-flow problem. We denote the capacities by $c$ and the costs by $a$. Vertex $s$ is the source and vertex $t$ is the sink, and we wish to send 4 units of flow from $s$ to $t$. (b) A solution to the minimum-cost flow problem in which 4 units of flow are sent from $s$ to $t$. For each edge, the flow and capacity are written as flow/capacity.

## Extension of the Maximum Flow Problem

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## Optimal Solution with total cost:

$$
\sum_{(u, v) \in E} a(u, v) f_{u v}=(2 \cdot 2)+(5 \cdot 2)+(3 \cdot 1)+(7 \cdot 1)+(1 \cdot 3)=27
$$


(a)

(b)

Figure 29.3 (a) An example of a minimum-cost-flow problem. We denote the capacities by $c$ and the costs by $a$. Vertex $s$ is the source and vertex $t$ is the sink, and we wish to send 4 units of flow from $s$ to $t$. (b) A solution to the minimum-cost flow problem in which 4 units of flow are sent from $s$ to $t$. For each edge, the flow and capacity are written as flow/capacity.

## Minimum Cost Flow as a LP

## Minimum Cost Flow as LP

minimise

$$
\sum_{(u, v) \in E} a(u, v) f_{u v}
$$

subject to

$$
\begin{aligned}
f_{u v} & \leq c(u, v) & & \text { for } u, v \in V \\
\sum_{v \in V} f_{v u}-\sum_{v \in V} f_{u v} & =0 & & \text { for } u \in V \backslash\{s, t\}, \\
\sum_{v \in V} f_{s v}-\sum_{v \in V} f_{v s} & =d, & & \\
f_{u v} & \geq 0 & & \text { for } u, v \in V .
\end{aligned}
$$

## Minimum Cost Flow as a LP

$$
\begin{aligned}
& \text { — Minimum Cost Flow as LP } \\
& \begin{array}{lrlrl}
\text { minimise } \\
\text { subject to }
\end{array} \\
& \qquad \begin{array}{rlrl}
\sum_{(u, v) \in E} a(u, v) f_{u v} & & \\
f_{u v} & \leq c(u, v) & & \text { for } u, v \in V, \\
\sum_{v \in V} f_{v u}-\sum_{v \in V} f_{u v} & =0 & & \text { for } u \in V \backslash\{s, t\}, \\
\sum_{v \in V} f_{s v}-\sum_{v \in V} f_{v s} & =d, & & \\
f_{u v} & \geq 0 & & \text { for } u, v \in V .
\end{array}
\end{aligned}
$$

Real power of Linear Programming comes from the ability to solve new problems!

## Outline

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Standard and Slack Forms

## Standard and Slack Forms

## Standard Form

$$
\begin{array}{ll}
\text { maximise } & \sum_{j=1}^{n} c_{j} x_{j} \\
\text { subject to } \\
& \sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i} \\
& \text { for } i=1,2, \ldots, m \\
x_{j} \geq 0 & \text { for } j=1,2, \ldots, n
\end{array}
$$

## Standard and Slack Forms

## Standard Form

$$
\begin{array}{ll}
\text { maximise } & \sum_{j=1}^{n} c_{j} x_{j}<\underbrace{}_{\text {Objective Function }} \\
\text { subject to } \\
& \sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i} \quad \text { for } i=1,2, \ldots, m \\
& x_{j} \geq 0 \quad \text { for } j=1,2, \ldots, n
\end{array}
$$

## Standard and Slack Forms



## Standard and Slack Forms



## Standard and Slack Forms



## Converting Linear Programs into Standard Form

## Reasons for a LP not being in standard form:

1. The objective might be a minimisation rather than maximisation.
2. There might be variables without nonnegativity constraints.
3. There might be equality constraints.
4. There might be inequality constraints (with $\geq$ instead of $\leq$ ).

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Goal: Convert linear program into an equivalent program which is in standard form

## Converting Linear Programs into Standard Form

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2. There might be variables without nonnegativity constraints.
3. There might be equality constraints.
4. There might be inequality constraints (with $\geq$ instead of $\leq$ ).

Goal: Convert linear program into an equivalent program which is in standard form


Equivalence: a correspondence (not necessarily a bijection) between solutions.

## Converting into Standard Form (1/5)

## Reasons for a LP not being in standard form:

1. The objective might be a minimisation rather than maximisation.

## Converting into Standard Form (1/5)

## Reasons for a LP not being in standard form:

1. The objective might be a minimisation rather than maximisation.


## Converting into Standard Form (1/5)

## Reasons for a LP not being in standard form:

1. The objective might be a minimisation rather than maximisation.

| minimise | $-2 x_{1}$ | + |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| subject to |  |  |  |  |  |
|  | $x_{1}$ | + | $x_{2}$ | $=$ | 7 |
|  | $\chi_{1}$ | - | $2 x_{2}$ | $\leq$ | 4 |
|  | $\chi_{1}$ |  |  | $\geq$ | 0 |

## Converting into Standard Form (1/5)

## Reasons for a LP not being in standard form:

1. The objective might be a minimisation rather than maximisation.

| minimise | $-2 x_{1}$ | + |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| subject to |  |  |  |  |  |
|  | $x_{1}$ | + | $x_{2}$ | $=$ | 7 |
|  | $x_{1}$ | - | $2 x_{2}$ | $\leq$ | 4 |
|  | $x_{1}$ |  |  | $\geq$ | 0 |
|  |  | N | Negate objective function |  | ve function |
| maximise | $2 x_{1}$ | - | $3 x_{2}$ |  |  |
| subject to |  |  |  |  |  |
|  | $x_{1}$ | + | $x_{2}$ | $=$ | 7 |
|  | $x_{1}$ | - | $2 x_{2}$ | $\leq$ | 4 |
|  | $x_{1}$ |  |  |  | 0 |

Converting into Standard Form (2/5)

## Reasons for a LP not being in standard form:

2. There might be variables without nonnegativity constraints.

## Converting into Standard Form (2/5)

## Reasons for a LP not being in standard form:

2. There might be variables without nonnegativity constraints.

$$
\begin{array}{lcl}
\operatorname{maximise} & 2 x_{1}-3 x_{2} & \\
\text { subject to } & & \\
& x_{1}+2 x_{2}=7 \\
& x_{1}-2 x_{2} \leq 4 \\
& x_{1} & \\
& \geq 0 \\
\hline
\end{array}
$$

## Converting into Standard Form (2/5)

## Reasons for a LP not being in standard form:

2. There might be variables without nonnegativity constraints.

$$
\begin{aligned}
& \text { maximise } 2 x_{1}-3 x_{2} \\
& \text { subject to } \\
& \text { Replace } x_{2} \text { by two non-negative } \\
& \text { variables } x_{2}^{\prime} \text { and } x_{2}^{\prime \prime}
\end{aligned}
$$

## Converting into Standard Form (2/5)

## Reasons for a LP not being in standard form:

2. There might be variables without nonnegativity constraints.
maximise
subject to


Replace $x_{2}$ by two non-negative variables $x_{2}^{\prime}$ and $x_{2}^{\prime \prime}$
maximise $2 x_{1}-3 x_{2}^{\prime}+3 x_{2}^{\prime \prime}$
subject to


## Converting into Standard Form (3/5)

## Reasons for a LP not being in standard form:

3. There might be equality constraints.

## Converting into Standard Form (3/5)

## Reasons for a LP not being in standard form:

3. There might be equality constraints.

$$
\begin{array}{lccccc}
\begin{array}{l}
\operatorname{maximise} \\
\text { subject to }
\end{array} & 2 x_{1}-3 x_{2}^{\prime}+3 x_{2}^{\prime \prime} & \\
& \begin{array}{|ccrrr}
x_{1} & + & x_{2}^{\prime} & - & x_{2}^{\prime \prime} \\
& x_{1}-2 & 2 x_{2}^{\prime} & + & 2 x_{2}^{\prime \prime} \\
& x_{1}, x_{2}^{\prime}, x_{2}^{\prime \prime} & & & \geq \\
& & & \geq
\end{array}
\end{array}
$$

## Converting into Standard Form (3/5)

## Reasons for a LP not being in standard form:

3. There might be equality constraints.
maximise $2 x_{1}-3 x_{2}^{\prime}+3 x_{2}^{\prime \prime}$
subject to


Replace each equality by two inequalities.

## Converting into Standard Form (3/5)

## Reasons for a LP not being in standard form:

3. There might be equality constraints.


Converting into Standard Form (4/5)

Reasons for a LP not being in standard form:
4. There might be inequality constraints (with $\geq$ instead of $\leq$ ).

## Converting into Standard Form (4/5)

## Reasons for a LP not being in standard form:

4. There might be inequality constraints (with $\geq$ instead of $\leq$ ).
maximise $2 x_{1}-3 x_{2}^{\prime}+3 x_{2}^{\prime \prime}$
subject to

| $x_{1}+$ | $x_{2}^{\prime}$ | - | $x_{2}^{\prime \prime}$ | $\leq$ | 7 |
| :---: | :---: | ---: | :---: | :---: | :---: |
| $x_{1}+$ | $x_{2}^{\prime}$ | - | $x_{2}^{\prime \prime}$ | $\geq$ | 7 |
| $x_{1}-22 x_{2}^{\prime}$ | $+2 x_{2}^{\prime \prime}$ | $\leq$ | 4 |  |  |
| $x_{1}, x_{2}^{\prime}, x_{2}^{\prime \prime}$ |  |  | $\geq$ | 0 |  |

## Converting into Standard Form (4/5)

## Reasons for a LP not being in standard form:

4. There might be inequality constraints (with $\geq$ instead of $\leq$ ).

$$
\begin{aligned}
& \text { maximise } 2 x_{1}-3 x_{2}^{\prime}+3 x_{2}^{\prime \prime} \\
& \text { subject to } \\
& \text { Negate respective inequalities. }
\end{aligned}
$$

## Converting into Standard Form (4/5)

## Reasons for a LP not being in standard form:

4. There might be inequality constraints (with $\geq$ instead of $\leq$ ).


Converting into Standard Form (5/5)

| maximise | $2 x_{1}$ | - | $3 x_{2}$ | + | $3 x_{3}$ |  |  |
| :--- | ---: | :--- | ---: | :--- | ---: | :--- | ---: |
| subject to |  |  |  |  |  |  |  |
|  | $x_{1}$ | + | $x_{2}$ | - | $x_{3}$ | $\leq$ | 7 |
|  | $-x_{1}$ | - | $x_{2}$ | + | $x_{3}$ | $\leq$ | -7 |
|  | $x_{1}$ | - | $2 x_{2}$ | + | $2 x_{3}$ | $\leq$ | 4 |
|  | $x_{1}, x_{2}, x_{3}$ |  |  | $\geq$ | 0 |  |  |

Converting into Standard Form (5/5)


## Converting into Standard Form (5/5)



It is always possible to convert a linear program into standard form.

## Converting Standard Form into Slack Form (1/3)

Goal: Convert standard form into slack form, where all constraints except for the non-negativity constraints are equalities.

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> For the simplex algorithm, it is more convenient to work with equality constraints.

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Goal: Convert standard form into slack form, where all constraints except for the non-negativity constraints are equalities.

For the simplex algorithm, it is more convenient to work with equality constraints.

Introducing Slack Variables

## Converting Standard Form into Slack Form (1/3)

Goal: Convert standard form into slack form, where all constraints except for the non-negativity constraints are equalities.

For the simplex algorithm, it is more convenient to work with equality constraints.

Introducing Slack Variables

- Let $\sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i}$ be an inequality constraint


## Converting Standard Form into Slack Form (1/3)

Goal: Convert standard form into slack form, where all constraints except for the non-negativity constraints are equalities.

For the simplex algorithm, it is more convenient to work with equality constraints.

Introducing Slack Variables

- Let $\sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i}$ be an inequality constraint
- Introduce a slack variable $s$ by


## Converting Standard Form into Slack Form (1/3)

Goal: Convert standard form into slack form, where all constraints except for the non-negativity constraints are equalities.

For the simplex algorithm, it is more convenient to work with equality constraints.

Introducing Slack Variables

- Let $\sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i}$ be an inequality constraint
- Introduce a slack variable $s$ by

$$
s=b_{i}-\sum_{j=1}^{n} a_{i j} x_{j}
$$

## Converting Standard Form into Slack Form (1/3)

Goal: Convert standard form into slack form, where all constraints except for the non-negativity constraints are equalities.

For the simplex algorithm, it is more convenient to work with equality constraints.

Introducing Slack Variables

- Let $\sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i}$ be an inequality constraint
- Introduce a slack variable $s$ by

$$
\begin{aligned}
& s=b_{i}-\sum_{j=1}^{n} a_{i j} x_{j} \\
& s \geq 0
\end{aligned}
$$

## Converting Standard Form into Slack Form (1/3)

Goal: Convert standard form into slack form, where all constraints except for the non-negativity constraints are equalities.

For the simplex algorithm, it is more convenient to work with equality constraints.

Introducing Slack Variables

- Let $\sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i}$ be an inequality constraint
- Introduce a slack variable $s$ by
$s$ measures the slack between the two sides of the inequality.

$$
\begin{aligned}
& s=b_{i}-\sum_{j=1}^{n} a_{i j} x_{j} \\
& s \geq 0
\end{aligned}
$$

## Converting Standard Form into Slack Form (1/3)

Goal: Convert standard form into slack form, where all constraints except for the non-negativity constraints are equalities.

For the simplex algorithm, it is more convenient to work with equality constraints.

Introducing Slack Variables

- Let $\sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i}$ be an inequality constraint
- Introduce a slack variable $s$ by
$s$ measures the slack between the two sides of the inequality.

$$
\left\{\begin{array}{l}
s=b_{i}-\sum_{j=1}^{n} a_{i j} x_{j} \\
s \geq 0
\end{array}\right.
$$

- Denote slack variable of the $i$-th inequality by $x_{n+i}$

Converting Standard Form into Slack Form (2/3)

$$
\begin{array}{lclllll}
\operatorname{maximise} & 2 x_{1} & - & 3 x_{2} & + & 3 x_{3} & \\
\text { subject to } & & & & & & \\
& x_{1} & + & x_{2} & - & x_{3} & \leq \\
& -x_{1} & - & x_{2} & + & x_{3} & \leq \\
& x_{1} & -7 \\
& x_{1}, x_{2} & + & 2 x_{3} & \leq & 4 \\
& x_{1}, x_{3} & & & \geq & 0
\end{array}
$$

## Converting Standard Form into Slack Form (2/3)

| maximise | $2 x_{1}$ | - | $3 x_{2}$ | + | $3 x_{3}$ |  |  |
| :--- | ---: | :--- | ---: | :--- | :--- | :--- | :--- |
| subject to |  |  |  |  |  |  |  |
|  | $x_{1}$ | + | $x_{2}$ | - | $x_{3}$ | $\leq$ | 7 |
|  | $-x_{1}$ | - | $x_{2}$ | + | $x_{3}$ | $\leq$ | -7 |
|  | $x_{1}$ | - | $2 x_{2}$ | + | $2 x_{3}$ | $\leq$ | 4 |
|  | $x_{1}, x_{2}, x_{3}$ |  |  |  | $\geq$ | 0 |  |
|  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |
|  |  |  |  |  |  |  |  |

## Converting Standard Form into Slack Form (2/3)


subject to

$$
x_{4}=7-x_{1}-x_{2}+x_{3}
$$

## Converting Standard Form into Slack Form (2/3)


subject to

$$
\begin{array}{rlrllllll}
x_{4} & = & 7 & - & x_{1} & - & x_{2} & + & x_{3} \\
x_{5} & = & -7 & + & x_{1} & + & x_{2} & - & x_{3}
\end{array}
$$

## Converting Standard Form into Slack Form (2/3)

| maximise | $2 x_{1}$ | - | $3 x_{2}$ | $+$ | $3 x_{3}$ |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $x_{1}$ | + | $x_{2}$ | - | $x_{3}$ | $\leq$ | 7 |
|  | $-x_{1}$ | - | $\chi_{2}$ | $+$ | $x_{3}$ | $\leq$ | -7 |
|  | $x_{1}$ | - | $2 x_{2}$ | + | $2 x_{3}$ | $\leq$ | 4 |
|  |  | X2 |  |  |  | $\geq$ | 0 |
|  |  |  | $\downarrow$ | trod | ice s | ck | riabl |

subject to

$$
\begin{array}{rrrrrlrlr}
x_{4} & = & 7 & - & x_{1} & - & x_{2} & + & x_{3} \\
x_{5} & = & -7 & + & x_{1} & + & x_{2} & - & x_{3} \\
x_{6} & = & 4 & - & x_{1} & + & 2 x_{2} & - & 2 x_{3}
\end{array}
$$

| maximise subject to | $2 x_{1}$ | - | $3 x_{2}$ | + | $3 x_{3}$ |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $x_{1}$ | $+$ | $x_{2}$ | - | $x_{3}$ | $\leq$ | 7 |
|  | $-x_{1}$ | - | $x_{2}$ | + | $x_{3}$ | $\leq$ | -7 |
|  | $x_{1}$ | - | $2 x_{2}$ | + | $2 x_{3}$ | $\leq$ | 4 |
|  |  | $\chi_{2}$ |  |  |  | $\geq$ | 0 |
|  |  |  |  | Introduce slack variables |  |  |  |

subject to

$$
\begin{array}{rcrrrlrlr}
x_{4} & = & 7 & - & x_{1} & - & x_{2} & + & x_{3} \\
x_{5} & = & -7 & + & x_{1} & + & x_{2} & - & x_{3} \\
x_{6} & = & 4 & - & x_{1} & + & 2 x_{2} & - & 2 x_{3} \\
& x_{1}, x_{2}, x_{3}, x_{4}, x_{5}, x_{6} & & \geq & 0 & &
\end{array}
$$

## Converting Standard Form into Slack Form (2/3)



Converting Standard Form into Slack Form (3/3)


| maximise | $2 x_{1}$ | - | $3 x_{2}$ | + | $3 x_{3}$ |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| subject to |  |  |  |  |  |  |  |  |

Use variable $z$ to denote objective function and omit the nonnegativity constraints.

## Converting Standard Form into Slack Form (3/3)

$$
\begin{aligned}
& \text { maximise } 2 x_{1}-3 x_{2}+3 x_{3} \\
& \text { subject to }
\end{aligned}
$$

> Use variable $z$ to denote objective function and omit the nonnegativity constraints.

## Converting Standard Form into Slack Form (3/3)

maximise $2 x_{1}-3 x_{2}+3 x_{3}$
subject to

Use variable $z$ to denote objective function and omit the nonnegativity constraints.

| $z$ | $=$ |  |  | $2 x_{1}$ | - | $3 x_{2}$ | + | $3 x_{3}$ |
| :--- | :--- | :--- | :--- | :--- | :--- | ---: | :--- | ---: |
| $x_{4}$ | $=$ | 7 | - | $x_{1}$ | - | $x_{2}$ | + | $x_{3}$ |
| $x_{5}$ | $=$ | -7 | + | $x_{1}$ | + | $x_{2}$ | - | $x_{3}$ |
| $x_{6}$ | $=$ | 4 | - | $x_{1}$ | + | $2 x_{2}$ | - | $2 x_{3}$ |

This is called slack form.

## Basic and Non-Basic Variables

| $z$ | $=$ |  |  | $2 x_{1}$ | - | $3 x_{2}$ | + | $3 x_{3}$ |
| :--- | :--- | :--- | :--- | ---: | :--- | ---: | :--- | ---: |
| $x_{4}$ | $=$ | 7 | - | $x_{1}$ | - | $x_{2}$ | + | $x_{3}$ |
| $x_{5}$ | $=$ | -7 | + | $x_{1}$ | + | $x_{2}$ | - | $x_{3}$ |
| $x_{6}$ | $=$ | 4 | - | $x_{1}$ | + | $2 x_{2}$ | - | $2 x_{3}$ |

## Basic and Non-Basic Variables



## Basic and Non-Basic Variables



## Basic and Non-Basic Variables



Slack Form (Formal Definition)
Slack form is given by a tuple ( $N, B, A, b, c, v$ ) so that

$$
\begin{aligned}
& z=v+\sum_{j \in N} c_{j} x_{j} \\
& x_{i}=b_{i}-\sum_{j \in N} a_{i j} x_{j} \quad \text { for } i \in B
\end{aligned}
$$

and all variables are non-negative.

## Basic and Non-Basic Variables

$$
\begin{array}{rlrllrlr}
z & = & & 2 x_{1} & - & 3 x_{2} & + & 3 x_{3} \\
x_{4} & = & 7 & - & x_{1} & - & x_{2} & + \\
x_{5} & = & -7 & + & x_{1} & + & x_{2} & - \\
x_{6} & = & 4 & - & x_{1} & + & 2 x_{2} & - \\
2 x_{3}
\end{array}
$$

Basic Variables: $B=\{4,5,6\}$
Non-Basic Variables: $N=\{1,2,3\}$

Slack Form (Formal Definition)
Slack form is given by a tuple ( $N, B, A, b, c, v$ ) so that

$$
\begin{aligned}
& z=v+\sum_{j \in N} c_{j} x_{j} \\
& x_{i}=b_{i}-\sum_{j \in N} a_{i j} x_{j} \quad \text { for } i \in B
\end{aligned}
$$

and all variables are non-negative.
Variables/Coefficients on the right hand side are indexed by $B$ and $N$.

## Slack Form (Example)

$$
\begin{aligned}
& z=28-\frac{x_{3}}{6}-\frac{x_{5}}{6}-\frac{2 x_{6}}{3} \\
& x_{1}=8+\frac{x_{3}}{6}+\frac{x_{5}}{6}-\frac{x_{6}}{3} \\
& x_{2}=4-\frac{8 x_{3}}{3}-\frac{2 x_{5}}{3}+\frac{x_{6}}{3} \\
& x_{4}=18-\frac{x_{3}}{2}+\frac{x_{5}}{2}
\end{aligned}
$$

## Slack Form (Example)

$$
\begin{aligned}
& z=28-\frac{x_{3}}{6}-\frac{x_{5}}{6}-\frac{2 x_{6}}{3} \\
& x_{1}=8+\frac{x_{3}}{6}+\frac{x_{5}}{6}-\frac{x_{6}}{3} \\
& x_{2}=4-\frac{8 x_{3}}{3}-\frac{2 x_{5}}{3}+\frac{x_{6}}{3} \\
& x_{4}=18-\frac{x_{3}}{2}+\frac{x_{5}}{2}
\end{aligned}
$$



Slack Form (Example)

$$
\begin{aligned}
& z=28-\frac{x_{3}}{6}-\frac{x_{5}}{6}-\frac{2 x_{6}}{3} \\
& x_{1}=8+\frac{x_{3}}{6}+\frac{x_{5}}{6}-\frac{x_{6}}{3} \\
& x_{2}=4-\frac{8 x_{3}}{3}-\frac{2 x_{5}}{3}+\frac{x_{6}}{3} \\
& x_{4}=18-\frac{x_{3}}{2}+\frac{x_{5}}{2}
\end{aligned}
$$

## Slack Form Notation

- $B=\{1,2,4\}, N=\{3,5,6\}$

Slack Form (Example)

$$
\begin{aligned}
& z=28-\frac{x_{3}}{6}-\frac{x_{5}}{6}-\frac{2 x_{6}}{3} \\
& x_{1}=8+\frac{x_{3}}{6}+\frac{x_{5}}{6}-\frac{x_{6}}{3} \\
& x_{2}=4-\frac{8 x_{3}}{3}-\frac{2 x_{5}}{3}+\frac{x_{6}}{3} \\
& x_{4}=18-\frac{x_{3}}{2}+\frac{x_{5}}{2}
\end{aligned}
$$

## Slack Form Notation

- $B=\{1,2,4\}, N=\{3,5,6\}$

$$
A=\left(\begin{array}{lll}
a_{13} & a_{15} & a_{16} \\
a_{23} & a_{25} & a_{26} \\
a_{43} & a_{45} & a_{46}
\end{array}\right)=\left(\begin{array}{ccc}
-1 / 6 & -1 / 6 & 1 / 3 \\
8 / 3 & 2 / 3 & -1 / 3 \\
1 / 2 & -1 / 2 & 0
\end{array}\right)
$$

Slack Form (Example)

$$
\begin{aligned}
& z=28-\frac{x_{3}}{6}-\frac{x_{5}}{6}-\frac{2 x_{6}}{3} \\
& x_{1}=8+\frac{x_{3}}{6}+\frac{x_{5}}{6}-\frac{x_{6}}{3} \\
& x_{2}=4-\frac{8 x_{3}}{3}-\frac{2 x_{5}}{3}+\frac{x_{6}}{3} \\
& x_{4}=18-\frac{x_{3}}{2}+\frac{x_{5}}{2}
\end{aligned}
$$

## Slack Form Notation

- $B=\{1,2,4\}, N=\{3,5,6\}$

$$
A=\left(\begin{array}{lll}
a_{13} & a_{15} & a_{16} \\
a_{23} & a_{25} & a_{26} \\
a_{43} & a_{45} & a_{46}
\end{array}\right)=\left(\begin{array}{ccc}
-1 / 6 & -1 / 6 & 1 / 3 \\
8 / 3 & 2 / 3 & -1 / 3 \\
1 / 2 & -1 / 2 & 0
\end{array}\right)
$$

$$
b=\left(\begin{array}{l}
b_{1} \\
b_{2} \\
b_{4}
\end{array}\right)=\left(\begin{array}{c}
8 \\
4 \\
18
\end{array}\right)
$$

Slack Form (Example)

$$
\begin{aligned}
& z=28-\frac{x_{3}}{6}-\frac{x_{5}}{6}-\frac{2 x_{6}}{3} \\
& x_{1}=8+\frac{x_{3}}{6}+\frac{x_{5}}{6}-\frac{x_{6}}{3} \\
& x_{2}=4-\frac{8 x_{3}}{3}-\frac{2 x_{5}}{3}+\frac{x_{6}}{3} \\
& x_{4}=18-\frac{x_{3}}{2}+\frac{x_{5}}{2}
\end{aligned}
$$

## Slack Form Notation

- $B=\{1,2,4\}, N=\{3,5,6\}$

$$
\begin{gathered}
A=\left(\begin{array}{lll}
a_{13} & a_{15} & a_{16} \\
a_{23} & a_{25} & a_{26} \\
a_{43} & a_{45} & a_{46}
\end{array}\right)=\left(\begin{array}{ccc}
-1 / 6 & -1 / 6 & 1 / 3 \\
8 / 3 & 2 / 3 & -1 / 3 \\
1 / 2 & -1 / 2 & 0
\end{array}\right) \\
b=\left(\begin{array}{l}
b_{1} \\
b_{2} \\
b_{4}
\end{array}\right)=\left(\begin{array}{c}
8 \\
4 \\
18
\end{array}\right), \quad c=\left(\begin{array}{l}
c_{3} \\
c_{5} \\
c_{6}
\end{array}\right)=\left(\begin{array}{l}
-1 / 6 \\
-1 / 6 \\
-2 / 3
\end{array}\right)
\end{gathered}
$$

Slack Form (Example)

$$
\begin{aligned}
& z=28-\frac{x_{3}}{6}-\frac{x_{5}}{6}-\frac{2 x_{6}}{3} \\
& x_{1}=8+\frac{x_{3}}{6}+\frac{x_{5}}{6}-\frac{x_{6}}{3} \\
& x_{2}=4-\frac{8 x_{3}}{3}-\frac{2 x_{5}}{3}+\frac{x_{6}}{3} \\
& x_{4}=18-\frac{x_{3}}{2}+\frac{x_{5}}{2}
\end{aligned}
$$

## Slack Form Notation

- $B=\{1,2,4\}, N=\{3,5,6\}$

$$
\begin{gathered}
A=\left(\begin{array}{lll}
a_{13} & a_{15} & a_{16} \\
a_{23} & a_{25} & a_{26} \\
a_{43} & a_{45} & a_{46}
\end{array}\right)=\left(\begin{array}{ccc}
-1 / 6 & -1 / 6 & 1 / 3 \\
8 / 3 & 2 / 3 & -1 / 3 \\
1 / 2 & -1 / 2 & 0
\end{array}\right) \\
b=\left(\begin{array}{l}
b_{1} \\
b_{2} \\
b_{4}
\end{array}\right)=\left(\begin{array}{c}
8 \\
4 \\
18
\end{array}\right), \quad c=\left(\begin{array}{l}
c_{3} \\
c_{5} \\
c_{6}
\end{array}\right)=\left(\begin{array}{l}
-1 / 6 \\
-1 / 6 \\
-2 / 3
\end{array}\right)
\end{gathered}
$$

- $v=28$

