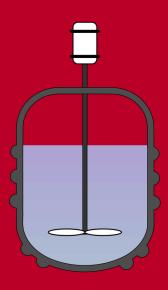
Carnegie Mellon University



Introductory Numerical Methods for Simulating Batch Reactors

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

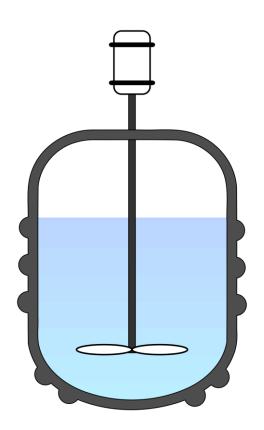
4. [Time Permitting] A familiarity of **other numerical methods/tools** for simulating batch reactors



- Overview
- Explicit Euler
- Batch Reactors
- Other Numerical Methods (If time allows)



- 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



Analytical vs. Numerical Methods

Analytical Methods

Separate and integrate

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

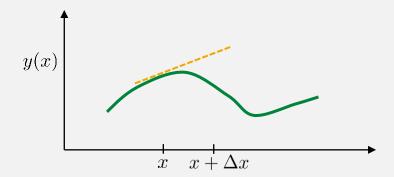
ODEs of special forms

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

Solving general ODEs is often difficult or not possible

Numerical Methods

- Seek to numerically approximate the solution
- Finite difference methods are common

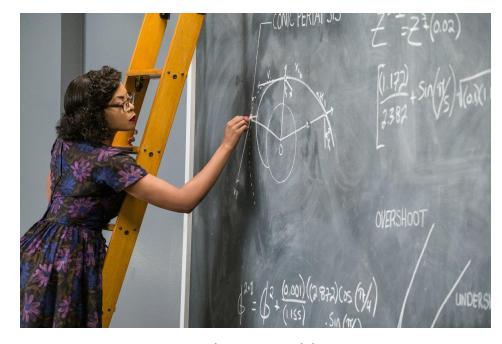


- More advanced methods are available
 - Not the focus of today



Outline

- Overview
- Explicit Euler
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- Other Numerical Methods





Katherine Goble



The Basics

Methodology

Consider a 1st order ODE

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

• Define time steps Δt

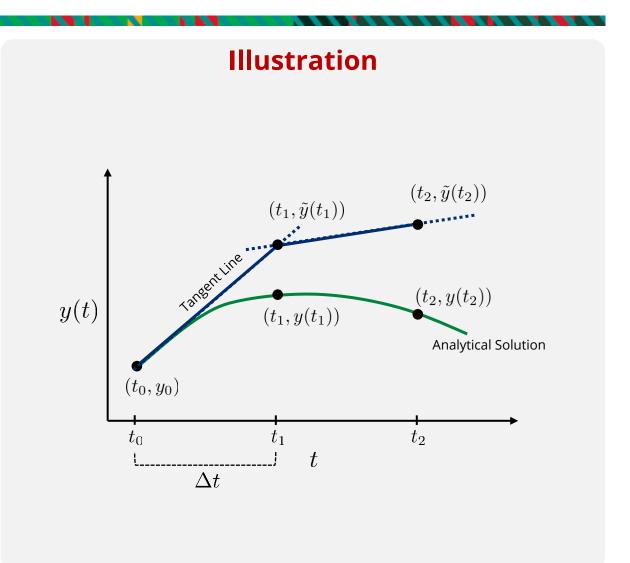
$$t \in [t_0, t_f] \qquad t_k = t_0 + k\Delta t$$

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

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





Simulating a System of ODEs

System of 1st order ODEs

General representation

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

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

$$\vdots$$

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

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

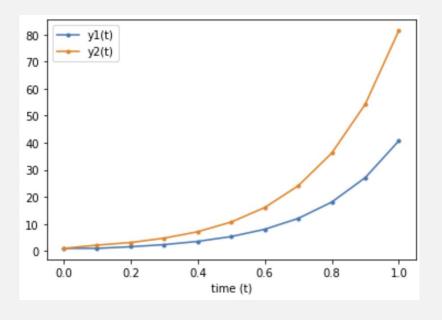
Vectorize

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

Vectorized Explicit Euler

Update rule uses vectorized representation

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



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

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

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

$$O(\Delta t)$$

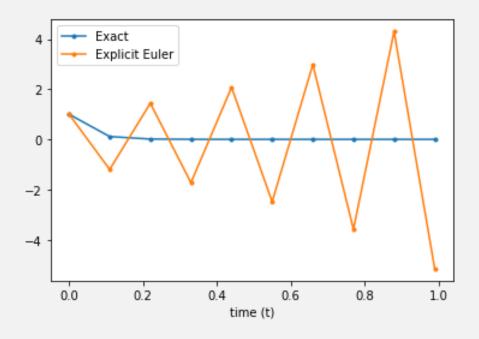
Hence, explicit Euler is a first order method

Higher order methods are available



Properties: Stability

Example



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

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

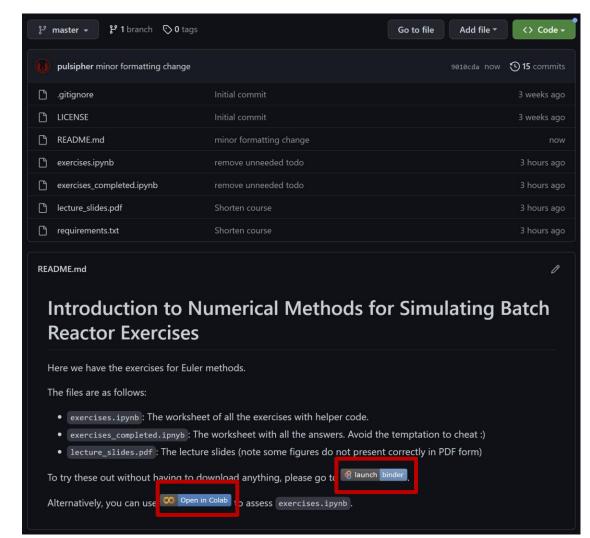


Tutorials

- 1. Go to https://github.com/pulsipher/eulercourse
- 2. Click on the "launch binder" button or click on the "Open in Colab" button
- 3. Open "exercise.ipynb" if you used binder



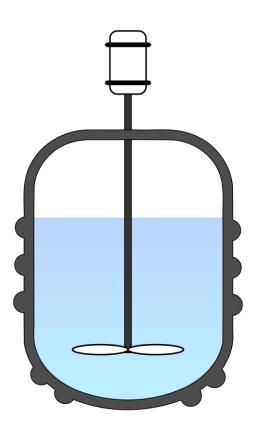






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Batch Reactor Modeling

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

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

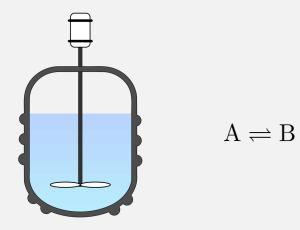
Species generation rates

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

Species balances

$$\frac{dc_i(t)}{dt} = g_i(\mathbf{c}, t), \quad i \in I$$
$$c_i(0) = c_{i,0}, \qquad i \in I$$

The Basics



Derivation from mole balance

$$\frac{N_i(t)}{dt} = F_i(t) - F_i(t) + \int_{-\infty}^{V} g_i(t)dV(t)$$

Assume perfect mixing and constant volume

$$\frac{N_i(t)}{dt} = g_i(t)V \qquad \qquad \frac{c_i(t)}{dt} = g_i(t)$$



Exercise: Batch Reactor

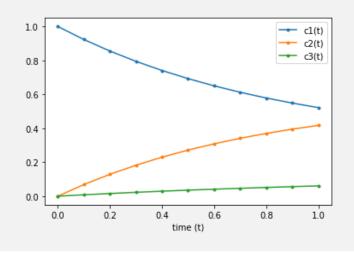
Exercise 1

• Simulate a batch reactor with the following reaction system using explicit Euler for $t \in [0, 1]$

$$A \rightleftharpoons B$$

$$A \rightleftharpoons C$$

• Experiment with different choices of Δt





Problem Information

$$R = 1.987$$

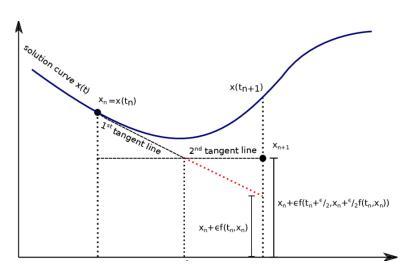
$$A = \begin{bmatrix} 3.6362e6 & 190.6879 \\ -2.5212e16 & 0 \\ 0 & -8.7409e24 \end{bmatrix}$$

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

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



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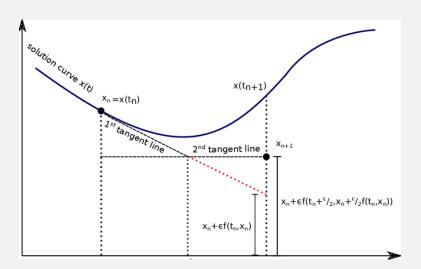




More Advanced Methods

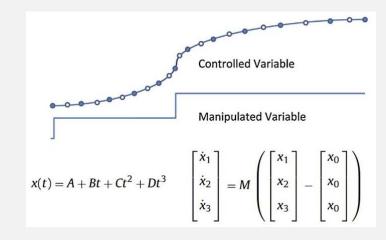
Runge-Kutta

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

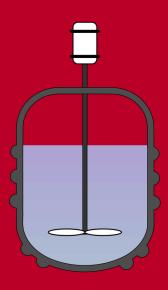
Simulate dynamics for particular systems







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