

# Formulating an Optimization Problem

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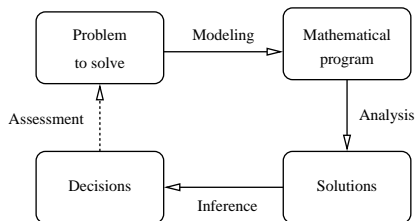
ChE 4G03: Optimization in Chemical Engineering

## Outline

- 1 The Importance of a Good Formulation
- 2 The Standard Formulation
- 3 Graphic Solution and Optimization Outcomes

## The Importance of a Good Formulation

### Model-based Optimization Procedure:



- 1 The conclusions are drawn from the model (mathematical program), **not** from the problem!
- 2 An **inadequate model** typically leads to **false conclusions**
- 3 The model must be **computationally tractable**: its analysis must be practical!

### Dofasco:

“We find it useful to formulate the optimization problem even if we cannot solve it...”

## Validity vs. Tractability

- 1 The **validity** of a model is the degree to which inferences drawn from the model hold for the real system
- 2 The **tractability** of a model is the degree to which this model admits convenient analysis – How much analysis is practical

### Tradeoff:

Decision makers almost always confront a **tradeoff** between **validity** of optimization models and their **tractability** to analysis

Very accurate  
over wide range  
of conditions  
Longer  
computations  
More complex

Less accurate  
over narrow range  
of conditions  
Shorter  
computations  
Less complex

# Optimization Model Formulation

## Standard Model:

Optimization models (also called **mathematical programs**) represent problem choices as **decision variables** and seek values that maximize (or minimize) **objective functions** of the decision variables subject to **constraints** on variable values expressing the limits on possible decision choices

$$\begin{aligned} \max_{\mathbf{x}} f(\mathbf{x}) & \quad \leftarrow \text{Objective function} \\ \text{s.t.} & \\ \mathbf{h}(\mathbf{x}) = \mathbf{0} & \quad \leftarrow \text{Equality constraints} \\ \mathbf{g}(\mathbf{x}) \leq \mathbf{0} & \quad \leftarrow \text{Inequality constraints} \\ \mathbf{x}_{\min} \leq \mathbf{x} \leq \mathbf{x}_{\max} & \quad \leftarrow \text{Variable bounds} \end{aligned}$$

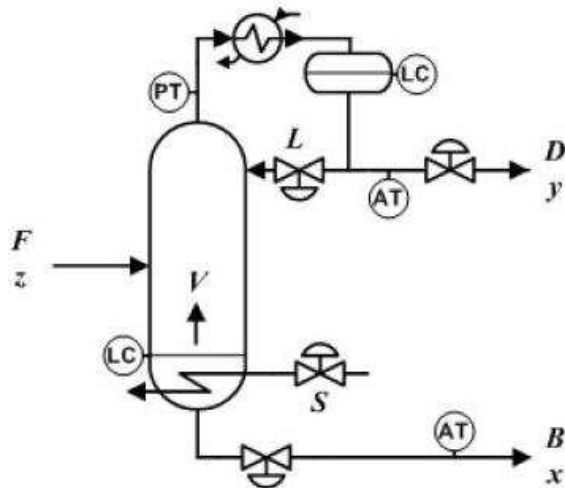
# Decision Variables, x

- At the **formulation stage**, decision variables can be grouped into two categories:
  - Independent variables**, whose values can be changed *independently* to modify the behavior of the system
  - Dependent variables**, whose behavior is determined by the values selected for the independent variables
 Such a grouping helps understanding!
- At the **solution stage**, independent and dependent variables need **not** be **distinguished**: the solution method just sees an optimization problem involving many variables

## Examples of Decision Variables:

- Design**: reactor volume, number of trays, heat exchanger area, etc.
- Operations**: temperature, flow, pressure, valve opening, etc.
- Management**: feed type, purchase price, sales price, etc.

# Decision Variables, x (cont'd)



**Exercise:**  
Identify **independent** and **dependent** variables

# Decision Variables, x (cont'd)

Discrete vs. Continuous Decision Variables:

- A variable is **discrete** if it is limited to a fixed or countable set of values (e.g., all-or-nothing, either-or decisions)
- A variable is **continuous** if it can take on any value in a specified interval

Problems with both discrete and continuous variables are called **mixed-integer programs**

## Heuristics:

When there is an option, modeling with **continuous variables is preferred to discrete variables** (for tractability reasons)

**Example:** Variable magnitudes are likely to be large enough that fractions have no practical importance

## Decision Variables, $x$ (cont'd)

**Exercise:** Decide whether a discrete or a continuous variable would be best employed to model each of the following quantities:

- 1 the optimal temperature of a chemical process
- 2 the warehouse slot assigned a particular product
- 3 whether a capital project is selected for investment
- 4 ball bearings in a plant that manufactures 10,000+/day
- 5 the number of aircraft produced on a defense contract

## Decision Variables, $x$ (cont'd)

**Exercise:**

A large manufacturer of corn seed operates  $\ell = 20$  facilities producing seeds of  $m = 25$  hybrid corn varieties and distributes them to customers in  $n = 30$  regions. The optimization problem is to carry out these production and distribution operations at minimum cost.

Let the three main dimensions in this problem be denoted as:

- $f$  = production facility number ( $f = 1, \dots, \ell$ )
- $v$  = hybrid variety number ( $v = 1, \dots, m$ )
- $r$  = sales region number ( $r = 1, \dots, n$ )

Questions:

- 1 Choose appropriate decision variables for a model of this problem (use indexing).
- 2 What is the total number of decision variables for this model?

## Decision Variables, $x$ (cont'd)

**Indexing:**

In real applications, optimization models quickly grow to thousands, sometimes millions, of variables: **indexed notational schemes** are needed to keep large models manageable!

- 1 **Indexes** (or **subscripts**) permit representing collections of similar quantities with a single symbol, e.g.

$$\{x_i : i = 1, \dots, 100\}$$

represents 100 similar values with the same  $x$  name, distinguishing them with the index  $i$

- 2 The **first step** in formulating a large optimization model is to choose appropriate indexes for the different dimensions of the problem; **multiple indexes** are extremely common!

## Variable Bounds, $x_{\min}$ , $x_{\max}$

**Variable bounds** specify the domain of definition for decision variables: the set of values for which the variables have meaning

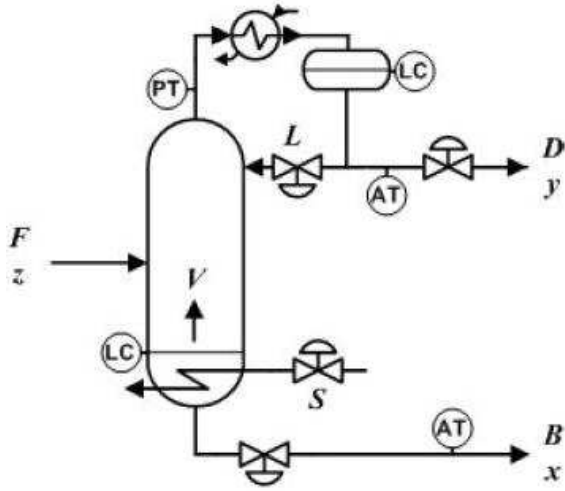
- 1 Setting the min and max values equal sets the variable to a constant value:

$$x_{\min,i} = x_{\max,i} = x_i, \quad \text{for some } i \in \{1, \dots, n\}$$

- 2 The most common variable bound constraint is **non-negativity**, e.g.

$$x_i \geq 0, \quad \forall i = 1, \dots, n$$

## Variable Bounds, $x_{\min}$ , $x_{\max}$ (cont'd)



**Exercise:**  
Propose bounds for variables in the process

## Main constraints, $g(x) \leq 0$ , $h(x) = 0$

**Main constraints** of optimization models specify the restrictions and interactions, other than variable bounds, that limit decision variable values.

### Inequality Constraints:

$$g(x) \leq 0$$

- 1 These are **one-way limits** on the system and are **essential** for optimization
- 2 Inequality constraints of the form  $\geq 0$  and  $\leq 0$  are equivalent. Do you see why?
- 3 There can be many of these inequalities, so that  $g(x)$  is a vector (**indexing** is used for constraints too!)
- 4 We must be careful to prevent defining a problem incorrectly with **no feasible region** or an **unbounded solution**

## Main constraints, $g(x) \leq 0$ , $h(x) = 0$ (cont'd)

### What limits the possible solutions to the problem?

- Safety
- Product quality (contracts)
- Equipment damage (long term)
- Equipment operation
- Legal/ethical considerations

### Examples:

- Max. investment available
- Max. flow rate due to pump limit
- Min. liquid flow rate on trays  $t = 1, \dots, N_t$
- Max. pressure in a closed vessel
- Max. region within which an approximation/simplification is acceptable

## Main constraints, $g(x) \leq 0$ , $h(x) = 0$ (cont'd)

### Equality Constraints:

$$h(x) = 0$$

- 1 These describe **interactions** between the variables in the model
- 2 Equality constraints are written with a **zero right-hand side** by convention
- 3 There can be many of these inequalities, so that  $h(x)$  is a vector (**indexing** is used for equality constraints too!)
- 4 There **cannot** be more (independent) equality constraints than decision variables in the model! Do you see why?

## Main constraints, $\mathbf{g}(\mathbf{x}) \leq \mathbf{0}$ , $\mathbf{h}(\mathbf{x}) = \mathbf{0}$ (cont'd)

### Examples of Equality Constraints:

- material, energy, current, etc. **balances**, e.g. for conserved quantity  $q$ :

$$\left\{ \begin{array}{c} \text{accumulation} \\ \text{of } q \end{array} \right\} = \left\{ \begin{array}{c} \text{rate of} \\ q \text{ in} \end{array} \right\} - \left\{ \begin{array}{c} \text{rate of} \\ q \text{ out} \end{array} \right\} + \left\{ \begin{array}{c} \text{generate} \\ \text{rate of } q \end{array} \right\}$$

- constitutive** relations, e.g.

$$Q = hA\Delta T$$
$$r_A = k_0 \exp\left(-\frac{E}{RT}\right)$$

- equilibrium** relations, e.g. VLE
- decisions** by the engineer, e.g.  $F_A - 2F_B = 0$
- behavior enforced by **controllers**, e.g. temperature set-point  $T_s = 231 \text{ K}$

## Objective Function, $f(\mathbf{x})$

Objective functions in optimization models tell us **how to rate decisions**

- We need a **quantitative** measure; a qualitative measure such as “good” or “bad” is **not** adequate!
- A **scalar objective** function is preferred for solving, even though multiple objectives are typical in real life!
- There is no fundamental or practical difference between max and min problems:

$$\max_{\mathbf{x}} f(\mathbf{x}) \Leftrightarrow \min_{\mathbf{x}} -f(\mathbf{x})$$

## Objective Function, $f(\mathbf{x})$

### How do we define a scalar that represents performance?

- Maximize profit (or minimize cost) – Economics
- Maximize product quality
- Minimize energy use
- Minimize polluting effluents
- Maximize safety
- etc.

**Exercise:** Write the expression for an objective function  $f(\mathbf{x})$  that depends on *all* variables  $x_i$ ,  $i = 1, \dots, n$ , and the cost associated with each variable is  $c_j$ .

- Express the answer as a summation of indexed variables
- Express the answer as a product of vectors

## Case Study: Model Formulation

A refinery distills crude petroleum from **two** sources, Saudi Arabia and Venezuela, into **three** main products: gasoline, jet fuel and lubricants. The two crudes differ in chemical composition and thus yield different product mixes (the remaining 10% of each barrel is lost to refining):

- Each barrel of Saudi crude yields 0.3 barrel of gasoline, 0.4 barrel of jet fuel, and 0.2 barrel of lubricants;
- Each barrel of Venezuelan crude yields 0.4 barrel of gasoline, 0.2 barrel of jet fuel, and 0.3 barrel of lubricants.

The crudes also differ in cost and availability:

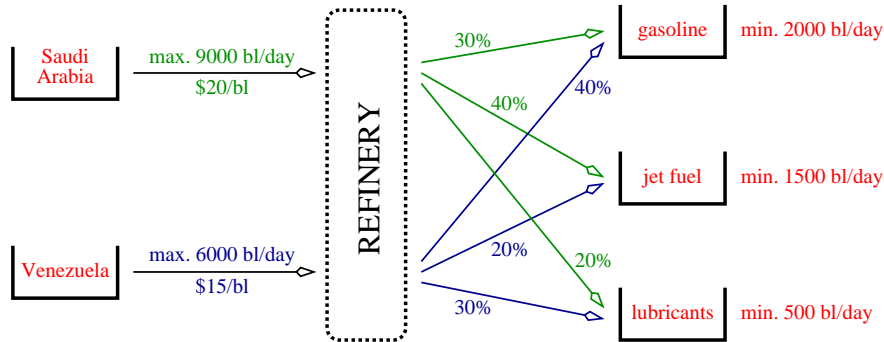
- Up to 9,000 barrels per day of Saudi crude are available at the cost \$20 per barrel;
- Up to 6,000 barrels per day of Saudi crude are also available at the lower cost \$15 per barrel.

Contracts with independent distributors require that the refinery produce 2,000 barrels per day of gasoline, 1,500 barrels per day of jet fuel, and 500 barrels per day of lubricants.

## Case Study: Model Formulation (cont'd)

### Question:

Formulate an optimization model in standard form so as to fulfill the requirements in the most efficient manner.



## Graphing Constraints

- Very simple graphic solution have enough power to deal with **tiny models**, such as 2- or 3-variable models
- The first issue in graphic solution is the **feasible set**:

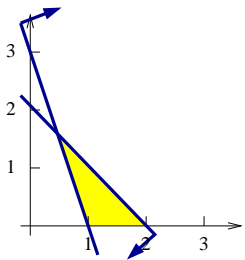
The feasible set  $S$  (or feasible region) of an optimization model is the collection of choices for decision variables satisfying **all** model constraints:

$$S \triangleq \{ \mathbf{x} : \mathbf{g}(\mathbf{x}) \leq \mathbf{0}, \mathbf{h}(\mathbf{x}) = \mathbf{0}, \mathbf{x}_{\min} \leq \mathbf{x} \leq \mathbf{x}_{\max} \}$$

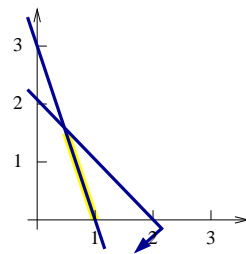
- The set of all points satisfying an equality constraint plots as a line or a curve (in 2-d)
- The set of all points satisfying an inequality constraint plots as a boundary line or curve, where the constraint holds with equality, along with all points on whichever side of the boundary satisfy the constraint as an inequality

## Examples of Feasible Regions

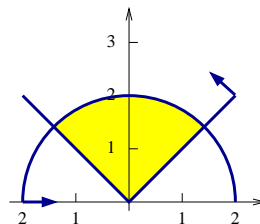
$$\begin{aligned} x_1 + x_2 &\leq 2 \\ 3x_1 + x_2 &\geq 3 \\ x_1, x_2 &\geq 0 \end{aligned}$$



$$\begin{aligned} x_1 + x_2 &\leq 2 \\ 3x_1 + x_2 &= 3 \\ x_1, x_2 &\geq 0 \end{aligned}$$



$$\begin{aligned} (x_1)^2 + (x_2)^2 &\leq 4 \\ |x_1 - x_2| &\leq 0 \end{aligned}$$



**Exercise:** Plot the feasible region in the two-crude case study.

## Graphing Objective Functions

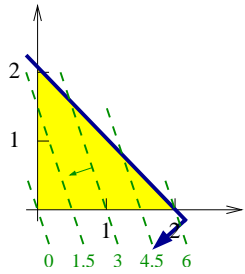
- To find out the best feasible point, the objective function must be introduced into a plot
- Objective functions are normally plotted in the same coordinate system as the feasible set by introducing **contours**:

The contour  $C_z$  of an objective function (in the decision variable space) is the line or curve passing through points having **equal objective values**  $z$ :

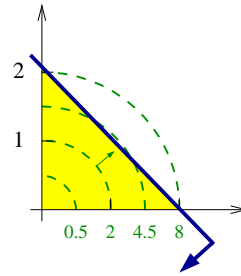
$$C_z \triangleq \{ \mathbf{x} : f(\mathbf{x}) = z \}$$

## Examples of Graphic Solutions

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



$$\begin{aligned} \max_{x_1, x_2} \quad & (x_1)^2 + (x_2)^2 \\ \text{s.t.} \quad & x_1 + x_2 \leq 2 \\ & x_1, x_2 \geq 0 \end{aligned}$$



**Exercise:** Plot the objective function contours in the two-crude case study.

## Optimization Outcomes

### Optimum

An **optimal solution**  $\mathbf{x}^*$  is a feasible choice for decision variables with objective function value at least equal to that of any other solution satisfying all constraints. For a minimization problem:

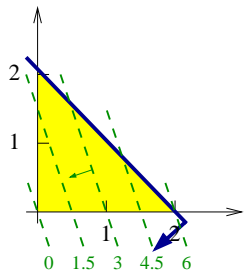
$$f(\mathbf{x}^*) \leq f(\mathbf{x}), \quad \forall \mathbf{x} \in S$$

- 1 Optimal solutions show graphically as points lying on the best objective function contour that intersects the feasible region
- 2 The **optimal value**  $f^*$  in an optimization model is the objective function value of any optimal solutions:  $f^* = f(\mathbf{x}^*)$
- 3 An optimization model can have **only one** optimal value
- 4 An optimization model may have:
  - ▶ a **unique** optimal solution
  - ▶ several **alternative** optimal solutions
  - ▶ **no** optimal solutions (unbounded or infeasible models)

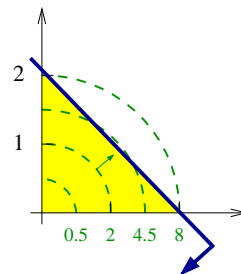
## Examples of Graphic Solutions

What is the **optimal solution**? What is the **optimal value**?

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



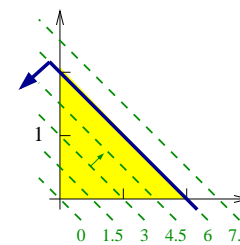
$$\begin{aligned} \max_{x_1, x_2} \quad & (x_1)^2 + (x_2)^2 \\ \text{s.t.} \quad & x_1 + x_2 \leq 2 \\ & x_1, x_2 \geq 0 \end{aligned}$$



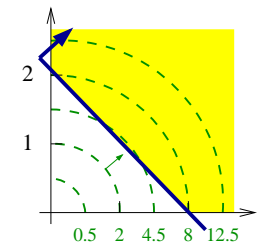
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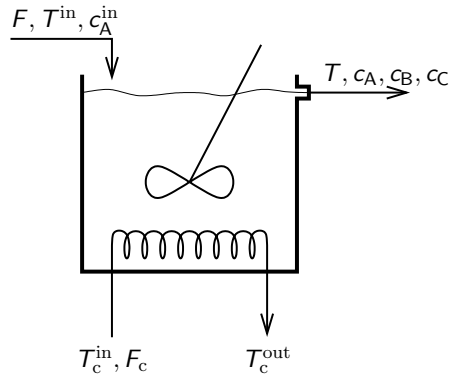
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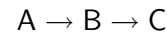
**Exercise:** Determine a graphic solution to the two-crude case study.

## Recognizing Optimization Opportunity

**Class Exercise:** What is optimum for the following process system?



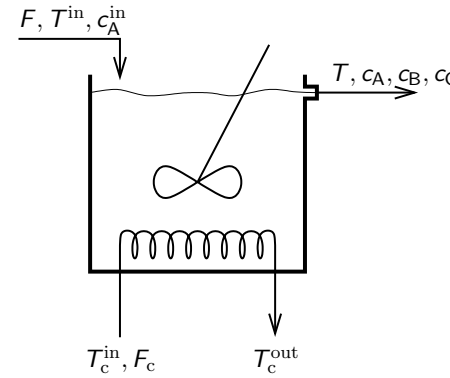
This is an isothermal CFSTR with the series reaction:



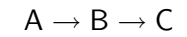
- The goal is to **maximize  $C_B$**  in the effluent at steady state
- Only the flow rate of feed  $F$  can be adjusted

## Recognizing Optimization Opportunity

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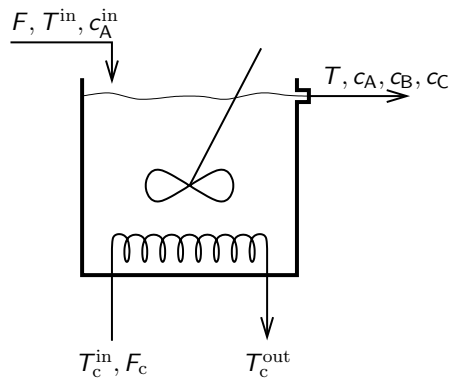
This is an isothermal CFSTR with the series reaction:



- The goal is to **maximize  $C_C$**  in the effluent at steady state
- Only the flow rate of feed  $F$  can be adjusted

## Recognizing Optimization Opportunity

**Class Exercise:** What is optimum for the following process system?



This is an isothermal CFSTR with the parallel reaction:



- The goal is to **maximize  $C_B$**  in the effluent at steady state
- Only the flow rate of feed  $F$  can be adjusted
- What other variable would you like to adjust?