

# Econ 172A - Slides from Lecture 1

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# WELCOME

1. Outlines Distributed.
2. Copying Information from Slides is Probably a Waste of Time

# INTRODUCTIONS

- ▶ Joel, 311 Econ, TuTh 2:30–3:20
- ▶ TAs
  1. Leland Farmer, 128 Econ, M 2:30–3:30
  2. Erin Giffin, 117 Econ, Tu 5:00–6:00

Class email: [econ172af12@gmail.com](mailto:econ172af12@gmail.com)

# COURSE MATERIAL

1. Book: Expensive, available, old editions OK. Buying it unnecessary.
2. Lecture “notes”: Class web page, Soft Reserves.
3. Old problems, exams, solutions: Class web page
4. Excel’s Solver
5. Sections Tuesdays in York Hall 226 from 8-9 PM and from 9-10 PM.

<http://www.econ.ucsd.edu/%7Ejsobel/172f12/172f12home.htm>

# Prerequisites

1. Whatever is in the catalog. (Intermediate Micro, Probability, Linear Algebra)
2. In fact: Linear Algebra, ability to follow a logical argument, spreadsheet experience.

# WAIT LIST

Not my job.

Go to departmental office, Sequoyah Hall 245.

# GRADING

1. 15% quizzes (4 given, best 3 count)
2. 35% midterm (November 1)
3. 50% final (December 10)

Ground rules: No books, no notes, no electronic aids.

Curve? Not really.

Straight scale? Not really.

# ADMINISTRATIVE

1. Wait List? 245 Sequoyah.
2. Don't have access to Excel? Installed on campus machines.
3. Google docs spreadsheet also has a solver.
4. Don't have access to campus machines? See me.
5. Can't make the quizzes, midterm, final at scheduled time?  
One quiz: Drop it. Otherwise: Drop class.
6. If you are not sure what cheating is, talk to me.

# HOW TO STUDY

1. Work problems.
2. Learn terminology and results from lecture.
3. Ask yourself questions.

# HOW NOT TO STUDY

1. “Read” lecture notes and book.
2. Review answers to problems prior to working problems.

# WHAT (I hope) YOU WILL LEARN

1. Translate verbal statement to math and back.  
(Econ/Management)
2. Basic Structure of Linear Programming. (Math)
3. What is an Algorithm. (Computer Science)
4. What solutions look like and how they change when problem changes. (Math/Econ)

# TOPICS

Problem Formulation

Graph

Duality

Complementary Slackness

Sensitivity Analysis

Integer Programming

Branch and Bound

Network Algorithms

Transportation and Assignment Problems

# WARNING

1. Second time slide presentation for UG.
2. It did not work well last time.
3. This time I plan to post slides but use a lot of chalk in lectures.

# OPERATIONS RESEARCH

1. Originally: research designed to “optimize” military operations.
2. Currently: Use of mathematical models to study problems that arise in managerial/industrial decision making.
3. Aspects:
  - 3.1 Pure Mathematics: When do problems have solutions? Characterization? Existence/properties of algorithms.
  - 3.2 Computer Science: Design of algorithms. Complexity of Problems.
  - 3.3 Management Science: Domain of applications. Interpretation of Solutions.

# MATHEMATICAL PROGRAMMING PROBLEM

Problem of the form:

$$\max f(x) \text{ subject to } x \in S.$$

1.  $f$  is called the **objective function**.
2.  $S$  is called the **feasible set** or **constraint set**.

# MEANING

$$\max f(x) \text{ subject to } x \in S.$$

- ▶ **solution:** Special choice of  $x$ ,  $x^*$ , satisfying
  1.  $x^* \in S$ . (“ $x^*$  is feasible.”)
  2. If  $x \in S$  then  $f(x^*) \geq f(x)$ . (“ $x^*$  is optimal.”)
- ▶ **value:**  $f(x^*)$ .

# PROPERTIES OF SOLUTIONS

1. Solutions: May not exist for two reasons.
  - 1.1  $S$  is empty (“problem is **not feasible**”).
  - 1.2 It is possible to make the value  $f(x)$  arbitrarily large (“problem is **unbounded**”).
2. Solution may be unique.
3. There may be more than one solution.

# PROPERTIES OF VALUES

Values must be unique (if they exist).

# MINIMIZATION PROBLEM

Same theory as max.

If you can solve:

$$\max f(x) \text{ subject to } x \in S,$$

then you can solve

$$\min g(x) \text{ subject to } x \in S$$

Solve:

$$\max -f(x) \text{ subject to } x \in S.$$

This describes the solution(s) to the min problem.

Value is multiplied by  $-1$ .

# SPECIAL KINDS OF MATH PROG PROBLEMS

1. Linear Programming ( $f$  linear function,  $S$  defined by linear inequalities)
2. Non-linear programming ( $f$  arbitrary function,  $S$  defined by non-linear inequalities)
3. Integer Programming (Some components of  $x$  constrained to be integers.)

Linear Programming – this class. “Discrete” methods.

Non-linear Programming – 172B. “Calculus” methods.

Integer Programming – this class. Specialized algorithms.

# LINEAR FUNCTION

$$f(x) = c \cdot x = \sum_{j=1}^n c_j x_j = c_1 x_1 + \cdots + c_n x_n$$

Here  $x = (x_1, \dots, x_n)$  are variables.  $c = (c_1, \dots, c_n)$  are constants.

# LINEAR CONSTRAINT

$$\sum_{j=1}^n a_{ij}x_j \leq b_i$$

or

$$Ax \leq b$$

where  $A$  is **technology matrix**, with  $n$  columns and  $m$  rows.  
 $b = (b_1, \dots, b_m)$ .

# DEFINING PROPERTIES OF LINEAR FUNCTIONS

1. Additivity:  $f(x + y) = f(x) + f(y)$
2. Constant Returns to Scale:  $f(ax) = af(x)$  for any number  $a$ .

# Economic Interpretation

- ▶ Additivity: You can combine independent production processes.
- ▶ Returns to Scale: Doubling Process Costs Twice as Much.

# GENERAL FORM OF LINEAR PROGRAM

$\max c \cdot x$  subject to  $Ax \leq b, x \geq 0$ .

- ▶  $x = (x_1, \dots, x_n)$  are the  $n$  variables.
- ▶  $c = (c_1, \dots, c_n)$  are the  $n$  coefficients of the objective function.
- ▶  $A$  is a matrix with  $n$  columns and  $m$  rows.
- ▶ The entry in row  $i$  and column  $j$  is  $a_{ij}$

It turns out that any linear programming problem can be written in this way.

# JARGON

Together  $A$ ,  $b$ ,  $c$  are the data of the problem (given constants).

Jargon:

- ▶  $c$ : coefficients of the objective function
- ▶  $b$ : Right hand side constants
- ▶  $A$ : technology
- ▶  $x \geq 0$ : nonnegativity constraint.

# WHEN IS LINEARITY SENSIBLE?

1. Typically a good assumption about pricing.  
(if  $p_j$  is price per unit of good  $j$ ;  $x_j$  is number of units of good  $j$  purchased, then

$$p \cdot x$$

is how much it costs to buy  $x$ .)

But: maybe volume discounts; maybe you can't buy half an apple.

2. Linearity is not a good model of utility (when there is decreasing marginal utility).
3. Linearity is not a good assumption when there are decreasing or increasing returns to scale.
4. Linearity is not a good assumption when there are indivisibilities (units come in whole numbers).