

MATH 464 – Lecture 25 (04/11/2023)

- Today:
- * illustration of Farkas' lemma
 - * asset pricing (application of Farkas' lemma).
 - * Optimal dual prices as marginal costs.

On AMPL: For the inventory problem, best to set param n as the # months so that you can use $1..n$ or $0..n-1$ or $0..n$ as the required sets to write, e.g., the balance constraint in one line:

$$s[i-1] + x[i] = \text{Demand}[i] + s[i]; \text{ for } i=1..n.$$

You could work with a set Months := m1 m2 ... m12; but then you can't directly index previous/next month: AMPL will **not** take m2+1 as m3!

Farkas' lemma

$$(a) \exists \bar{x} \geq \bar{0} \text{ s.t. } A\bar{x} = \bar{b}$$

$$(b) \exists \bar{p} \text{ s.t. } \bar{p}^T A \geq \bar{0} \text{ and } \bar{p}^T \bar{b} < 0.$$

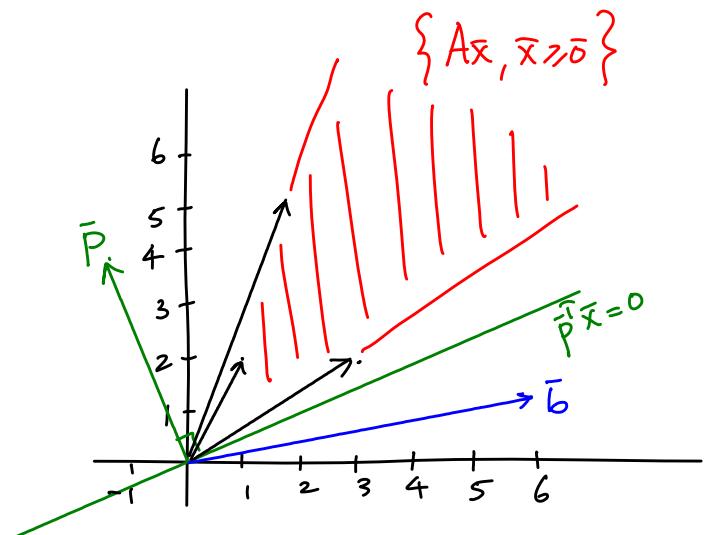
Illustration in 2D

$$A = \begin{bmatrix} 1 & 3 & 2 \\ 2 & 2 & 5 \end{bmatrix}, \bar{b} = \begin{bmatrix} 6 \\ 1 \end{bmatrix}.$$

Does there exist $\bar{x} \geq \bar{0}$ s.t. $A\bar{x} = \bar{b}$?
complicating part.

No, as seen in the picture!

\bar{b} is not in the cone $\{\bar{A}\bar{x} | \bar{x} \geq \bar{0}\}$.



For $\bar{p} = \begin{bmatrix} -1 \\ 4 \end{bmatrix}$, we get $\bar{p}^T A = [-1 \ 4] \begin{bmatrix} 1 & 3 & 2 \\ 2 & 2 & 5 \end{bmatrix} = [7 \ 5 \ 18] \geq \bar{0}$

$$\text{and } \bar{p}^T \bar{b} = [-1 \ 4] \begin{bmatrix} 6 \\ 1 \end{bmatrix} = -2.$$

We see that $\{\bar{A}\bar{x} | \bar{x} \geq \bar{0}\}$ and \bar{b} lie on opposite sides of $\{\bar{x} | \bar{p}^T \bar{x} = 0\}$. Hence the vector \bar{p} gives a certificate of infeasibility of the original system. Note that there could be multiple such \bar{p} vectors possible here - any solution to the alternative system in (b) will do.

In this example, note that the $\bar{x} \geq \bar{0}$ restriction is indeed the "complicating" constraint. If we wanted to check if $A\bar{x} = \bar{b}$ (for any \bar{x}), then answer would have been YES, since the columns of A span \mathbb{R}^2 here. But we could present Farkas' lemma in an equivalent form without $\bar{x} \geq \bar{0}$ in alternative (a).

Alternative forms of Farkas' lemma

Exactly one of the following two alternatives hold:

$$(a) \exists \bar{x} \text{ s.t. } A\bar{x} = \bar{b}.$$

$$(b) \exists \bar{p} \text{ s.t. } \bar{p}^T A = \bar{0}, \bar{p}^T \bar{b} = -1. \rightarrow \text{similar to } [\bar{0} \ \bar{0} \dots \bar{0} | \blacksquare]$$

where $\blacksquare \neq 0$, which indicates $A\bar{x} = \bar{b}$ is inconsistent.

Idea: If (b) holds, we can derive (by taking linear combinations of rows) $0 = -1$ from $A\bar{x} = \bar{b}$, and hence (a) cannot hold.

Proof (part of).

(\Leftarrow) If (a) does not hold, (b) holds.

$$(P) \quad \begin{array}{ll} \max & \bar{0}^T \bar{x} \\ \text{s.t.} & A\bar{x} = \bar{b} \quad \bar{p} \end{array} \quad \begin{array}{ll} \min & \bar{p}^T \bar{b} \\ \text{s.t.} & \bar{p}^T A = \bar{0}^T \end{array} \quad (D)$$

(a) does not hold means (P) is infeasible. So (D) is either infeasible or unbounded. But $\bar{p} = \bar{0}$ is feasible for (D), so (D) is unbounded. Hence $\exists \bar{p} \text{ s.t. } \bar{p}^T A = \bar{0}^T$ and $\bar{p}^T \bar{b} < 0$. We can scale \bar{p} such that $\bar{p}^T \bar{b} = -1$ holds. Hence (b) holds.

Application of Farkas' Lemma - Asset Pricing

There are n assets, and after a period of investment there are m possible states (or scenarios). Let r_{ij} be the return from asset j under state/scenario i . Hence if you invest \$1 in asset j , and the state turns out to be i , you expect to receive $\$r_{ij}$.

Assumption 1 r_{ij} is same for both $x_j \geq 0$ and $x_j < 0$.

($x_j < 0$ means you're in a "short" position — you sell $|x_j|$ at the start, with the promise to buy back at the end at rate r_{ij} if state is i . The payoff of $r_{ij}x_j < 0$ here, meaning you have to pay $r_{ij}|x_j|$).

Assumption 1 might not hold in general, but that's not too critical for our illustration here.

Let $R = [r_{ij}] \in \mathbb{R}^{m \times n}$, and $\bar{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ be the amount of assets at start ($x_j \geq 0$ and $x_j < 0$ are both possible).

Let \bar{w} represent the wealth at the end of the period. Then we write $\bar{w} = R\bar{x}$. Let \bar{p} be the prices at the start for each asset. Then $\bar{p}^T \bar{x}$ is the total cost of the portfolio.

An arbitrage is a price such that $R\bar{x} \geq \bar{p}$ but $\bar{p}^T \bar{x} < 0$, i.e., we get a non-negative return for a negative price.

Assumption 2 Prices will equilibrate to avoid arbitrages.

"No arbitrages" means "if $R\bar{x} \geq \bar{0}$ then $\bar{p}^T \bar{x} \geq \bar{0}$ ".

(BT-1LD Theorem 4.8) There exist no arbitrages iff there exists

$$\bar{q}_j = \begin{bmatrix} q_1 \\ \vdots \\ q_m \end{bmatrix} \geq \bar{0} \text{ such that } \bar{p} = R^T \bar{q}_j, \bar{q}_j \geq \bar{0}.$$

Intuitively, this theorem says that there will be no arbitrages iff the prices are some non-negative linear combination of the return rates $R = [r_{ij}]$. This result is usually presented independent of any connections to linear programming. But we can prove it by a direct application of Farkas' lemma.

Proof Follows from Farkas' lemma

- (a) $\exists \bar{x} \geq \bar{0}$ s.t. $A\bar{x} = \bar{b}$.
- (b) $\exists \bar{p}$ s.t. $\bar{p}^T A \geq \bar{0}, \bar{p}^T \bar{b} < 0$.

We construct one alternative to represent arbitrages. The other alternative then presents the criterion given in the theorem for existence of \bar{q}_j .

$$(a) \quad \begin{aligned} \bar{R}^T \bar{q}_j &= \bar{p} \\ \bar{q}_j &\geq \bar{0} \end{aligned}$$

$$\begin{aligned} R\bar{x} &\geq \bar{0} & (b) \\ \bar{p}^T \bar{x} &< 0 \\ (\text{arbitrage}) \end{aligned}$$

Optimal dual variables as marginal costs

$$(P) \quad \begin{array}{l} \min \bar{c}^T \bar{x} \\ \text{s.t. } A\bar{x} = \bar{b} \\ \bar{x} \geq \bar{0} \end{array}$$

Assume (P) has optimal solution \bar{x}^* that is non-degenerate, and let B be the corresponding basis matrix.

Then $\bar{x}_B = B^{-1}\bar{b} \geq \bar{0}$. Optimal cost is $\bar{c}_B^T \bar{x}_B$.

Also, the reduced costs: $\bar{c}'^T = \bar{c}^T - \bar{c}_B^T B^{-1}A \geq \bar{0}$ (optimality conditions).

Consider changing \bar{b} to $\bar{b} + \bar{d}$ (rhs vector). Since $\bar{x}_B > \bar{0}$, for \bar{d} small enough, the new optimal solution $\bar{x}_B = B^{-1}(\bar{b} + \bar{d}) \geq \bar{0}$. And hence the new optimal cost is $\bar{c}_B^T B^{-1}(\bar{b} + \bar{d}) = \underbrace{\bar{c}_B^T B^{-1} \bar{b}}_{\text{original cost}} + \underbrace{\bar{c}_B^T B^{-1} \bar{d}}_{\bar{p}^T}$.

Hence we get new optimal cost = old optimal cost + $\bar{p}^T \bar{d}$. Thus \bar{p} can be interpreted as the marginal cost (or shadow price) for increasing each rhs entry by 1 unit. In particular, p_i is the shadow price for "resource" i , i.e., the price to pay for increasing b_i to $b_i + 1$.

Dual Simplex Method

Caution! This is **NOT** applying simplex method to the dual LP!

Tableau for the primal simplex method:

$$(P) \begin{array}{ll} \min & \bar{c}^T \bar{x} \\ \text{s.t.} & \bar{A}\bar{x} = \bar{b} \\ & \bar{x} \geq 0 \end{array}$$

$$\max \bar{p}^T \bar{b} \\ \text{s.t. } \bar{p}^T \bar{A} \leq \bar{c}^T \quad (D)$$

allow < 0
strive to make
them ≥ 0 .

$-\bar{c}_B^T \bar{x}_B$	$\bar{c}^T - \bar{c}_B^T \bar{B}^{-1} \bar{A}$
$\bar{B}^{-1} \bar{b}$	$\bar{B}^{-1} \bar{A}$

$\bar{c}' \geq 0$ (maintain)

more in the next lecture...