# MATB61 Midterm Preparation - Summary of Theorems

### Refer to Textbook Chapters:

- 1 Introduction to Linear Programming
- 2 The Simplex Method
- 3 Further Topics in Linear Programming

#### 0.1 Summary of Important Theorems and Definitions in Chapter 1

**Definition 1.1** A vector  $x \in \mathbb{R}^n$  satisfying the constraints of a linear programming problem is called a **feasible solution** to the problem.

**Definition 1.2** A feasible solution that maximizes or minimizes the objective function of a linear programming problem is called an **optimal solution**.

**Definition 1.3** A subset S of  $\mathbb{R}^n$  is called convex if for any two distinct points  $x_1$  and  $x_2$  in S, the line segment joining  $x_1$  and  $x_2$  lies in S. That is, S is convex if, whenever  $x_1$  and  $x_2 \in S$ , so is  $x = \lambda x_1 + (1 - \lambda)x_2$  for  $0 \le \lambda \le 1$ .

**Definition 1.4** A point  $x \in \mathbb{R}^n$  is a convex combination of the points  $x_1, x_2, ..., x_r$  in  $\mathbb{R}^n$  if for some real numbers  $c_1, c_2, ..., c_r$  which satisfy  $\sum_{i=1}^r c_i = 1$  and  $c_i \geq 0, 1 \leq i \leq r$ , we have  $x = \sum_{i=1}^r c_i x_i$ .

**Theorem 1.1** A closed half-space is a convex set.

**Theorem 1.2** A hyper-plane is a convex set.

**Theorem 1.3** The intersection of a finite collection of convex sets is convex.

**Theorem 1.4** Let A be an  $m \times n$  matrix, and let b be a vector in  $\mathbb{R}^m$ . The set of solutions to the system of linear equations Ax = b, if non-empty, is a convex set.

**Theorem 1.5** The set of all convex combinations of a finite set of points in  $\mathbb{R}^n$  is a convex set.

**Definition 1.5** A point u in a convex set S is called an extreme point of S if it is not an interior point of any line segment in S. That is, u is an extreme point of S if there are no distinct points  $x_1$  and  $x_2$  in S such that  $u = \lambda x_1 + (1 - \lambda)x_2, 0 < \lambda < 1$ .

**Theorem 1.6** Let S be a convex set in  $\mathbb{R}^n$ . A point u in S is an extreme point of S if and only if u is not a convex combination of other points of S.

**Theorem 1.7** Let S be the set of feasible solutions to a general linear programming problem.

- 1. If S is non-empty and bounded, then an optimal solution to the problem exists and occurs at an extreme point.
- 2. If S is non-empty and not bounded and if an optimal solution to the problem exists, then an optimal solution occurs at an extreme point.

3. If an optimal solution to the problem does not exist, then either S is empty or S is unbounded.

**Theorem 1.8** Suppose that the last m columns of A, which we denote by  $A'_1, A'_2, ..., A'_m$ , are linearly independent and suppose that  $x'_1A'_1 + x'_2A'_2 + ... + x'_mA'_m = b$ , where  $x'_i \geq 0$  for i = 1, 2, ..., m. Then the point  $x = (0, 0, ..., 0, x'_1, x'_2, ..., x'_m)$  is an extreme point of S.

**Theorem 1.9** If  $x = (x_1, x_2, ..., x_s)$  is an extreme point of S, then the columns of A that correspond to positive  $x_j$  form a linearly independent set of vectors in  $R^m$ .

**Theorem 1.10** At most m components of any extreme points of S can be positive. The rest must be zero.

**Theorem 1.11** Consider the linear programming problem

$$\max \quad z = c^{\top} x \tag{1}$$

subject to 
$$Ax = b$$
  
  $x \ge 0$  (2)

For the linear programming problem determined by (1) and (2), every basic feasible solution is an extreme point, and, conversely, every extreme point is a basic feasible solution.

**Theorem 1.12** The problem determined by (1) and (2) has a finite number of basic feasible solutions.

**Theorem 1.13** Every extreme point of S yields an extreme point of S' when slack variables are added. Conversely, every extreme point of S', when truncated, yields an extreme point of S.

**Theorem 1.14** The convex set S of all feasible solutions to a linear programming problem in standard form has a finite number of extreme points.

**Definition 1.6** A basic feasible solution to the linear programming problem given by (1) and (2) is a basic solution that is also feasible.

## 1.2 Summary of Important Theorems and Definitions in Chapter 23

**Definition 2.1** Two distinct extreme points in S' are said to be adjacent if as basic feasible solutions they have all but one basic variable in common.

#### 2.3 Summary of Important Theorems and Definitions in Chapter 3

**Theorem 3.1** Given a primal problem as follows

$$\max \quad z = c^{\top} x$$

$$subject \ to \quad Ax \le b$$

$$x > 0$$

The dual of its dual is again the primal problem.

**Theorem 3.2** The linear programming problem in canonical form given by

$$\begin{aligned} \max \quad z &= c^\top x \\ subject \ to \quad Ax &= b \\ x &\geq 0 \end{aligned}$$

has for its dual the linear programming problem

$$\max \quad z' = b^{\top} w$$
 subject to  $A^T w \ge c$  
$$w \text{ unrestricted}$$