

# Optimization – Problem Formulation

# NVF, Mathematical Models, and Optimization

→ Given a set of parameters and constraints, how do we optimize all or part of our NVF?

→ We will look briefly at model-based optimization

- We will not dive deep into the theory, rather we will provide an overview of optimization and show how to use Excel to perform optimization

→ Given a problem, we develop a mathematical model to describe the problem.

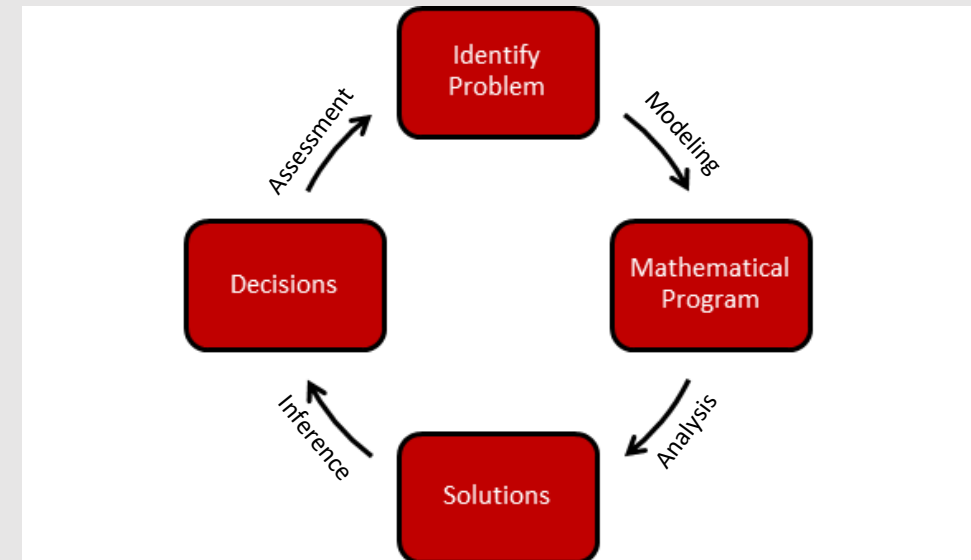
- Models include variables, constraints, objective function
- Example: Maximize net value considering labour, material, environment etc.

# What is Model-based Optimization?

→ Models are not perfect, they are a simplification of the real problem

- In model-based optimization, we draw conclusions from the *model* of the system, NOT from the problem itself
- It is critical to **develop an appropriate model for your system/problem/scenario**

→ Model-based optimization is typically iterative



# Components of an Optimization Model

- Optimization models (mathematical programs) typically include the following components
- **Decision Variables:** (i.e., independent variables or terms dependent on them) What you're going to change (directly or indirectly) to find the best value of the objective you can.
  - **Constraints:** Set allowed values of decision variables (i.e., allowed inputs, i.e., the “domain” of the objective function).
  - **Objective:** The thing you will optimize (maximize or minimize) (i.e., NV)
    - Note: technically the objective is also a decision variable, just not one that you directly decide.
  - **Objective Function:** A mathematical function that determines the objective as a function of the decision variables (i.e., the NVF).

# Components of an Optimization Model

→ Typical set up for a generalized optimization program:

$$\min_{\mathbf{x}} \phi = f(\mathbf{x})$$

← **Objective Function**

**s.t.**

← “Subject to”

$$h(\mathbf{x}) = 0$$

← Equality Constraints

$$g(\mathbf{x}) \leq 0$$

← Inequality Constraints

$$\mathbf{x}_{lb} \leq \mathbf{x} \leq \mathbf{x}_{ub}$$

← Variable Bounds

# Components of an Optimization Model

## →Parameters:

- Constants; terms you don't get to decide when optimizing

## →Decision Variables:

- **Independent Variables (i.e., “direct” decision variables)**
  - Things you can change directly, your inputs
  - Example: hourly wage and labour hours
- **Dependent Variables (i.e., “indirect” decision variables)**
  - Convenient combinations of independent variables that you might use as an intermediate term (or the objective itself).
  - Example: total labour cost (the product of hourly wage and labour hours)
  - Though it's possible to write the objective in terms of only independent variables, it likely leads to a very tough to read objective function where the physical meaning of terms is hidden

# Decision variables can be discrete or continuous

→ Discrete variables,  $x_D$

- Limited to a fixed or countable set of values (all-or-nothing, either-or, must be produced/purchased in integer quantities, *etc.*)

$$x_D \in I = \{0, 1, 2, \dots\}$$

→ Continuous variables,  $x_C$

- May take on *any value* in a specified interval

$$x_C \in S \subset \mathbb{R} \quad \text{Where “} \subset \text{” means “is a subset of”}$$

→ It is much easier to model with continuous variables than with discrete variables

→ If a discrete variable scale is large enough that rounding to the nearest integer causes *minimal loss of model accuracy*, then we can model that variable *continuously*

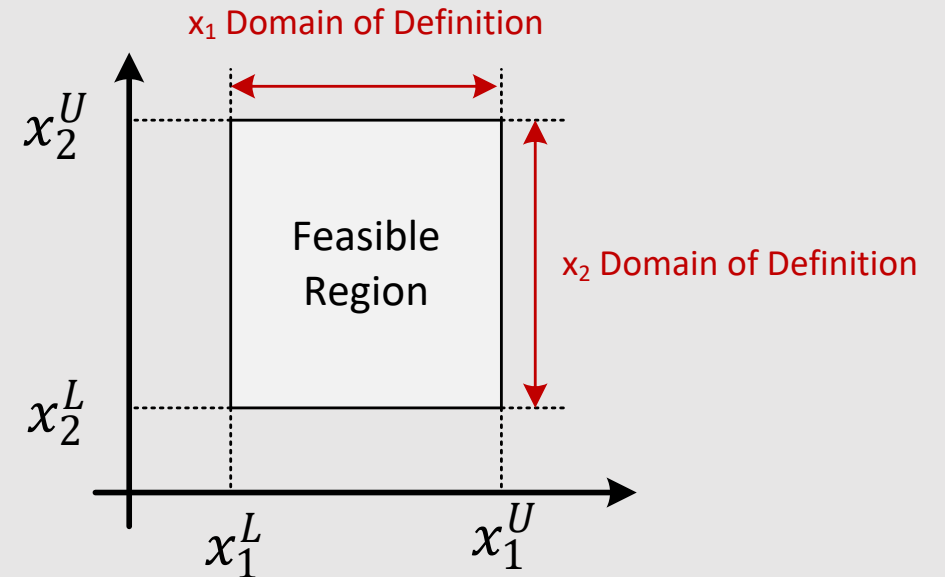
# Constraints

→ Decision variables are usually *bounded* (finite allowed range) because of:

- **Physical reality:** e.g., cannot purchase negative raw materials, cannot have a temperature less than 0 K
- **Model assumptions:** e.g., cannot sleep less than 6 hours per day, pipe ID is around 1/8"

→ The *domain of definition* is defined by a decision variable's upper ( $x^U$ ) and lower bounds ( $x^L$ )

- The combination of domains of definition for all decision variables is the **feasible region** or **feasible set**





# Graphing Model Constraints

→The feasible set (or region)  $\mathcal{S}$  of an optimization model is the collection of decision variables that satisfy *all* the model constraints:

$$\mathcal{S} \triangleq \{x : g(x) \leq 0, h(x) = 0, x^L \leq x \leq x^U\}$$

→The set of all points satisfying  $h(x) = 0$  results in a line or vector when plotting the feasible set.

→The set of all points satisfying  $g(x) \leq 0$  results in a region bounded above ( $\leq$ ) or below ( $\geq$ ) by a line defining the inequality.

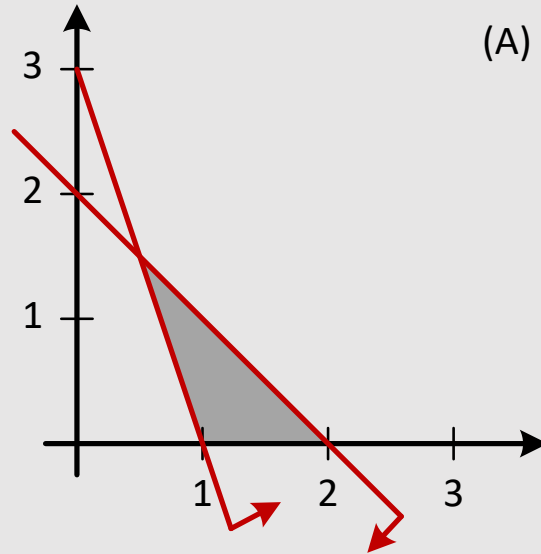
# Graphing Model Constraints Examples

*Constraint set (A)*

$$x_1 + x_2 \leq 2$$

$$3x_1 + x_2 \geq 3$$

$$x_1, x_2 \geq 0$$

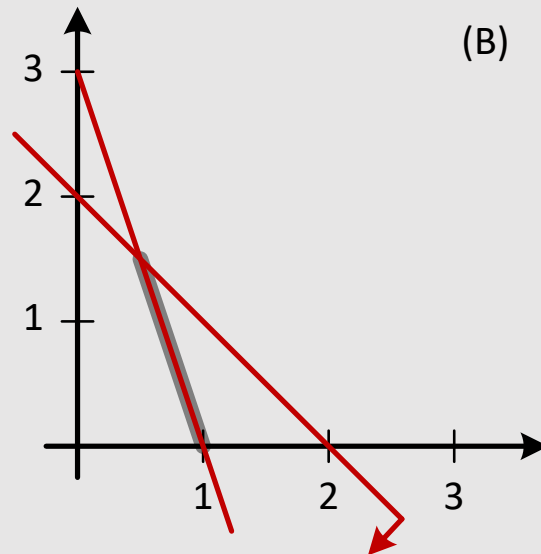


*Constraint set (B)*

$$x_1 + x_2 \leq 2$$

$$3x_1 + x_2 = 3$$

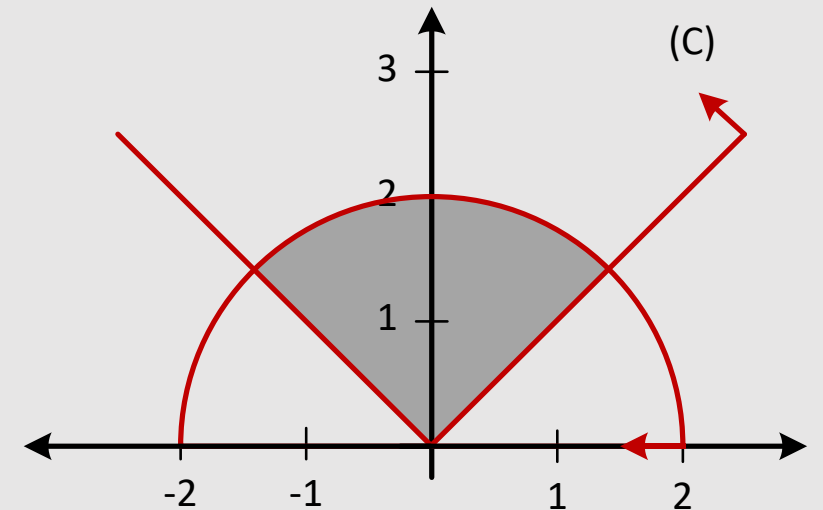
$$x_1, x_2 \geq 0$$



*Constraint set (C)*

$$x_1^2 + x_2^2 \leq 4$$

$$|x_1| - x_2 \leq 0$$



# Constraints + Objective Visualizing

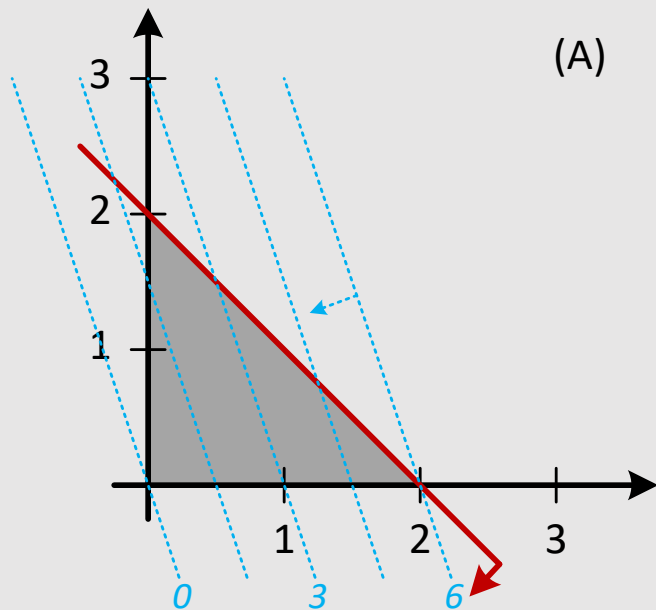
## Problem (A)

$$\min_{x_1, x_2} \phi = 3x_1 + x_2$$

s.t.

$$x_1 + x_2 \leq 2$$

$$x_1, x_2 \geq 0$$



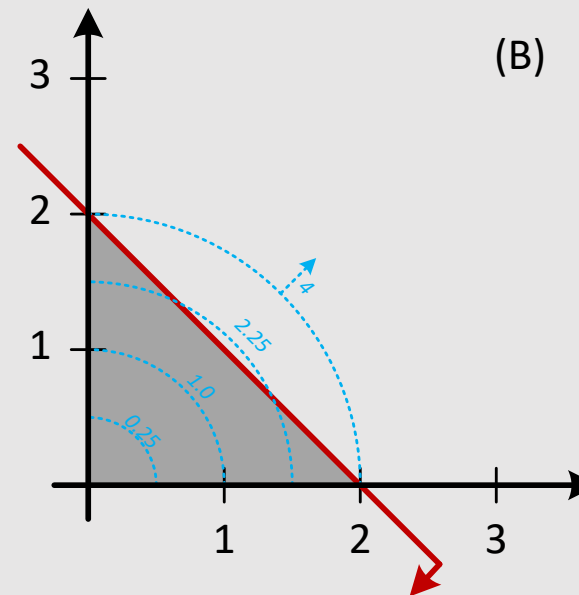
## Problem (B)

$$\max_{x_1, x_2} \phi = x_1^2 + x_2^2$$

s.t.

$$x_1 + x_2 \leq 2$$

$$x_1, x_2 \geq 0$$



# Optimization Outcomes

- Optimal solutions are points lying on the best objective function contour that intersects with at least one boundary of the feasible region.
- The *optimal value*  $\phi^*$  is defined to be the value of the objective at the optimum(s):  
 $\phi^* \triangleq \phi(x^*)$ .
- An optimization model can have only one true optimal value. It may have:
  - A **unique** optimal solution.
  - Several **alternative** solutions  $x^*$  yielding the *same* optimal  $\phi^*$ .
  - **No** optimal solutions (either the problem is unbounded or infeasible).

# Optimization Outcomes

Considering the same optimization problems as before:

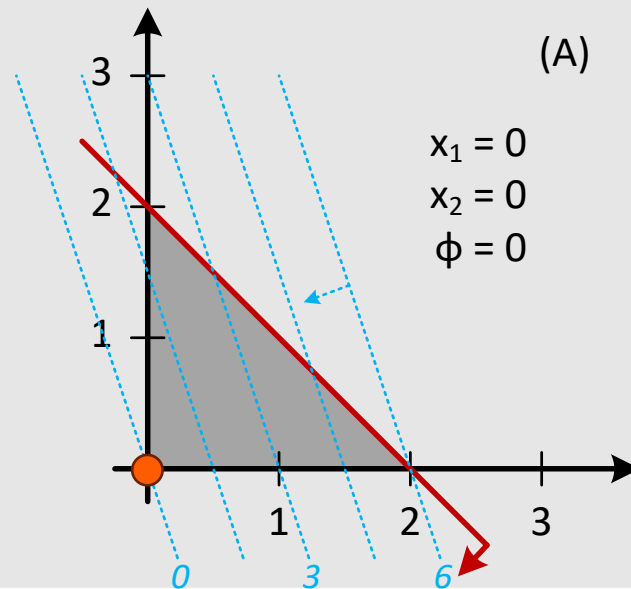
**Problem (A)**

$$\min_{x_1, x_2} \phi = 3x_1 + x_2$$

s.t.

$$x_1 + x_2 \leq 2$$

$$x_1, x_2 \geq 0$$



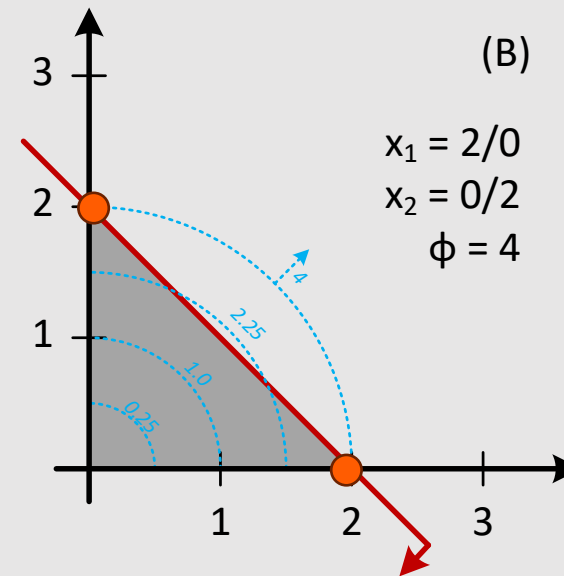
**Problem (B)**

$$\max_{x_1, x_2} \phi = x_1^2 + x_2^2$$

s.t.

$$x_1 + x_2 \leq 2$$

$$x_1, x_2 \geq 0$$



# Optimization Outcomes

→ Some optimization problems have degeneracy and unboundedness

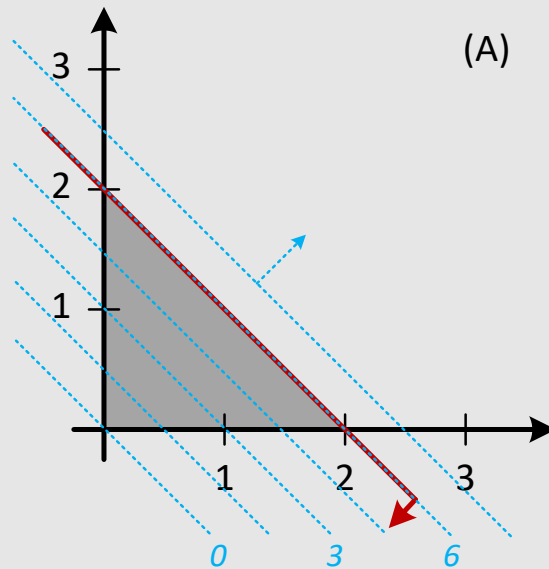
**Problem (A)**

$$\max_{x_1, x_2} \phi = 3x_1 + 3x_2$$

s.t.

$$x_1 + x_2 \leq 2$$

$$x_1, x_2 \geq 0$$



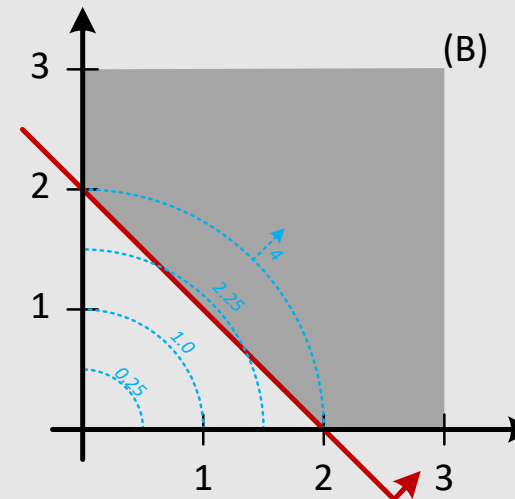
**Problem (B)**

$$\max_{x_1, x_2} \phi = x_1^2 + x_2^2$$

s.t.

$$x_1 + x_2 \geq 2$$

$$x_1, x_2 \geq 0$$



# Comments on Constraints

→ A constraint  $g(x) \leq 0$  is said to be:

- **Active** (or **binding**) at some point  $x^*$  if  $g(x^*) = 0$ .
- **Inactive** at some point  $x^*$  if  $g(x^*) < 0$ .

→ **Active constraints:**

- The set of constraints that are active at the optimal solution are known as the **active set**.
- Equality constraints are **ALWAYS** active at any feasible optimal point.
- No constraints (inequality or equality) may be violated at any optimal point.

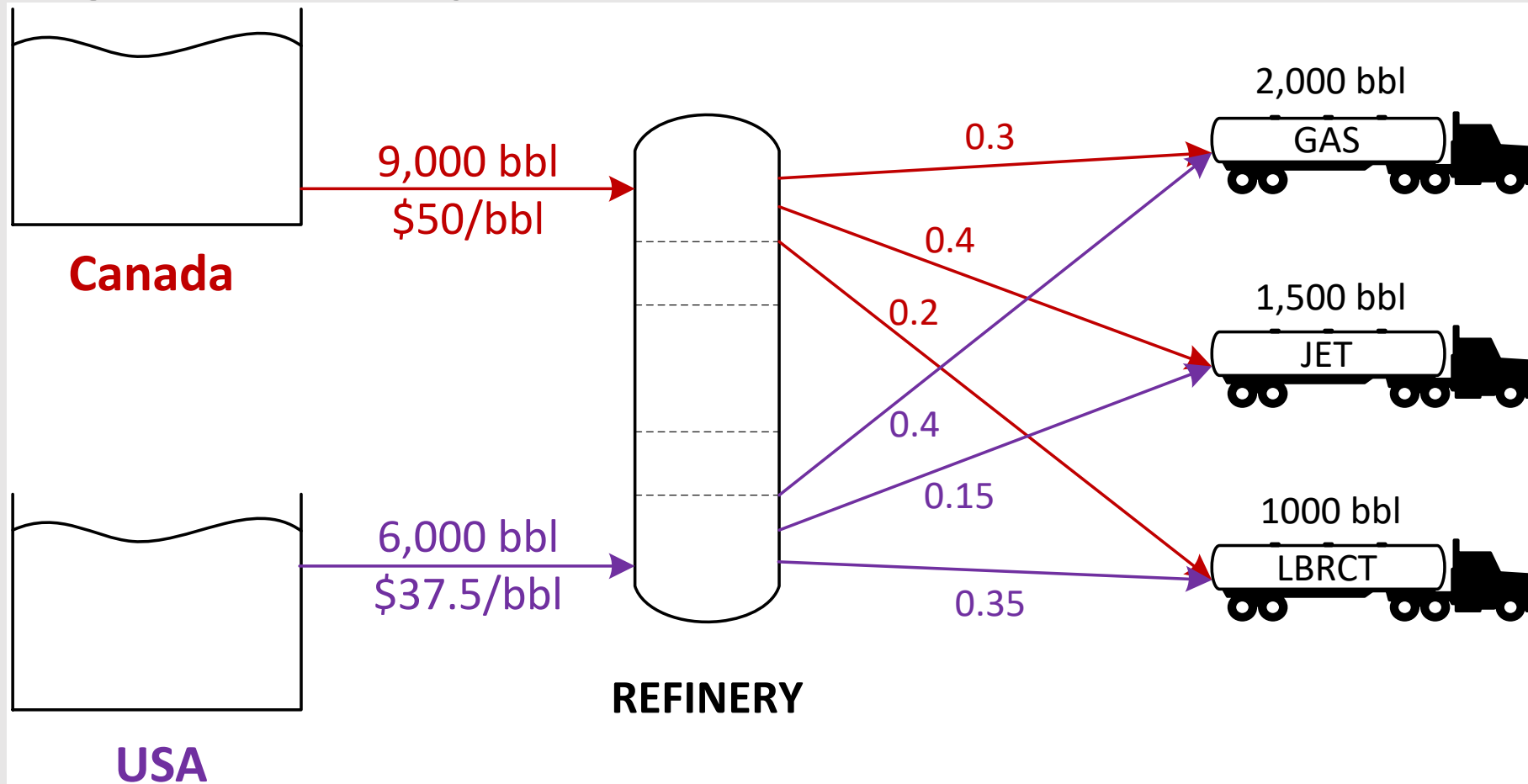
# Optimization Formulation Example

- A refinery distills crude petroleum into three products: gasoline, jet fuel, and lubricants.
- Your plant receives crude oil shipments from Canada and USA. You want to minimize the cost.
  - Each barrel from Canada yields 0.3 barrels of gasoline, 0.4 barrels of jet fuel, and 0.2 barrels of lubricants.
  - Each barrel from USA yields 0.4 barrels of gasoline, 0.15 barrels of jet fuel, and 0.35 barrels of lubricants.
  - The remaining 0.1 (10%) from both sources is lost to the refining process.
  - The Canadian oil costs your refinery \$50 per barrel and is available up to 9,000 barrels per day.
  - The American crude costs your refinery \$37.5 but only available up to 6,000 barrels per day.
- You have a contract with local distributors to provide per day
  - 2,000 barrels of gasoline
  - 1,500 barrels of jet fuel
  - 1000 barrels of lubricant



# Linear Optimization Example

→ Draw a diagram of the supply network



# Optimization Formulation Example

→ Formulate the optimization problem:

$$\begin{array}{llll} \min_{x_1, x_2} \phi & = & 50x_1 + 37.5x_2 & \\ & \text{Subject to} & & \\ 0.3x_1 + 0.4x_2 & \geq & 2,000 & \\ 0.4x_1 + 0.15x_2 & \geq & 1,500 & \\ 0.2x_1 + 0.35x_2 & \geq & 1,000 & \\ x_1 & \leq & 9,000 & \\ x_2 & \leq & 6,000 & \\ x_i & \geq & 0 & (\forall i) \end{array}$$

# Optimization Example

→ For this problem, we can start with a graphical model:

- Plot the constraints and objective function
- Visually identify the optimal value and feasible set

