

Foundations of Chemical Kinetics Lecture 16:
The kinetic Monte Carlo method
The multi-molecule simulation method

Marc R. Roussel

November 17, 2021

Simulating many molecules

- What we have so far is a method for simulating a single molecule.
We can repeat single-molecule simulations many times to collect statistics.
- Often, we are less interested in detailed single-molecule trajectories than in the evolution of a probability distribution.
- KMC can handle this case, too. The basic idea is that if $w_{ij}dt$ is the probability that a single molecule executes a jump from state i to state j in time dt , then if there are N_i molecules in state i , the probability that **one** of them jumps from state i to state j in time dt is $N_i w_{ij}dt$ **provided** dt is sufficiently small that there is a negligible probability that there is more than one jump in that time.
- Instead of tracking the states of individual molecules, we will keep track of how many molecules are in each state.

System, states and propensities

- The **system** consists of a number of molecules distributed over the accessible molecular states.
- The **state of the system** is a vector of populations of molecules in each **molecular state**: $\mathbf{N} = (N_1, N_2, \dots, N_s, \dots)$.
- The **propensity** of a jump from molecular state i to molecular state j is $a_{ij} = N_i w_{ij}$.
This is a probability per unit time of **one** molecule making a jump from molecular state i to state j .
- The KMC algorithm for many molecules is much the same, except that
 - Propensities are used to determine jump probabilities.
 - Instead of updating the state of one molecule, we update the populations of the “from” and “to” states.

The Kinetic Monte Carlo (KMC) Algorithm

Simulating a population of molecules

Initialize:

- 1 Store the w_{ij} values.
- 2 Store the initial N_i populations.
- 3 Set $t = 0$.

The Kinetic Monte Carlo (KMC) Algorithm

Simulating a population of molecules (continued)

At each step:

- 1 Calculate the propensities $a_{ij} = w_{ij} N_i$.
- 2 Calculate $a_{\text{tot}} = \sum_{i,j} a_{ij}$.
- 3 Generate two random numbers $r_{1,2} \in (0, 1)$.
- 4 Use r_1 to generate the time to next jump:

$$\Delta t = -\ln r_1 / a_{\text{tot}}$$

The Kinetic Monte Carlo (KMC) Algorithm

Simulating a population of molecules (continued)

- 5 Use r_2 to pick the destination of the jump as follows:
 - 1 Arrange the a_{ij} so that they are stored in a vector $\hat{\mathbf{a}}$.
The best way to do this will depend on the structure of the problem.
 - 2 Find the smallest k such that

$$\sum_{\ell=1}^k \hat{a}_{\ell} > r_2 a_{\text{tot}}$$

then work backwards to determine the (i, j) values corresponding to k .

How we find the (i, j) values corresponding to k will depend on how we 'linearized' the a_{ij} values.

The Kinetic Monte Carlo (KMC) Algorithm

Simulating a population of molecules (continued)

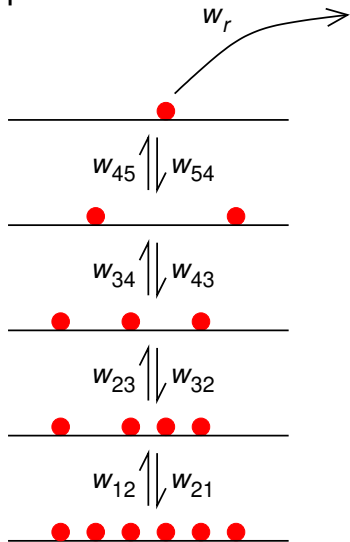
- 6 Add Δt to the simulation time.
- 7 Adjust populations according to the jump:

$$N_i = N_i - 1$$

$$N_j = N_j + 1$$

- 8 Recalculate (at least) a_{ik} and $a_{jk} \forall k$.

A simple model of a unimolecular reaction



- Assume equally spaced levels

A simple model of a unimolecular reaction (continued)

- Strong collision assumption implies

$$w_{\text{down}} \equiv w_{21} = w_{32} = w_{43} = w_{54} = Z_A = \sigma \bar{u}_{\text{ref}} L[\text{B}] \quad (\text{B} = \text{bath gas})$$

- Assume $w_{\text{down}} = 10^9 \text{ s}^{-1}$.

- From the equilibrium conditions,

$$w_{\text{up}} \equiv w_{12} = w_{23} = w_{34} = w_{45} = Z_A e^{-\Delta\epsilon/k_B T}$$

- Assume $w_{\text{up}} = 0.8 w_{\text{down}}$.

- w_r is related to the rate of IVR and crossing the transition-state dividing surface.

- Assume $w_r = 10^{13} \text{ s}^{-1}$.

- rate constant = $\frac{\text{rate of reaction}}{\text{number of unreacted molecules}} = -\frac{1}{N} \frac{dN}{dt} =$

$$-\frac{d \ln N}{dt}$$

$$\text{where } N = \sum N_i.$$

A simple model of a unimolecular reaction (continued)

- Initial condition: $N_1 = N$, $N_2 = N_3 = N_4 = N_5 = 0$
- Organization of the $\hat{\mathbf{a}}$ vector:
 $[a_{12} \ a_{23} \ a_{34} \ a_{45} \ a_{21} \ a_{32} \ a_{43} \ a_{54} \ a_r]$

Some useful Matlab knowledge

- If you run a program repeatedly, there is a chance that data stored in previous runs will interfere with your current calculation. The `clear` command clears all variables.
- We can operate on parts of a vector by referring to ranges of indices, e.g. if `a` is the propensity vector of length 9, `N` is the population vector of length 5, and `wup` is a scalar,

$$a(1:4) = wup*N(1:4);$$

will calculate the propensities of the upward jumps and store them in the first 4 elements of the propensity vector.

- Because vector operations are very efficient in Matlab, it may be advantageous to do a few vector operations rather than write something more complicated to try to save arithmetic operations.
- The keyword `end` can be used as an index into a vector to denote the last element of the vector.

Some useful Matlab knowledge

Conditional execution

- The ability to only run code if certain conditions are met is central to a lot of programming.
while loops provide a form of conditional execution.
- `if...elseif...else` structures are often useful.

General syntax:

```
if condition  
    code to execute  
elseif another condition  
    alternative code  
else  
    Default code  
end
```

Some useful Matlab knowledge

Conditional execution (continued)

- `elseif` is optional, but you can have many `elseif` statements if needed.
- `else` is optional.
If present, it is executed if none of the previous conditions are true.
- The conditions are tested in order. Once a true condition is found, the code in that block is executed, and then the program skips to end of the `if` control structure.
No more than one statement in an `if` control structure will ever be executed.

Two powerful Matlab functions

`cumsum()` calculates cumulative sums of a vector.

Example: `cumsum(a)` returns the vector
 $(a_1, a_1 + a_2, \dots, \sum_{\ell} a_{\ell})$.

Note that a_{tot} is just the last element returned by `cumsum()`.

`find()` has many useful options for locating quantities matching a particular condition.

The KMC jump destination selection rule, finding the smallest k for which

$$\sum_{\ell=1}^k \hat{a}_{\ell} > r_2 a_{\text{tot}}$$

can be coded as follows in Matlab:

```
a_sums = cumsum(a);
```

```
atot = cumsum(end);
```

```
k = find(a_sums > r2*atot,1,'first');
```

Fitting and plotting in Matlab

`coeffs = polyfit(x,y,degree)` fits a polynomial of the selected degree to (x, y) data.

`coeffs` is a vector of coefficients stored in descending order of exponent.

`hold` tells Matlab to keep adding to a plot rather than erasing and starting over.

`hold off` tells Matlab that the plot is complete.