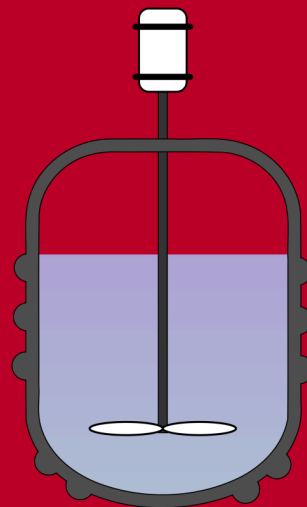


Carnegie Mellon University



# Introductory Numerical Methods for Simulating Batch Reactors

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**Dr. Joshua Pulsipher**



# Learning Outcomes

1. The relative advantages/disadvantages of using **explicit Euler** methods
2. How to implement explicit Euler to **simulate ODEs** using common computation environments (e.g., Python)
3. How to **simulate batch reactors** using numerical methods
4. A familiarity of **other numerical methods/tools** for simulating batch reactors



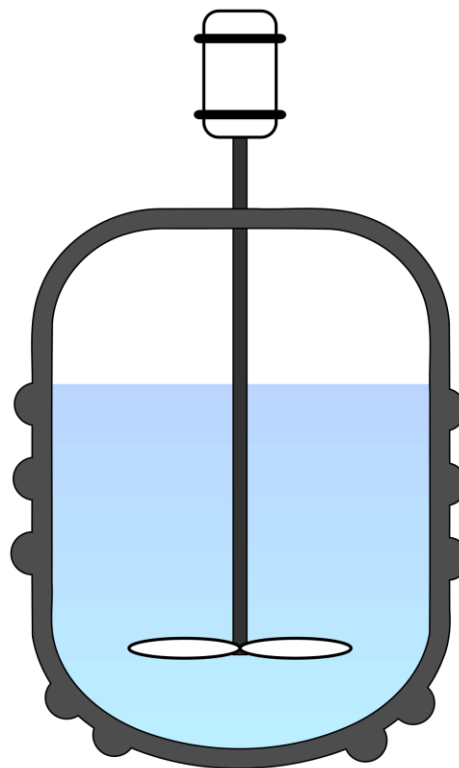
# Outline

- Overview
- Explicit Euler
- Batch Reactors
- Other Numerical Methods



# Outline

- **Overview**
- Explicit Euler
- Batch Reactors
- Other Numerical Methods





# Simulating Dynamic Systems

- Simulating dynamic systems is vital for enabling engineering applications



Curiosity Rover



Batch Reactor

- Simulate using **numerical methods to approximate dynamics** (e.g., differential equations)
- Enables us to **computationally experiment** and implement **automation**



# Differential Equations

## Types

- Ordinary differential equations (ODEs)

- Today's focus

$$\frac{dy(t)}{dt} = f(y(t), t)$$
$$y(0) = y_0$$

- Partial differential equations (PDEs)

$$\frac{\partial y_c(t, x)}{\partial t} = \xi(x) \left( \frac{\partial^2 y_c(t, x)}{\partial x_1^2} + \frac{\partial^2 y_c(t, x)}{\partial x_2^2} \right) + y_g(t, x)$$
$$y_c(0, x), y_c(t, \text{boundary}) = 0$$

## Applications

- Transient flow balance

$$\frac{df(t)}{dt} = f_{in}(t) - f_{out}(t) + f_{gen}(t)$$

- Heat/mass transfer

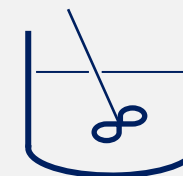
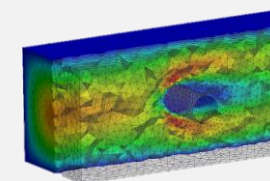
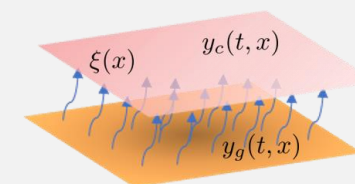
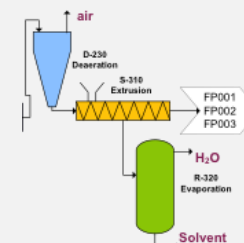
$$\frac{\partial T(t)}{\partial t} - \alpha \frac{\partial^2 T(t)}{\partial x^2} = 0$$

- Fluid flow

$$\rho \frac{DV(t, x)}{Dt} = -\nabla p + \rho g(x) + \mu \nabla^2 V(t, x)$$

- Kinetics

$$\frac{dc(t)}{dt} = kc(t)^\alpha$$





# Analytical vs. Numerical Methods

## Analytical Methods

- Separate and integrate

$$g(y) \frac{dy}{dx} = h(x) \quad \Rightarrow \quad \int g(y) dy = \int h(x) dx + C$$

- ODEs of special forms

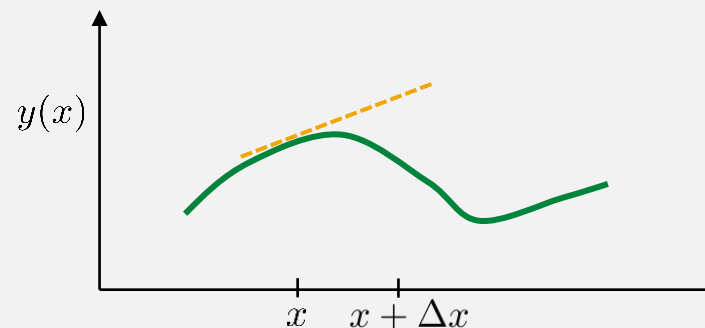
$$\frac{dy}{dx} = \frac{x+y}{x-y} \quad \Rightarrow \quad \frac{1}{2} \log \left( \frac{y^2}{x^2} + 1 \right) - \tan^{-1} \left( \frac{y}{x} \right) = C - \log(x)$$

- Solving general ODEs is often difficult or not possible

## Numerical Methods

- Seek to numerically approximate the solution

- Finite difference methods** are common



- More advanced methods are available
  - Not the focus of today



# My Teaching Philosophy

**Idea:** Promote a **tutorial-like** format that encourages active engagement.

## Active Learning

- Mastery comes through **deliberate practice**
- Magnify class time to gain **guided hands-on experiences**



## In-Class Exercises

- We will be using Jupyter notebooks with a Julia and/or a Python kernel today
- No downloads/installation are/is needed
- <https://pulsipher.info/teaching/courses.html>

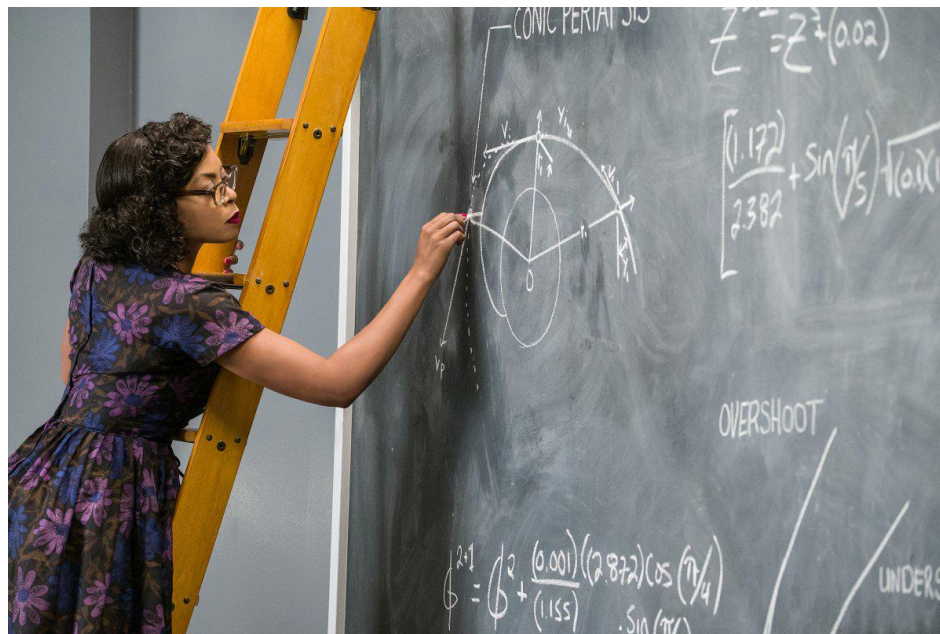






# Outline

- Overview
- **Explicit Euler**
- Batch Reactors
- Other Numerical Methods





# The Basics

## Methodology

- Consider a 1<sup>st</sup> order ODE

$$\frac{dy(t)}{dt} = f(y(t), t)$$
$$y(0) = y_0$$

- Define time steps  $\Delta t$

$$t \in [t_0, t_f] \quad t_k = t_0 + k\Delta t$$

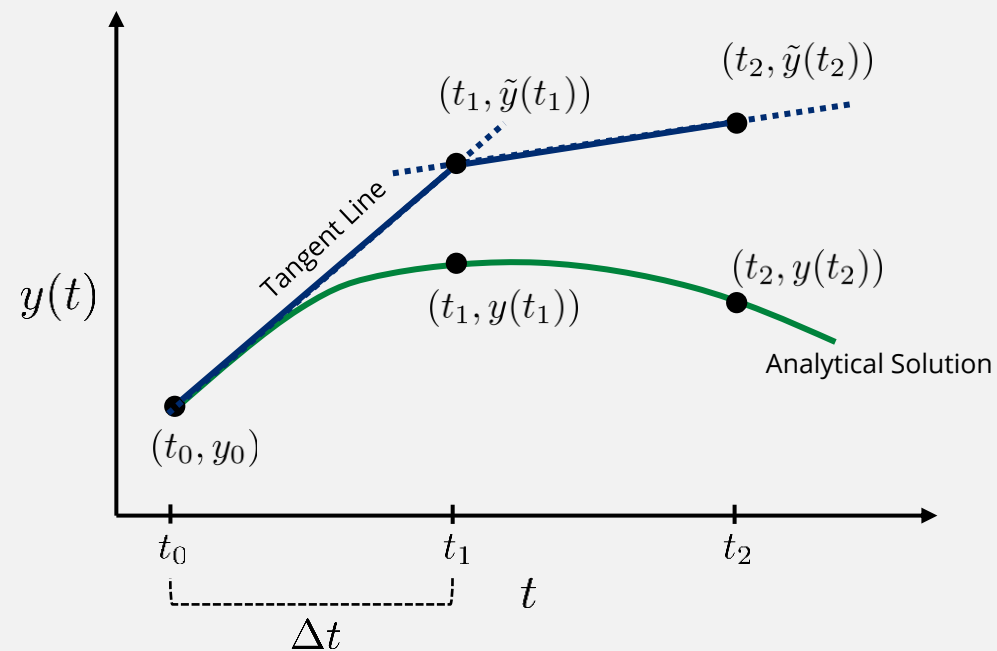
- Approximate derivative as finite difference

$$\left. \frac{dy(t)}{dt} \right|_{t_k} \approx \frac{\tilde{y}(t_{k+1}) - \tilde{y}(t_k)}{\Delta t}$$

- Define update rule

$$\tilde{y}(t_{k+1}) = \tilde{y}(t_k) + f(y(t_k), t_k)\Delta t$$

## Illustration





# Properties: Error

## Local Truncation Error (LTE)

- Recall update rule

$$\tilde{y}(t_{k+1}) = \tilde{y}(t_k) + f(y(t_k), t_k) \Delta t$$

- Taylor series expansion of analytic solution

$$y(t_k + \Delta t) = y(t_k) + \Delta t \left. \frac{dy(t)}{dt} \right|_{t_k} + O(\Delta t^2)$$

- Difference w/ explicit Euler

$$y(t_k + \Delta t) - \tilde{y}(t_{k+1}) = O(\Delta t^2)$$

- Hence, the error incurred after one step is

$$O(\Delta t^2)$$

## Global Truncation Error (GTE)

- The number of steps

$$\frac{t - t_0}{\Delta t} \propto \frac{1}{\Delta t}$$

- Multiplying this with the LTE, we get GTE that is

$$O(\Delta t)$$

- Hence, explicit Euler is a **first order method**

- Higher order methods are available



# Tutorial 1

## Problem Setup

- Solve the ODE

$$\frac{dy(t)}{dt} = e^{-t}$$
$$y(0) = -1$$

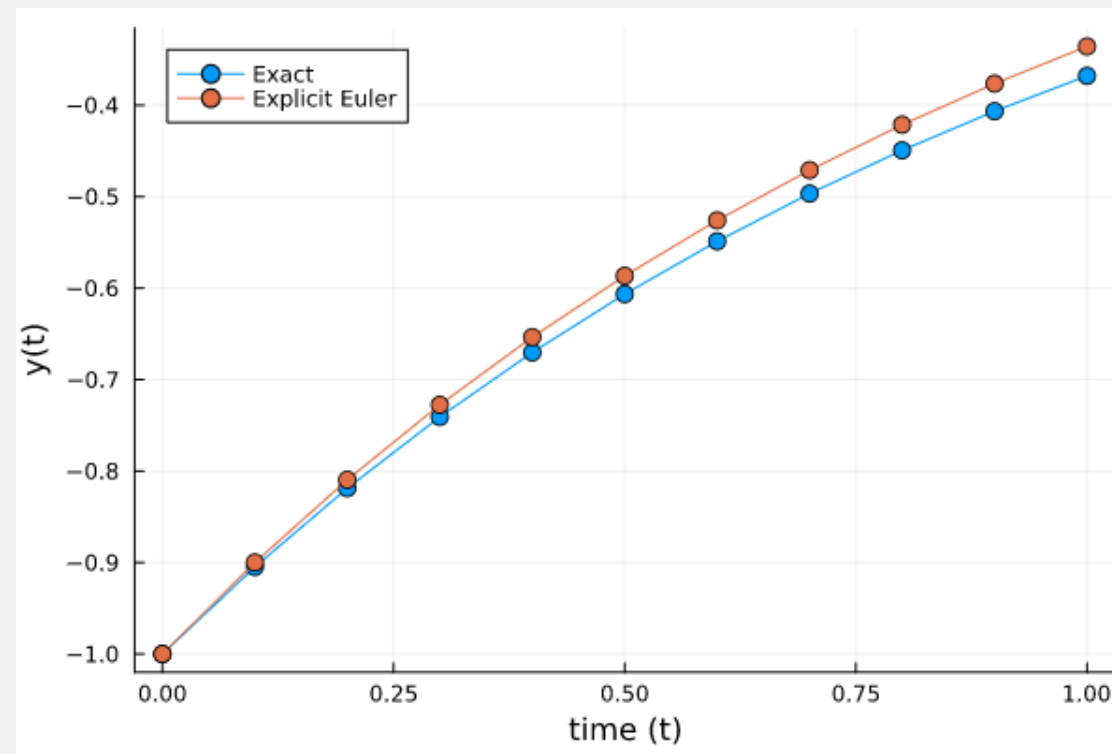
- Specifications

- $t \in [0, 1]$
- $\Delta t = 0.1$

- Plot the result against the analytical answer

- Experiment with varied  $\Delta t$

## Expected Answer





# Simulating a System of ODEs

## System of 1<sup>st</sup> order ODEs

- General representation

$$\frac{dy_1(t)}{dt} = f_1(y_1, y_2, \dots, y_n)$$

$$\frac{dy_2(t)}{dt} = f_2(y_1, y_2, \dots, y_n)$$

$\vdots$

$$\frac{dy_n(t)}{dt} = f_n(y_1, y_2, \dots, y_n)$$

$$y_1(0) = y_{1,0}, y_2(0) = y_{2,0}, \dots, y_n(0) = y_{n,0}$$

- Vectorize

$$\frac{d\mathbf{y}(t)}{dt} = \mathbf{f}(\mathbf{y}(t), t)$$

$$\mathbf{y}(0) = \mathbf{y}_0$$

We can represent a higher order ODE as a 1<sup>st</sup> order system

## Vectorized Explicit Euler

- Update rule uses vectorized representation

$$\tilde{\mathbf{y}}(t_{k+1}) = \tilde{\mathbf{y}}(t_k) + \mathbf{f}(\mathbf{y}(t_k), t_k) \Delta t$$

- Tutorial 2:** Simulate coupled ODEs w/  $\Delta t = 0.01$  and  $t \in [0, 1]$

$$\frac{dy_1(t)}{dt} = -5y_1(t) + 5y_2(t)$$

$$\frac{dy_2(t)}{dt} = 14y_1(t) - 2y_2(t)$$

$$y_1(0) = y_2(0) = 1$$



# Properties: Stability

## Example 1

- Simulate ODE w/  $\Delta t = 0.1$  in  $t \in [0, 1]$

$$\frac{dy(t)}{dt} = -20y(t)$$
$$y(0) = 1$$

- Compare w/ analytic answer

## Linear Stability

- Consider the linear ODE

$$\frac{dy(t)}{dt} = \lambda y(t)$$

- For a stable solution we must have

$$|1 + \lambda \Delta t| < 1$$

## Stiff ODEs

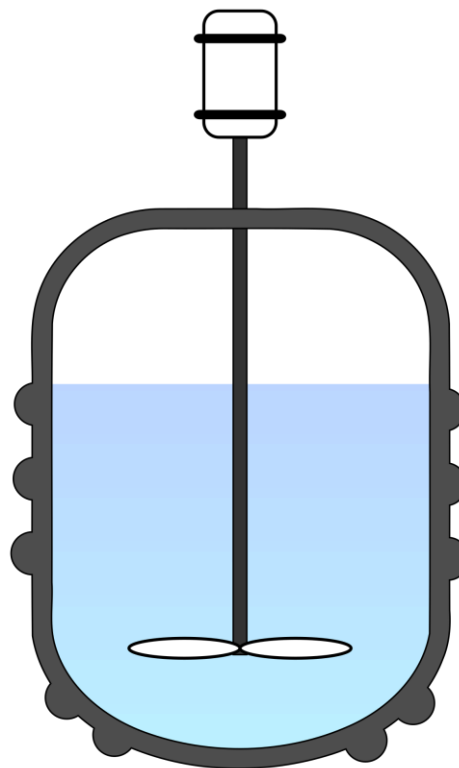
- Systems that exhibit **numerical instability**
- Precise mathematical definition is nontrivial
- Common with **reaction systems**
  - Coexistence of small and large rate constants
- So, what can we do? → Use **implicit methods**

$$\tilde{y}(t_{k+1}) = \tilde{y}(t_k) + f(y(t_{k+1}), t_{k+1}) \Delta t$$



# Outline

- Overview
- Explicit Euler
- **Batch Reactors**
- Other Numerical Methods





# Batch Reactor Modeling

## Simple ODE Model

- Arrhenius equation for species  $i$  and reaction  $j$

$$k_{ij}(t) = A_{ij} \exp\left(\frac{-E_{a,ij}}{RT(t)}\right)$$

- Reaction rates

$$r_j(\mathbf{c}, t) = \sum_{i \in I} k_{ij}(t) c_i^{\beta_{ij}}(t)$$

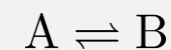
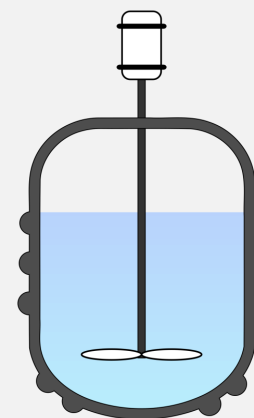
- Species generation rates

$$r_i(\mathbf{c}, t) = \sum_{j \in J} \gamma_{ij} r_j(\mathbf{c}, t)$$

- Species balances

$$\begin{aligned} \frac{dc_i(t)}{dt} &= r_i(\mathbf{c}, t), \quad i \in I \\ c_i(0) &= c_{i,0}, \quad i \in I \end{aligned}$$

## The Basics



- Derivation from mole balance

$$\frac{N_i(t)}{dt} = \cancel{F_{i,in}(t)} - \cancel{F_{i,out}(t)} + \int^V r_i(t) dV(t)$$

- Assume perfect mixing and constant volume

$$\frac{N_i(t)}{dt} = r_i(t)V \quad \rightarrow \quad \frac{c_i(t)}{dt} = r_i(t)$$

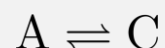
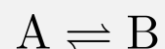




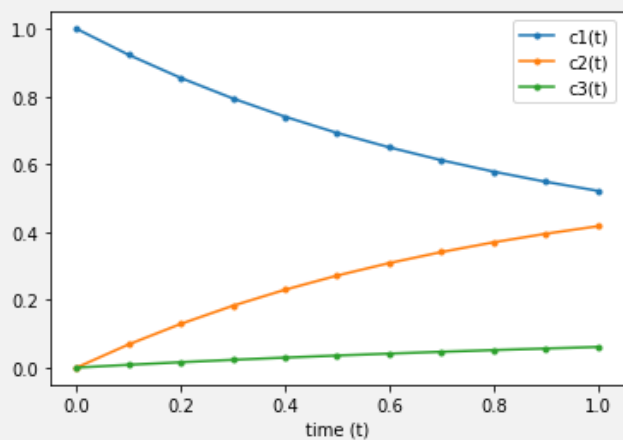
# Exercise: Batch Reactor

## Exercise 1

- Simulate the following reaction system using explicit Euler for  $t \in [0, 1]$



- Experiment with different choices of  $\Delta t$



## Problem Information

$$R = 1.987$$

$$A = \begin{bmatrix} 3.6362e6 & 190.6879 \\ -2.5212e16 & 0 \\ 0 & -8.7409e24 \end{bmatrix}$$

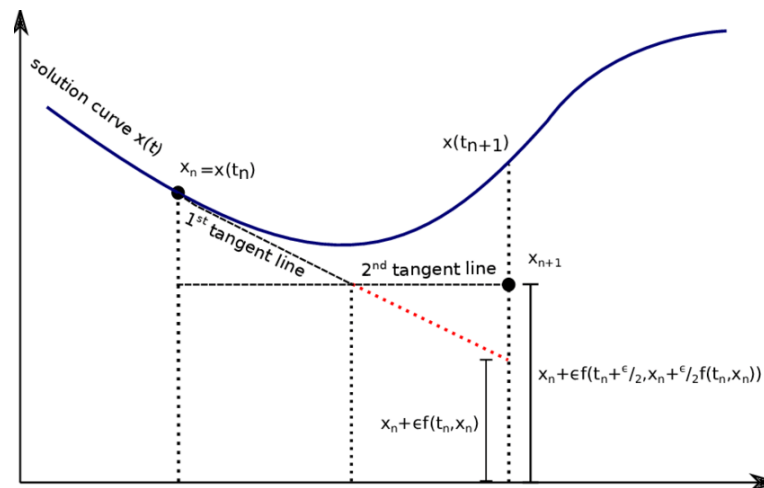
$$E_a = \begin{bmatrix} 10000 & 5000 \\ 25000 & 0 \\ 0 & 40000 \end{bmatrix} \quad \beta = 1 \quad \gamma = \begin{bmatrix} -1 & -1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$T(t) = \begin{cases} 333, & t < 0.5 \\ 325, & t \geq 0.5 \end{cases} \quad \mathbf{c}_0 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$



# Outline

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- **Other Numerical Methods**

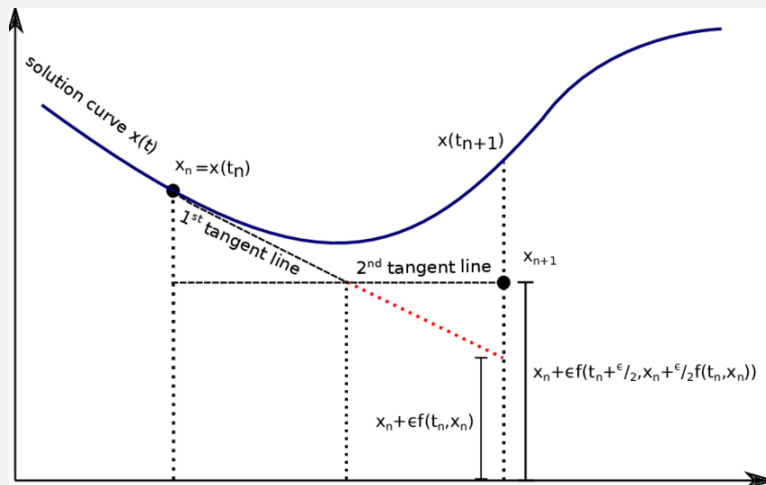




# More Advanced Methods

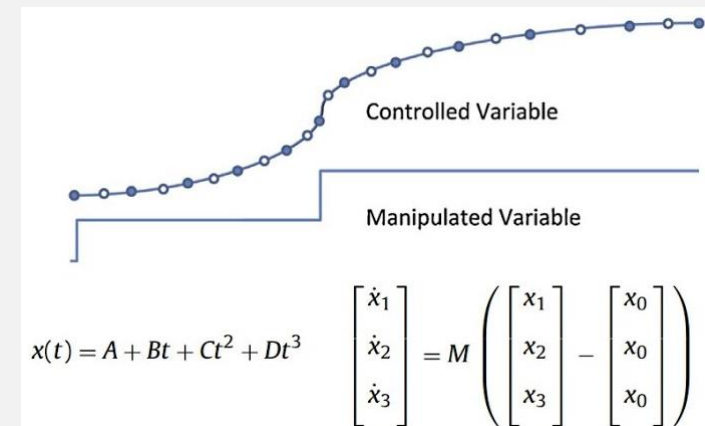
## Runge-Kutta

- Family of explicit and implicit iterative methods
- Various orders based on GTE  $O(h^p)$ 
  - 1<sup>st</sup> order methods are the Euler methods
  - 4<sup>th</sup> order methods are popular



## Orthogonal Collocation over Finite Elements

- The discretization uses finite elements
- We approximate the solution in each element as a polynomial function
- End up solving a system of linear equations

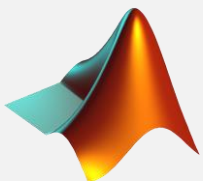




# Common Simulation Tools

## ODE Integrators

- Common in scripting languages
- Provide numerical solutions to ODE systems



## Symbolic Solvers

- Can provide analytic solutions when possible
- Typically, not used for large problems



## Optimization Tools

- Can incorporate differential equations when solving optimization problems

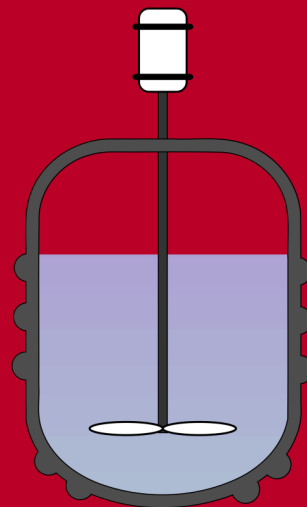


## Problem Specific

- Simulate dynamics for particular systems



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