

Linear Programming

0.1 Linear Programming (LP) Structure

- **Standard LP Form:**

$$\max \pi'x \quad \text{s.t. } Ax \leq b, x \geq 0$$

- Characteristics:
 - Linear objective.
 - Linear constraints.
 - Nonnegativity.

Can also be written in matrix notation.

Economic interpretation:

- π : profit per unit.
 - A : resource use matrix.
 - b : resource endowment.
-

0.2 Model Building Process (McCarl framework)

1. Identify **decision variables**.
 2. State the **objective**.
 3. Identify and formulate **constraints**.
 4. Collect data.
 5. Translate into computer-readable form.
 6. Solve and interpret.
-

0.3 Farm LP Example

A farmer has 500 acres of land available and is deciding how to allocate it between wheat and corn. Each acre of wheat yields a profit of \$200, requires 3 hours of labor, and 4 units of fertilizer. Each acre of corn yields a profit of \$300, requires 4 hours of labor, and 3 units of fertilizer. The farm has at most 1,800 hours of labor available and 2,000 units of fertilizer.

Formulate this situation as a linear programming problem. Clearly define the decision variables, write down the objective function representing total profit, and specify the constraints that capture the land, labor, fertilizer, and nonnegativity restrictions.

Decision variables.

Let W = acres of wheat, C = acres of corn.

0.4 Scalar (algebraic) form

$$\begin{aligned} \max_{W, C} \quad & 200W + 300C \\ \text{s.t.} \quad & W + C \leq 500 \quad (\text{land}) \\ & 3W + 4C \leq 1800 \quad (\text{labor}) \\ & 4W + 3C \leq 2000 \quad (\text{fertilizer}) \\ & W, C \geq 0. \end{aligned}$$

0.5 Matrix (compact) form

$$\begin{aligned} \max_{x \in \mathbb{R}_{\geq 0}^2} \quad & c^\top x \\ \text{s.t.} \quad & Ax \leq b, \end{aligned} \quad \text{with} \quad x = \begin{bmatrix} W \\ C \end{bmatrix}, \quad c = \begin{bmatrix} 200 \\ 300 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 1 \\ 3 & 4 \\ 4 & 3 \end{bmatrix}, \quad b = \begin{bmatrix} 500 \\ 1800 \\ 2000 \end{bmatrix}.$$

1 Summary

- Optimization = decision making under constraints.
- Mathematical programming provides a general framework: decision variables, objective, constraints.
- Linear programs (LPs): linear objective + linear constraints + nonnegativity.
- Example: Farm resource allocation with land, labor, fertilizer.

2 Day 2: Linear Programming in Practice

2.1 Review

- General form of a mathematical program.
 - Structure of a linear program:
 - Linear objective
 - Linear constraints
 - Nonnegativity
 - Wheat–Corn farm allocation example in scalar and matrix notation.
-

2.2 7 Assumptions of Linear Programming

Linear programming models rely on a set of assumptions that make them tractable but also limit their realism. McCarl & Spreen (Ch. 2.4) identify **seven important assumptions**. The first three involve the *appropriateness of the formulation*; the last four describe *mathematical properties* of the LP model.

2.2.1 1. Objective Function Appropriateness

- The objective function is assumed to be the **sole criterion** for evaluating solutions.
- This means the decision maker's preferences can be fully represented by a single linear function (e.g., profit, cost, utility).
- In practice, decisions may depend on multiple objectives (profit, risk, leisure), but LP assumes one dominates.

2.2.2 2. Decision Variable Appropriateness

- All relevant decision variables must be included, and each must be **fully controllable** by the decision maker.
- Omitting key variables or including variables outside the decision maker's control invalidates the formulation.

2.2.3 3. Constraint Appropriateness

- Constraints must **accurately and completely capture** the limits faced by the decision maker:
 - They fully describe resource, technological, and institutional limits.
 - Resources within a constraint are **homogeneous** and freely substitutable among activities.
 - No constraint should arbitrarily rule out feasible choices.
 - Constraints cannot be bent outside the model.

2.2.4 4. Proportionality

- Contributions of activities to the objective function are **proportional** to their level.
- Likewise, resource use is proportional: doubling an activity doubles its input use.
- This rules out fixed costs, economies of scale, or price effects that depend on output level.

2.2.5 5. Additivity

- Total contributions to the objective and resource use are the **sum of individual contributions**.
- No interactions among variables are allowed (e.g., no multiplicative terms).

2.2.6 6. Divisibility

- Decision variables can take on **fractional values**.
- This assumes continuous activities (e.g., acres of land).
- When variables must be integer (e.g., number of tractors), integer programming is required instead.

2.2.7 7. Certainty

- All parameters (objective coefficients, resource availability, input-output coefficients) are **known with certainty**.
- LP is thus a deterministic model.
- In practice, parameters are often estimated, and uncertainty can be explored with sensitivity or stochastic programming.

2.3 Teaching Note

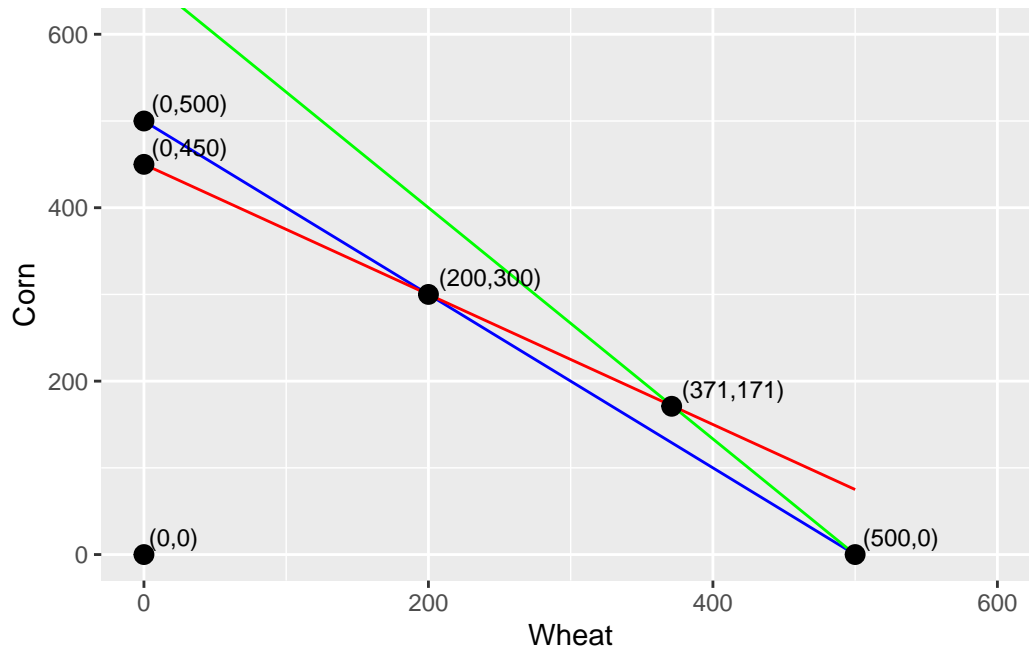
These assumptions both **enable LP to be solvable** and **limit realism**. They provide a natural segue to later topics in the course:

- Multi-objective programming (relax objective function assumption)
 - Integer programming (relax divisibility)
 - Stochastic programming (relax certainty)
 - Nonlinear programming (relax proportionality and additivity)
-

2.4 Graphical Solution Method (2 variables)

Step 1. Draw the constraints.

- Land: $W + C \leq 500$
- Labor: $3W + 4C \leq 1800$
- Fertilizer: $4W + 3C \leq 2000$
- Nonnegativity: $W, C \geq 0$
- Plot based on endpoints. set one var 0 and devote all resources to that var.



Step 2. Identify the feasible region.

- Intersection of all constraints in the (W, C) plane.
- Polygon bounded by lines.

Step 3. Plot the objective function.

- Profit = $200W + 300C$.
- Show isoprofit lines:
 - Suppose we plot \$6000 profit. All wheat no corn, then all corn no wheat.
 - lines of constant profit slope $-\frac{200}{300} = -\frac{2}{3}$.

Step 4. Locate the optimum.

- Slide the isoprofit line outward until the last point of contact with the feasible region.
- Optimum is always at a **corner point** (fundamental theorem of LP).

2.5 Simplex Method

2.5.1 Why We Need It

- The graphical method only works for **two variables**.
- Real problems may involve **hundreds or thousands** of variables.
- Key geometric fact:
 - The feasible region of an LP is a **convex polytope**.
 - The **optimal solution lies at a vertex (corner point)**. What about the problem makes this a fact?
- The simplex method provides a **systematic way** to move from vertex to vertex until the best one is found.

2.5.2 Core Idea

- Start at a **basic feasible solution (BFS)** — a corner point of the feasible region.

- At each step:
 1. Compute **reduced costs** (how much the objective improves if a variable increases from 0).
 2. Identify an **entering variable** (the candidate to increase).
 3. Determine which constraint binds first — this sets the **leaving variable**.
 4. **Pivot** to a new BFS.
 - Stop when no variable can improve the objective — this is optimal.
-

2.6 Intuition

- Simplex is like **walking along the edges** of the feasible polygon.
 - At each corner, ask: “*If I move along this edge, does profit go up?*”
 - Continue until no edge yields improvement.
 - The same logic applies in higher dimensions, even though we cannot draw the polytope.
-

2.7 Simplex Pivot — Tiny Worked Example

We use the smallest LP that still shows the mechanics:

Problem

$$\begin{aligned} \max z &= 3x_1 + 2x_2 \\ \text{s.t. } x_1 + x_2 &\leq 4, \quad x_1 \leq 2, \quad x_1, x_2 \geq 0. \end{aligned}$$

Standard form (add slacks s_1, s_2)

$$\begin{aligned} x_1 + x_2 + s_1 &= 4, \\ x_1 + s_2 &= 2, \\ z - 3x_1 - 2x_2 &= 0, \quad s_1, s_2 \geq 0. \end{aligned}$$

- Slack variables are added to “ \leq ” constraints to convert them into equalities, making the LP system compatible with the simplex algorithm.

- They measure the unused portion of a resource — e.g., if a land constraint is $x_1 + x_2 \leq 500$ and only 400 acres are used, the slack variable equals 100.
 - Slack variables always have a zero coefficient in the objective function, since they do not directly contribute to profit or cost.
 - Each slack variable typically appears with a coefficient of +1 in one constraint and 0 elsewhere, so they “fill the gap” between resource availability and resource use.
 - At optimality, a nonzero slack indicates an unused resource. Checking which slack variables are positive helps interpret whether constraints are binding or loose.
-

2.8 Initial tableau and choice of entering/leaving variables

Initial Basic Feasible Solution (BFS): $x_1 = x_2 = 0 \Rightarrow s_1 = 4, s_2 = 2, z = 0$.

Tableau

	x_1	x_2	s_1	s_2	RHS
s_1	1	1	1	0	4
s_2	1	0	0	1	2
z	-3	-2	0	0	0

- **Entering variable:** look at the objective row; most negative reduced cost is under x_1 (coefficient -3) \rightarrow **enter** x_1 .
- **Leaving variable (ratio test):** divide RHS by the positive entries in the x_1 column:
Row s_1 : $4/1 = 4$, Row s_2 : $2/1 = 2$. Minimum is 2 \rightarrow **leave** s_2 .
- **Pivot element:** the entry at row s_2 , column x_1 (which is 1).

We will **pivot on that 1**, swapping $s_2 \leftrightarrow x_1$.

2.9 Row operations (make pivot column a unit vector)

Goal: pivot column (x_1) should become $(0, 1, 0)^\top$.

- 1) **Normalize pivot row** (already 1, so no change):

$$(s_2) : [1 \ 0 \ 0 \ 1 \mid 2].$$

- 2) **Zero out the other entries in the x_1 column:**

- Row s_1 : $(s_1) \leftarrow (s_1) - 1 \cdot (s_2)$

$$[1 \ 1 \ 1 \ 0 \mid 4] - [1 \ 0 \ 0 \ 1 \mid 2] = [0 \ 1 \ 1 \ -1 \mid 2].$$

- Row z : $(z) \leftarrow (z) + 3 \cdot (s_2)$

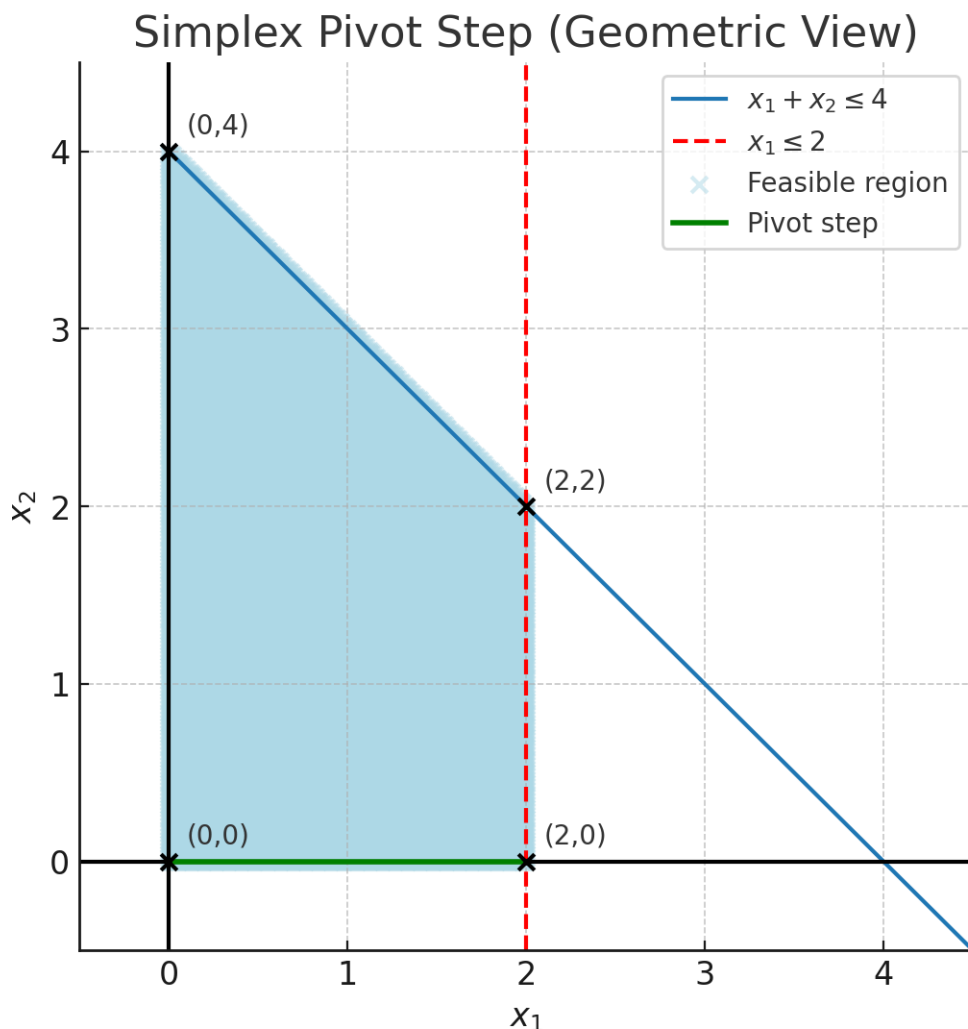
$$[-3 \ -2 \ 0 \ 0 \mid 0] + 3 \cdot [1 \ 0 \ 0 \ 1 \mid 2] = [0 \ -2 \ 0 \ 3 \mid 6].$$

2.10 New tableau (after one pivot)

	x_1	x_2	s_1	s_2	RHS
s_1	0	1	1	-1	2
x_1	1	0	0	1	2
z	0	-2	0	3	6

- **Basis after pivot:** $\{s_1, x_1\}$.
- **Current solution:** $x_1 = 2$, $x_2 = 0$, $s_1 = 2$, $s_2 = 0$, $z = 6$.
- The column labels show that x_1 **entered** (blue row) and s_2 **left**.

Next step (if continuing): the most negative in the z -row is under x_2 (-2), so x_2 would enter next; the algorithm would pivot again and reach the optimum at $(x_1, x_2) = (2, 2)$ with $z = 10$.



2.11 Excel Solver

Excel implements the simplex method in the solver add-on. See [LP_land_alloc.xlsx](#)

2.12 Summary

- Linear programming problems have:
 - **Decision variables** (choices to make),
 - **Objective function** (profit, cost, etc.),
 - **Constraints** (resource limits, requirements).
- Graphical method (2 variables) shows:
 - Feasible region = convex polygon.
 - Optimum occurs at a **corner point**.
- **Fundamental theorem of LP:** optimum is always at a vertex of the feasible region.
- **Simplex method** generalizes:
 - Moves systematically from one **basic feasible solution (BFS)** to another.
 - Uses **entering and leaving variables** to pivot.
 - Stops when no further improvement is possible.
- **Slack variables** convert inequalities to equalities and measure unused resources.
- Tableau row operations implement pivots (targeted Gaussian elimination).

3 Sensitivity Analysis

3.1 Review

- LPs make 7 assumptions. What are a few?
- How do LPs algorithms solve? What is the fundamental theorem of LP?

3.2 Motivation

- LPs give an **optimal solution** and an **objective value**.
 - But: parameters (profits, resources, input requirements) are rarely known with certainty.
 - We need to know how **robust** the solution is.
-

3.3 What is Sensitivity Analysis?

- Also called **post-optimality analysis**.
 - Asks: *how much can model parameters change before the current solution changes?*
 - Focus on three categories:
 1. RHS (resources)
 2. Objective coefficients (profits/costs)
 3. Technical coefficients (input-output relationships)
-

3.4 RHS (Right-Hand Side) Ranging

- Suppose resource availability changes:

$$b_{new} = b_{old} + \Delta r$$

- Current solution remains optimal **as long as basic variables stay ≥ 0** .
 - Interpretation:
 - Within the allowable range, the **shadow price** is valid.
 - Outside the range, the optimal activity mix shifts.
-

3.5 Objective Coefficient Ranging

- How much can a profit coefficient change before the basis changes?
 - For **nonbasic variables**: reduced costs must remain ≥ 0 .
 - For **basic variables**: check feasibility of objective row with new coefficient.
-

3.6 Technical Coefficient Changes

- What if technology or input requirements change?
 - Example: wheat now needs 3.5 instead of 3 labor hours per acre.
 - Sensitivity analysis uses shadow prices to approximate the effect on objective value.
 - Interpretation: tighter labor constraints may shift which crop dominates.
-

3.7 The 100% Rule

The 100% Rule provides a way to evaluate whether simultaneous changes in objective function coefficients or right-hand side (RHS) values will preserve the current optimal basis — that is, whether the current solution remains optimal without re-solving the problem.

This rule applies separately to:

- Changes in objective function coefficients (i.e., the c_i 's), and
- Changes in the RHS values of constraints (i.e., the b_j 's).

Objective Function Coefficients

Suppose multiple objective coefficients change. For each decision variable x_i , let:

- Δc_i be the change in the objective coefficient.
- AL_i, AU_i be the allowable decrease and increase, respectively, as reported in the sensitivity analysis.

Define the proportion of the allowable range used for each change:

$$r_i = \begin{cases} \frac{|\Delta c_i|}{AL_i} & \text{if } \Delta c_i < 0 \\ \frac{|\Delta c_i|}{AU_i} & \text{if } \Delta c_i > 0 \end{cases}$$

Then, compute: $\sum_{i \in \text{changed vars}} r_i$. If: $\sum r_i \leq 1$, then the current optimal basis remains optimal, although the optimal value of the objective function will generally change. If the sum exceeds 1, the basis may change, and re-solving the problem is required.

3.8 Wheat–Corn Example (Excel Solver Report):

3.8.1 Excel top panel: Variable Cells

Interpretation of columns:

- **Final Value** – the optimal level of the decision variable (e.g., acres of wheat or corn).
- **Reduced Cost** – for nonbasic variables (value = 0), how much the objective coefficient must improve before the variable would enter the solution. Zero if the variable is positive in the solution.
- **Objective Coefficient** – the profit (or cost) per unit used in the objective function.
- **Allowable Increase / Decrease** – the range over which the objective coefficient can change without altering the current optimal basis (solution structure).

Interpretation of results

- Corn: optimal solution plants 450 acres of corn. Reduced cost = 0 because it's in the basis.
- Wheat: optimal solution plants 0 acres of wheat. Reduced cost = -25 means if wheat's profit increased by more than \$25/acre (from 200 \rightarrow 225), it would enter the solution.

Ranges:

- Wheat: profit can increase up to +25 before wheat enters.
- Corn: profit can fall as much as 33.3 (300 \rightarrow 266.7) before solution changes.
- Interpretation: Corn dominates under current prices. Wheat only becomes attractive if its relative profit improves substantially.

3.8.2 Excel bottom panel: Constraints

Interpretation of columns:

- **Final Value** – the amount of the resource actually used at the solution.
- **Constraint R.H. Side** – the available amount of the resource (the right-hand side of the inequality).
- **Shadow Price** – the marginal value of relaxing the constraint (increase in objective if RHS increases by 1 unit), valid only within the allowable range.
- **Allowable Increase / Decrease** – the range over which the shadow price remains valid and the current basis stays optimal.

Interpretation of results:

- Land: only 450 acres used out of 500. Slack = 50 acres. Shadow price = 0 because land is not binding. You can increase land indefinitely without improving profit (since labor is the true bottleneck).
- Labor: fully used (1800/1800). Shadow price = 75 means each additional labor hour would increase profit by \$75, valid for up to +200 extra hours.
- Fertilizer: only 1350 units used of 2000. Slack = 650. Shadow price = 0 because it's not binding.

Economic Takeaways for Discussion

- Which resource is scarce? Labor.
 - How much should the agent be willing to pay for an additional unit of labor? 75 per labor hour. This is also the opportunity cost of labor on this operation.
 - Why does wheat drop out of the solution? Because relative to corn it uses too much labor per profit dollar.
 - Policy thought experiment: If labor availability were increased by 200 hours, profit would rise by \$15,000 (200×75).
-

3.9 Economic Interpretation

- **Shadow prices:** marginal value of resources.
 - **Allowable ranges:** robustness of those shadow prices.
 - **Managerial use:**
 - Identify which resources are most binding.
 - Assess which profit coefficients are critical.
 - Evaluate new technologies or policy changes.
-

3.10 Summary

- Sensitivity analysis extends LP results beyond a single point solution.
 - Provides insight into:
 - Resource valuation (RHS changes),
 - Profit robustness (objective changes),
 - Technology shifts (coefficient changes).
-

4 Getting Started with R & RStudio

4.1 RStudio Orientation (2–3 min tour)

- **Source Pane (top-left):** where you edit scripts (.R) and notebooks (.qmd, .Rmd).

- **Console (bottom-left):** runs commands immediately (> prompt).
- **Environment/History (top-right):** objects in memory (data, vectors, functions).
- **Files/Plots/Packages/Help (bottom-right):** manage files, see plots, install packages, read docs.

Workflow tips - Create a **Project** (File → New Project...) for the course; it pins your working directory.
- Put scripts in a `code/` folder, data in `data/`, and outputs in `out/`. - Use **scripts** for anything you might need to re-run. Avoid one-off console work for assignments.

4.2 Preview

- Next class: **duality** — formalize the relationship between primal and dual problems.
- Show how shadow prices emerge naturally from the dual formulation.