

Duality

- Primal dual program
- Strong duality
- Farkas Lemma

Primal and dual programs

- If we transform the dual into an equivalent minimization problem and then form its dual, we obtain a problem equivalent to the original problem
- the dual of the dual is the primal.**

Primal

$$\begin{array}{ll} \min_x & c^T x \\ \text{s.t.} & \left. \begin{array}{l} a_i^T x \geq b_i \quad i \in M_1 \\ a_i^T x \leq b_i \quad i \in N_2 \\ a_i^T x = b_i \quad i \in M_3 \end{array} \right\} \quad \left. \begin{array}{l} x_j \geq 0 \quad j \in N_1 \\ x_j \leq 0 \quad j \in N_2 \\ x_j - \text{free} \quad j \in N_3 \end{array} \right\} \\ & \end{array}$$

dual variable
for each constraint

$$\max_y \quad y^T b$$
$$y_i \geq 0$$
$$y_i \leq 0$$
$$y_i - \text{free}$$
$$a_j^T y \leq c_j$$
$$a_j^T y \geq c_j$$
$$a_j^T y = c_j$$

Weak/strong duality

Thm. If x is primal feasible and y is dual feasible, then $c^T x \geq b^T y$.

- If the primal optimal cost is $-\infty$, then the dual is infeasible.

Suppose the optimal cost is $-\infty$ and dual has a feasible solution $y \Rightarrow b^T y \leq -\infty$, which is impossible.

- If the dual optimal cost is ∞ , then the primal is infeasible.

Thm If a linear programming problem has a optimal solution, so does its dual, and $p^* = d^*$.

Sufficient condition.

Suppose that (x, y, z) is primal/dual feasible.

By weak duality: $x^T c + z^T z = b^T y$

By strong duality, if (x, y, z) is primal/dual optimal:

$$z^T x = 0$$

conversely: If $z^T x = 0$, then

- $c^T x$ achieves upper bound
- $b^T y$ achieves lower bound

therefore (x, y, z) is primal/dual optimal.

Theorem: The primal/dual (x, y, z) is optimal iff

$$Ax = b, \quad x \geq 0, \quad A^T y + z = c, \quad z \geq 0, \quad z^T x = 0$$

Relationship between primal and dual LPs.

Recall: For linear program, exactly one of the following three possibilities hold:

- ① There is an optimal solution
- ② The problem is unbounded
- ③ The problem is infeasible.

$$c^T x \geq b^T y$$

Suppose $p^* = -\infty$ and dual problem has feasible y . By weak duality, $b^T y \leq -\infty$ which is impossible.

	finite optimal	unbounded	infeasible
dual			
finite optimal	✓	✗	✗
unbounded	✗	✗	✓
infeasible	✗	✓	✓

Interpretation of dual variables

primal	dual
$\min_{\mathbf{x}} \mathbf{c}^T \mathbf{x}$ \mathbf{x} s.t. $\mathbf{A}\mathbf{x} = \mathbf{b}$ $\mathbf{x} \geq 0$	$\max_{\mathbf{y}, \mathbf{z}} \mathbf{b}^T \mathbf{y}$ \mathbf{y}, \mathbf{z} s.t. $\mathbf{A}^T \mathbf{y} + \mathbf{z} = \mathbf{c}$ $\mathbf{z} \geq 0$

Suppose \mathbf{x}^* is optimal and non-degenerate, then

$$\mathbf{x}^* = \begin{bmatrix} \mathbf{B}^{-1} \mathbf{b} \\ \mathbf{0} \end{bmatrix} > 0 \quad \text{and} \quad \mathbf{x}_B^*(\varepsilon) = \mathbf{B}^{-1}(\mathbf{b} + \varepsilon \Delta \mathbf{b}) > 0 \quad \text{for small } \varepsilon.$$

$\xrightarrow{\mathbf{B}^T \mathbf{y} = \mathbf{c}_B}$

Reduced cost $\mathbf{z}^* = \mathbf{c} - \mathbf{A}^T \mathbf{y}^*$ doesn't change. Thus $(\mathbf{x}^*(\varepsilon), \mathbf{y}^*, \mathbf{z}^*)$

is primal/dual optimal and optimal cost is

$$\mathbf{c}_B^T \mathbf{x}_B^* = \mathbf{c}_B^T \mathbf{B}^{-1}(\mathbf{b} + \varepsilon \Delta \mathbf{b}) = \mathbf{y}^T (\mathbf{b} + \varepsilon \Delta \mathbf{b}) = \mathbf{y}^T \mathbf{b} + \varepsilon \mathbf{y}^T \Delta \mathbf{b}$$

Small change of \mathbf{b} results in small change in optimal cost.

Certificate of infeasibility

- Consider standard form constraints: $Ax = b, x \geq 0$
- Certificate of infeasibility: If there exists some vector y s.t. $A^T y \geq 0$ and $b^T y < 0$, standard form constraint is infeasible.

$$A^T y \geq 0 \Rightarrow x^T A^T y \geq 0 \text{ for all } x \geq 0$$

$$b^T y < 0 \Rightarrow x^T A^T y \neq b^T y \text{ for all } x \geq 0$$

$$\Rightarrow Ax \neq b \text{ for all } x \geq 0$$

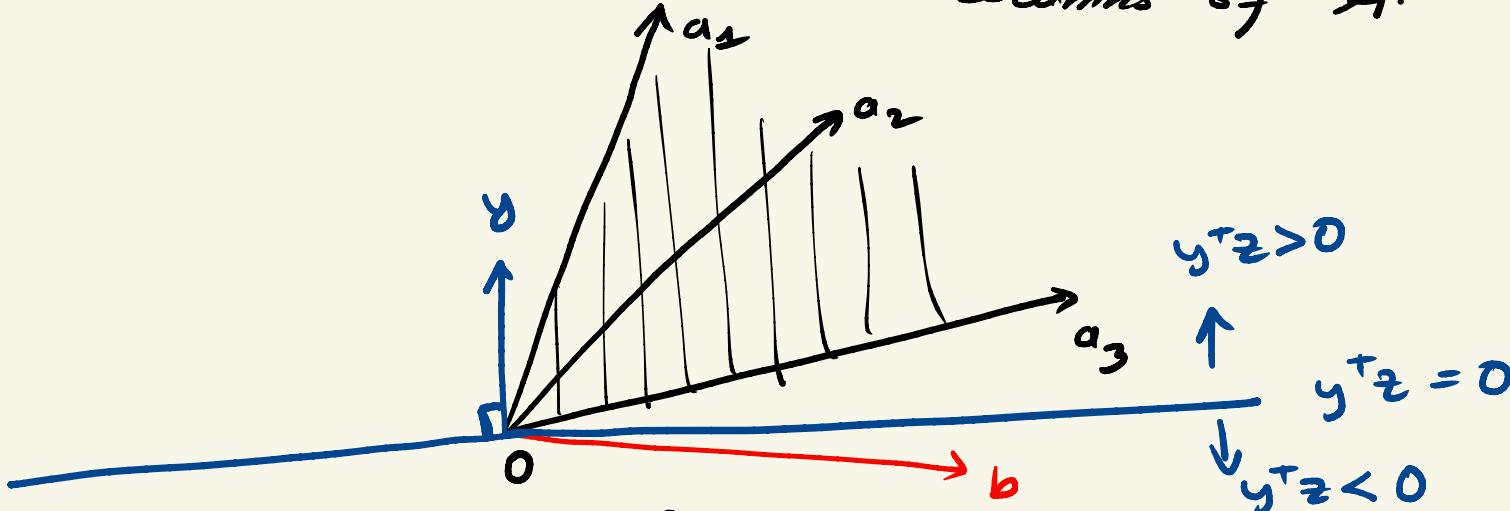
Farkas Lemma: Let $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$. Then exactly one of the following two alternatives hold:

- (a) There exists some $x \geq 0$ such that $Ax = b$
- (b) There exists some vector y s.t. $A^T y \geq 0$ & $b^T y < 0$

Geometric view of Farkas's lemma

Let a_i be i^{th} column of A : $A = \begin{bmatrix} a_1 & a_2 & \dots & a_n \end{bmatrix}$

Then $Ax = \sum_{i=1}^n a_i x_i \rightarrow$ linear combination of columns of A .



In figure, $Ax \neq b$ for any $x \geq 0$ because we
the hyperplane $\{z \mid y^T z = 0\}$ separates b and $\{Ax \mid x \geq 0\}$

Separating hyperplane theorem

- Every polyhedron $D = \{x \mid Ax \geq b\}$ is closed.
 - Separating hyperplane: Let S be a non empty closed convex subset of \mathbb{R}^n and $x^* \in \mathbb{R}^n$ be a vector that does not belong to S . Then there exists some vector $c \in \mathbb{R}^n$ such that $c^T x^* < c^T x \quad \forall x \in S$.
 - $S = \{Ax \mid x \geq 0\}$ is closed
- $S' = \{(x, y) \mid Ax = y, x \geq 0\}$ - is a polyhedron
- and $S = \text{proj}_y(S')$

