

Rare Events & Nudged Elastic Band method

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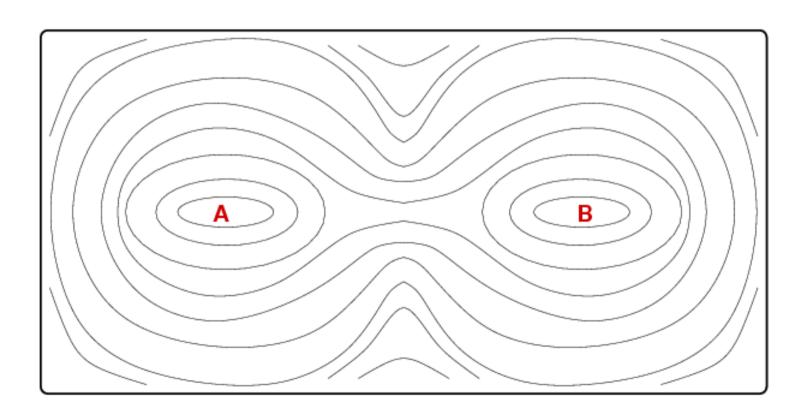
Bangalore, July 2006

Potential Energy Landscape



Let's consider a potential energy function (as defined in the lecture about dynamics) with two minima $\{q_A^{3N}\}$ and $\{q_B^{3N}\}$.

A two-dimensional example:

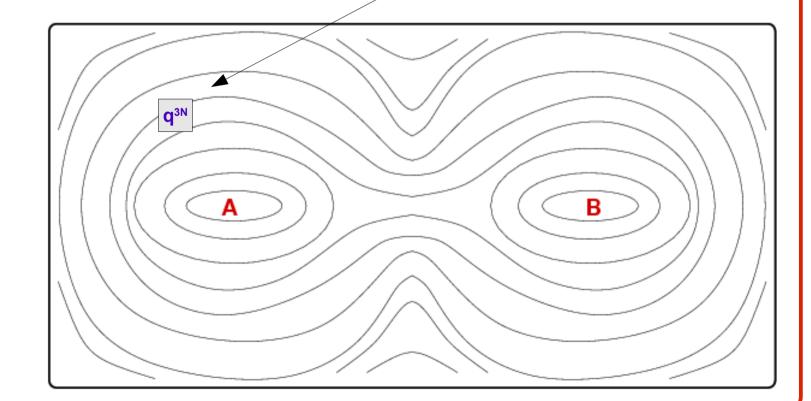


Potential Energy Landscape



The probability of finding the system in the neighbourhood of any configuration $\{q^{3N}\}$ is proportional to the Boltzmann factor computed at such configuration:

$$\mathcal{P}\left(q^{3N}\right) \sim \frac{\mathrm{e}^{-\beta U\left(q^{3N}\right)}}{\mathcal{Z}}$$

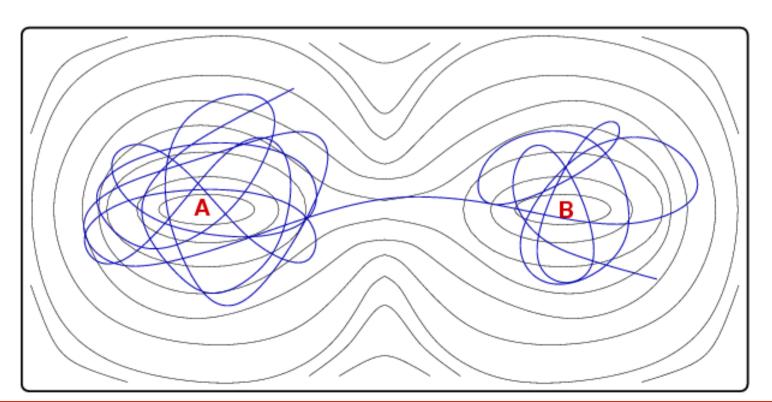


DEMOCRITOS DEmocritos Modeling Center for INFIM

Dynamics in the Potential Energy Landscape

Equivalently (ergodicity) the relative time spent around a configuration $\{q^{3N}\}$ is proportional to the Boltzmann factor.

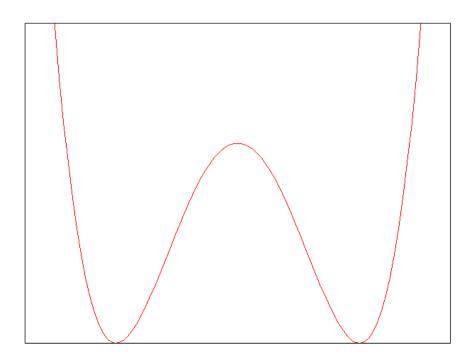
The system spends more time around configurations of low potential energy. Fluctuations are responsible for the transition between different minima.





1-dimensional brownian motion in a double-well potential:

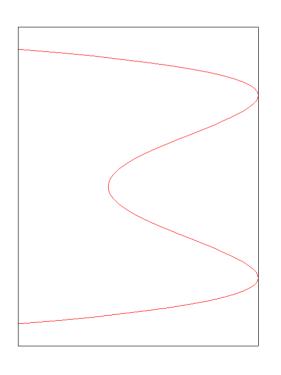
$$dx(t) = -\nabla V(x(t)) dt + \xi(t)$$

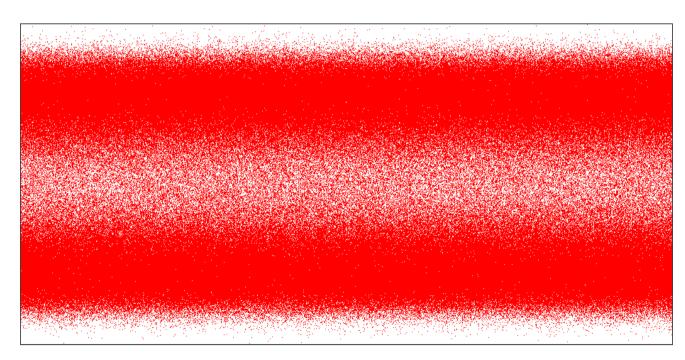




Brownian motion in a double-well potential at a temperature $K_{\rm B}T\sim 0.5~E_{_{A}}$ (50% of the barrier height):

diffusive behaviour

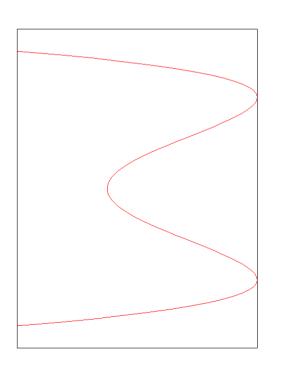


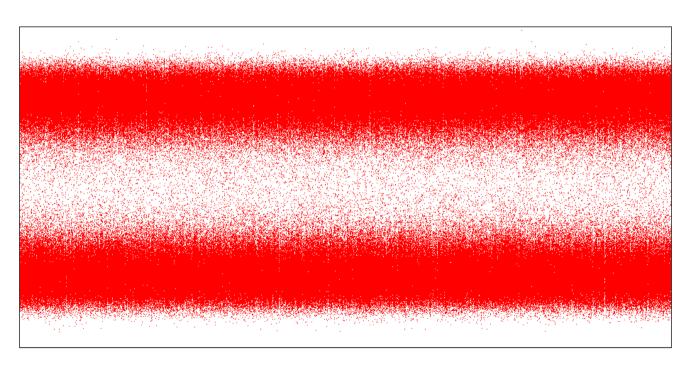


time



Brownian motion in a double-well potential at a temperature $K_BT \sim 0.2 E_A$ (20% of the barrier height):



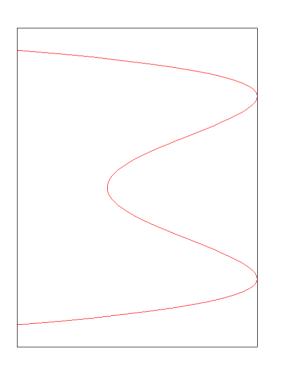


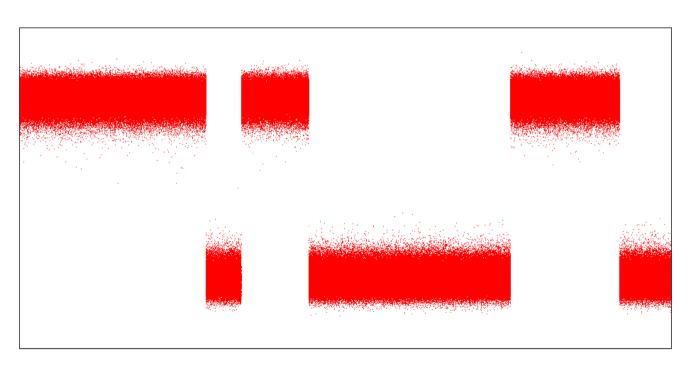
time



Brownian motion in a double-well potential at a temperature $K_{\rm B}T \sim 0.08~E_{\rm A}$ (8% of the barrier height):

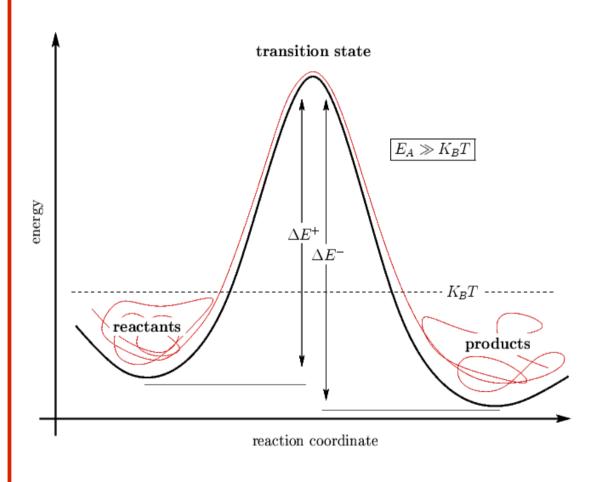
"instantonic" behaviour

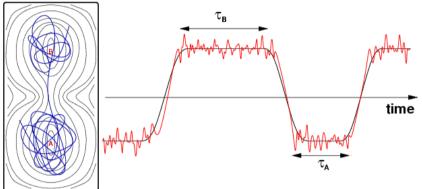




time







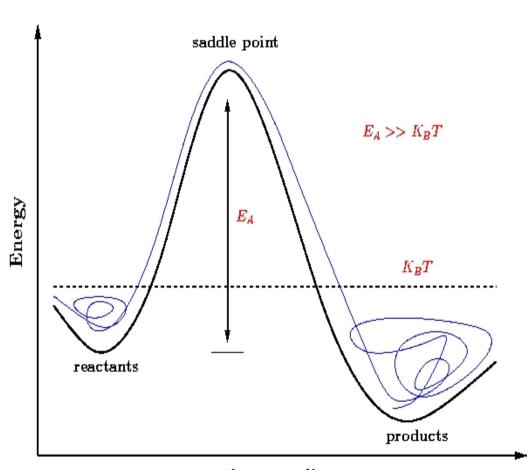
$$\tau_A \sim e^{\beta \Delta E^+}$$

$$\tau_B \sim e^{\beta \Delta E^-}$$

$$\frac{\tau_A}{\tau_B} = e^{\beta(E_B - E_A)}$$

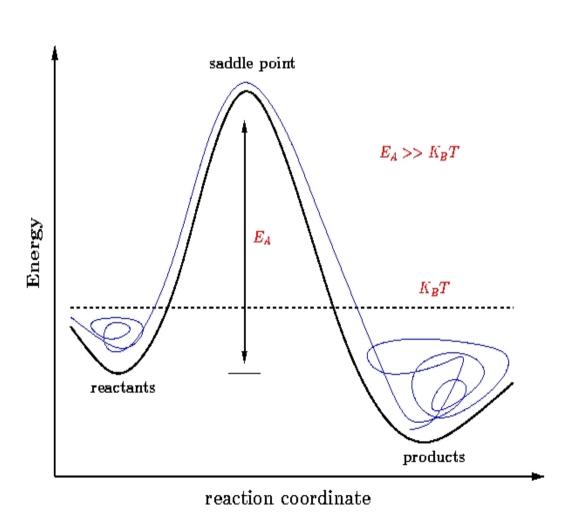


What is the characteristic time scale of this transition process?



reaction coordinate



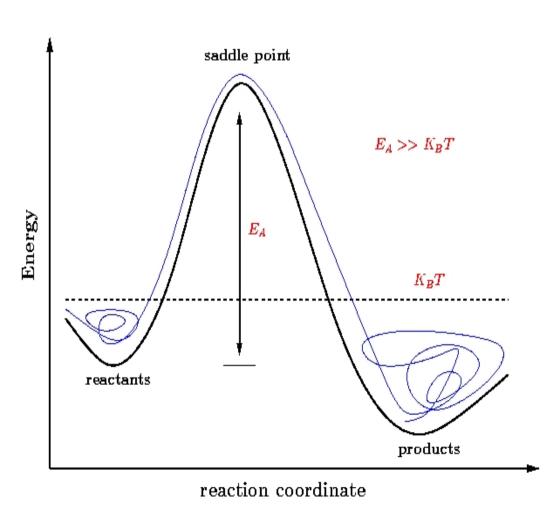


What is the characteristic time scale of this transition process?

Van't-Hoff - Arrhenius (1890)

$$t_{\text{jump}} \sim t_{\text{vib}} \cdot e^{\frac{E_A}{K_B T}}$$



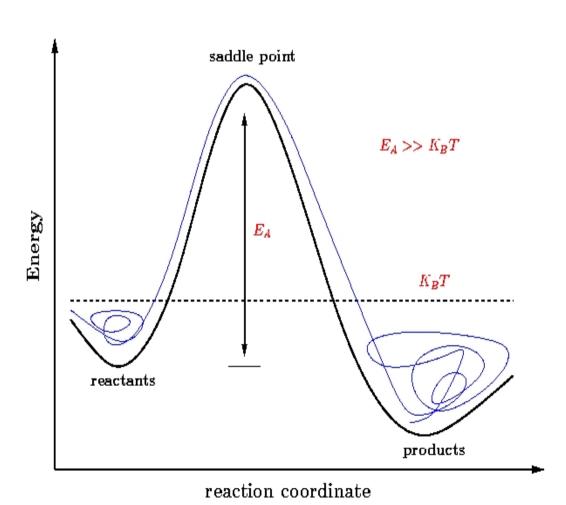


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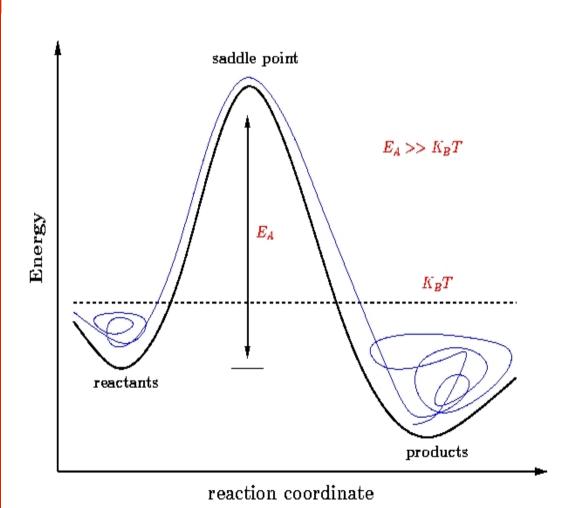
$$t_{\text{jump}} \sim t_{\text{vib}} \cdot \mathrm{e}^{\frac{E_A}{K_B T}}$$
 $t_{\text{vib}} \sim 10^{-13} \, \mathrm{s}$ $T = 300 \, \mathrm{K}$





Assuming a time step of one femtosecond, $\sim 10^{15}$ steps of Molecular Dynamics are necessary to have a sizeable probability of observing at least a transition from reactants to products!

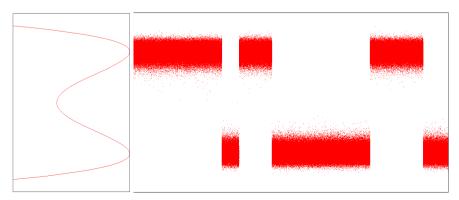




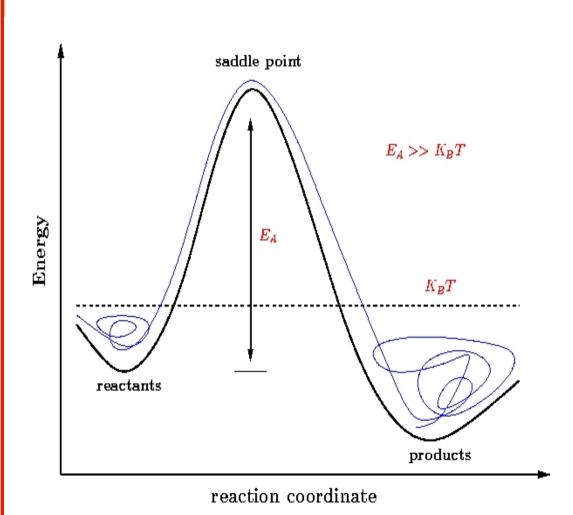
Assuming a time step of one femto-second, $\sim 10^{15}$ steps of Molecular Dynamics are necessary to have a sizeable probability of observing at least a transition from reactants to products!

Nevertheless, when an appropriate fluctuation occurs, the process is extremely fast:

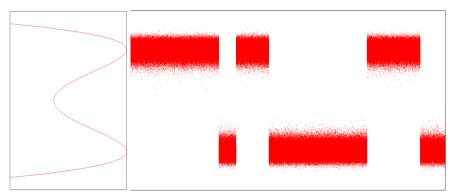
 $t \sim femto-seconds (10^{-15} s)$





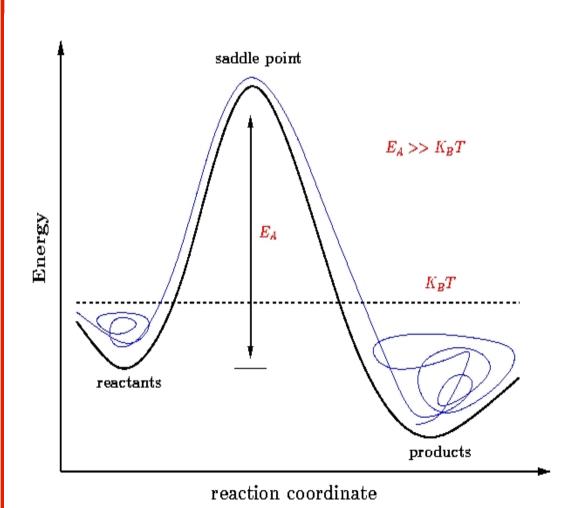


What is macroscopically perceived as a slow process is instead a rare event.



The jumps are uncorrelated





Alternative approach:

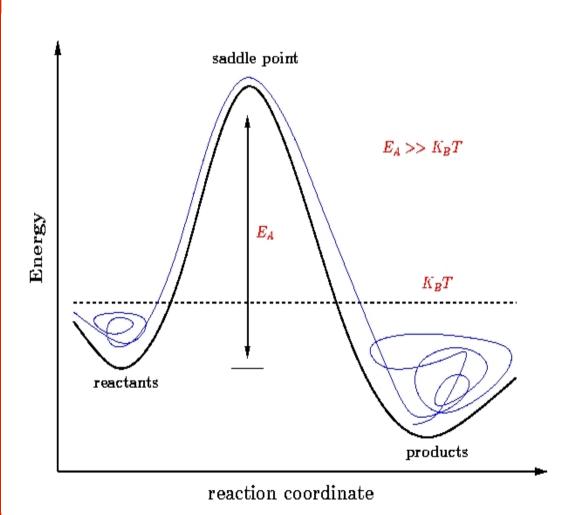
The transition rate can be estimated using equilibrium statistical mechanics:

once the saddle point has been identified we can use <u>harmonic</u> <u>Transition State Theory</u> (hTST) to compute the rate constant:

$$k_{\text{reactants} \to \text{products}} = \mathcal{A} \cdot e^{-\frac{E_A}{K_B T}}$$

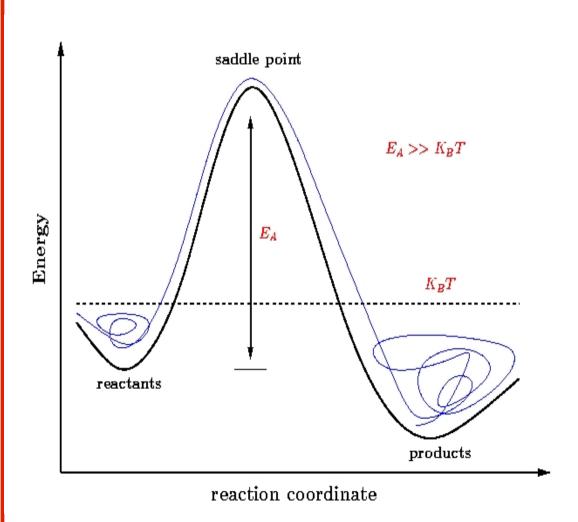
$$\mathcal{A} = \frac{\prod_{i=1}^{3N} \nu_i^{ ext{reactants}}}{\prod_{i=1}^{3N-1} \nu_i^{ ext{saddle point}}}$$





The goal is locating all the relevant saddle points, but:





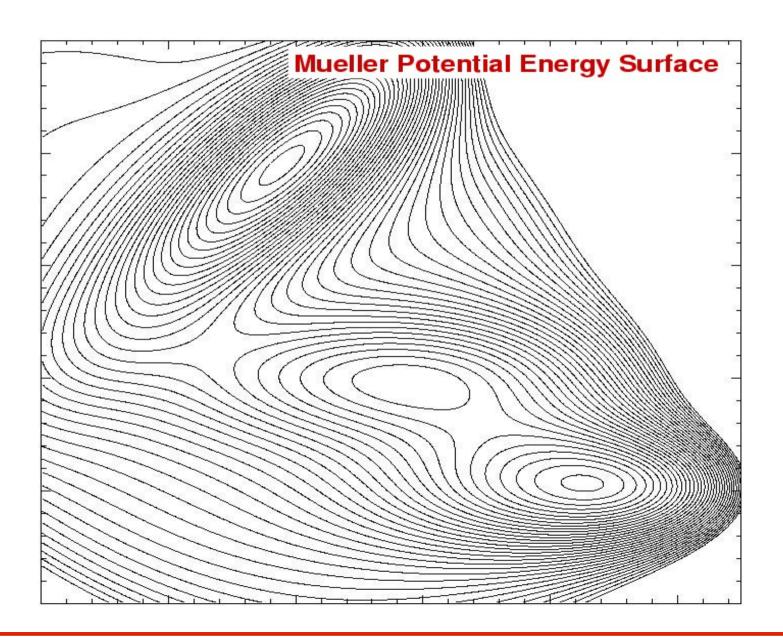
The goal is locating all the relevant saddle points, but:

saddle points are unstable

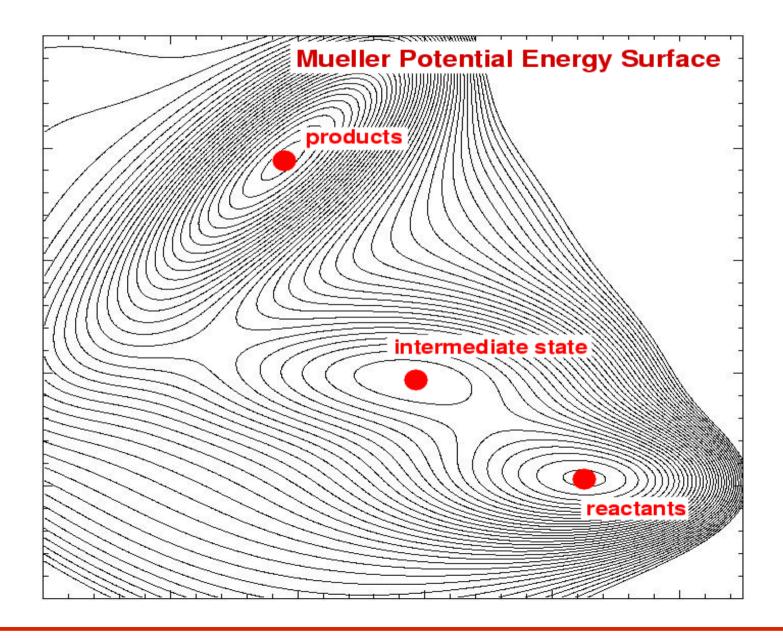


their direct location is a rather difficult task

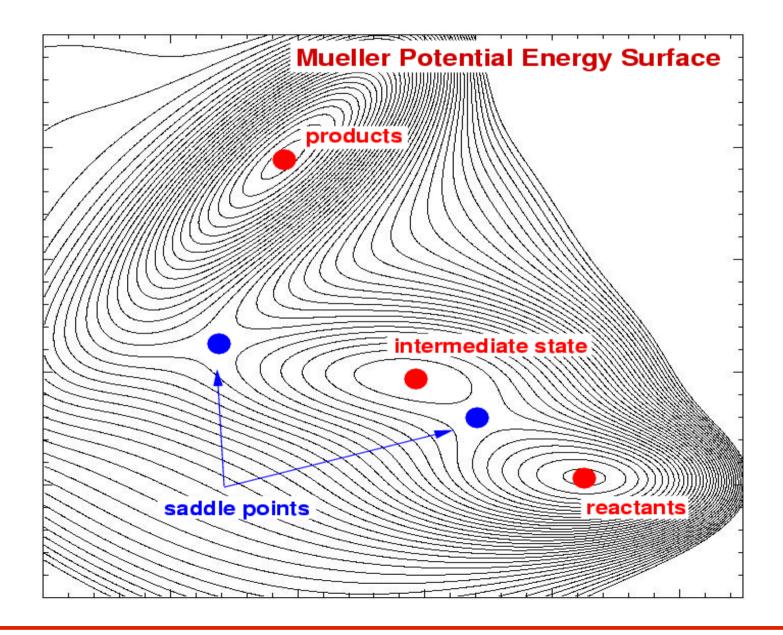




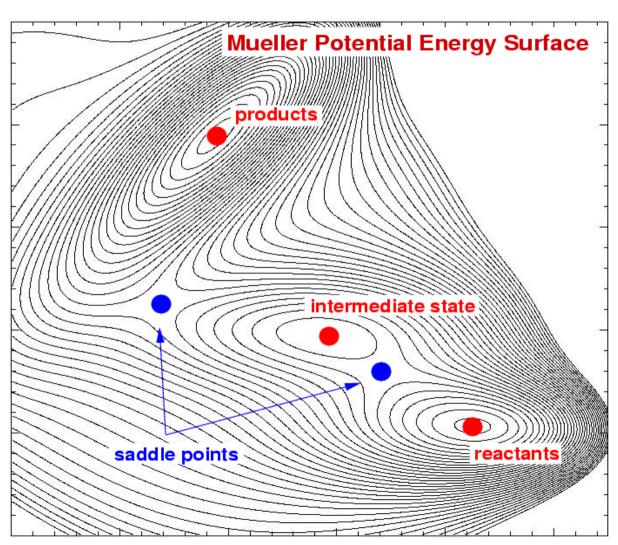










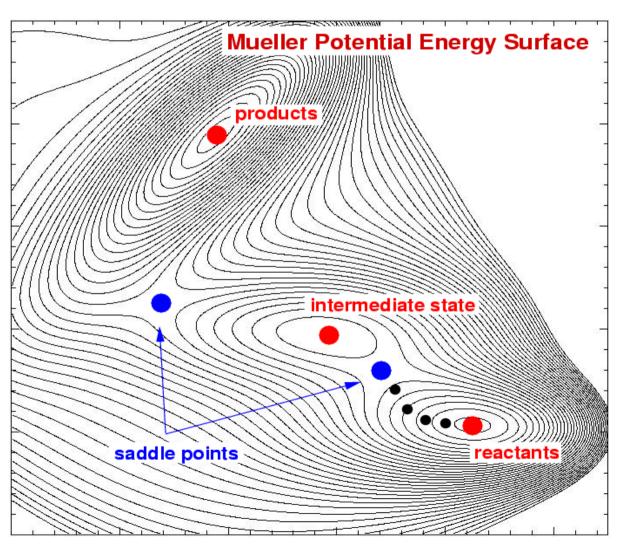


The path characterized by the "highest" transition probability, at zero temperature, is the Minimum Energy Path.

$$\nabla V(x(s)) - \tau(s) \; (\tau(s)|\nabla V(x(s))) = 0$$

$$\text{normalised tangent}$$

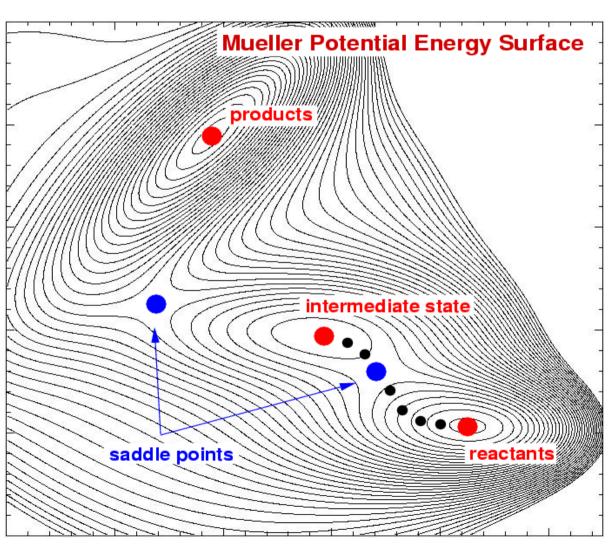




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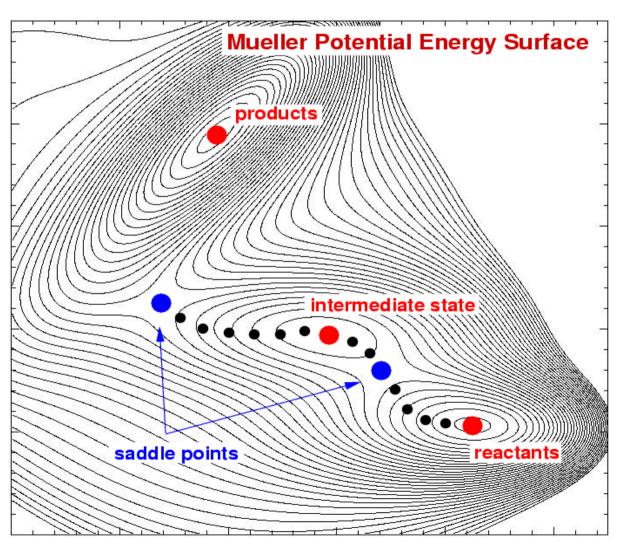


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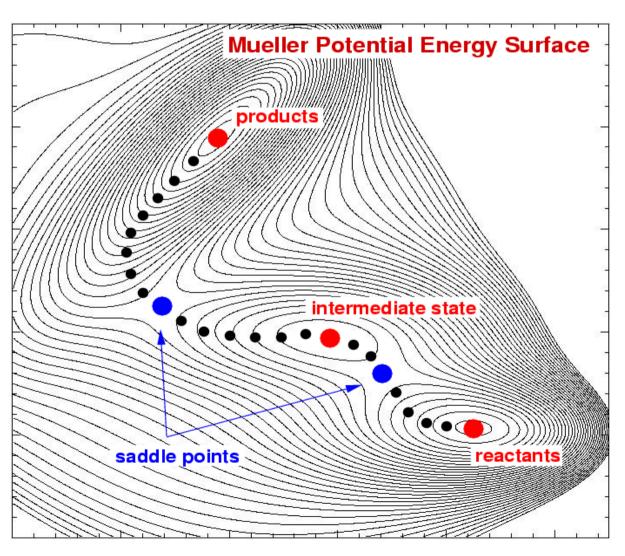




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$$\nabla V(x(s)) - \tau(s) \; (\tau(s)|\nabla V(x(s))) = 0$$
 normalised tangent





The path characterized by the "highest" transition probability, at zero temperature, is the Minimum Energy Path.

MEP: the components of the force orthogonal to the path are zero.

The MEP crosses the saddle points.



1) Path discretisation ("chain of images"):

$$s \longrightarrow i \cdot \delta s$$

$$x(s) \longrightarrow x_s$$

$$au(s) \longrightarrow au_i = rac{x_{i+1} - x_{i-1}}{|x_{i+1} - x_{i-1}|}$$



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2) Orthogonal forces:

$$F(x_i)_{\perp} = -\left[\nabla V(x_i) - \tau_i \left(\tau_i | \nabla V(x_i)\right)\right]$$



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4) path dynamics
 (steepest-descent) :

$$x_i^{k+1} = x_i^k + \lambda F(x_i^k)_{\perp}$$



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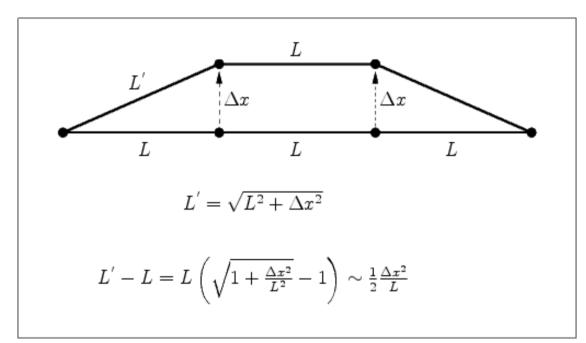
$$x_i^{k+1} = x_i^k + \lambda F(x_i^k)_{\perp}$$

5) Alternatively, Broyden acceleration:

$$x_i^{k+1} = x_i^k + J^{-1} F(x_i^k)_{\perp}$$



The path dynamics does not preserve the inter-image distance (path's parametrisation):

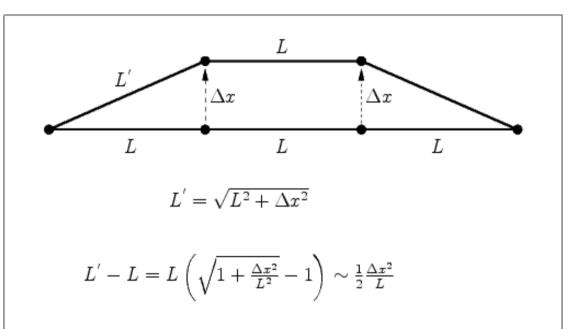




The path dynamics does not preserve the inter-image distance (path's parametrisation):

Consequences:

1) Many images are required to represent the path.

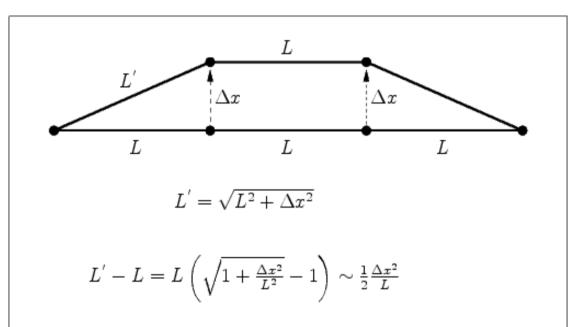




The path dynamics does not preserve the inter-image distance (path's parametrisation):

Consequences:

- 1) Many images are required to represent the path.
- 2) The images can eventually slide down to the two minima.

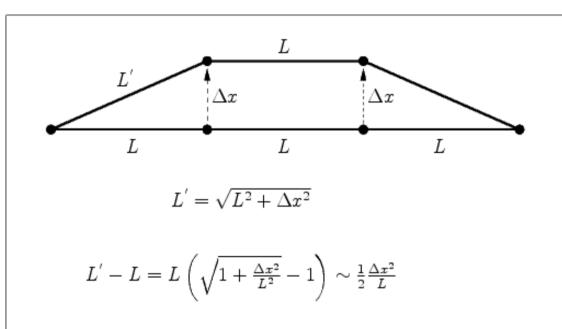




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Possible solutions:

- 1) NEB: the images are connected by springs.
- 2) STRING: images are kept equispaced using Lagrange constraints.

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Nudged Elastic Band method



→ Subsequent images of the chain are connected by springs (to enforce continuity).

Nudged Elastic Band method



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- Each image feels forces due to <u>external potential + springs</u>.

Nudged Elastic Band method



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- Each image feels forces due to <u>external potential + springs</u>.
- → NEB idea [1,2]: elastic forces are projected along the path and external forces are projected orthogonally to the path.

[1] G.Mills and H.Jonsson, Phys. Rev. Lett. 72, 1124 (1994)

[2] G.Henkelman and H.Jonsson, J. Chem. Phys. 133, 9978 (2000)

Nudged Elastic Band method



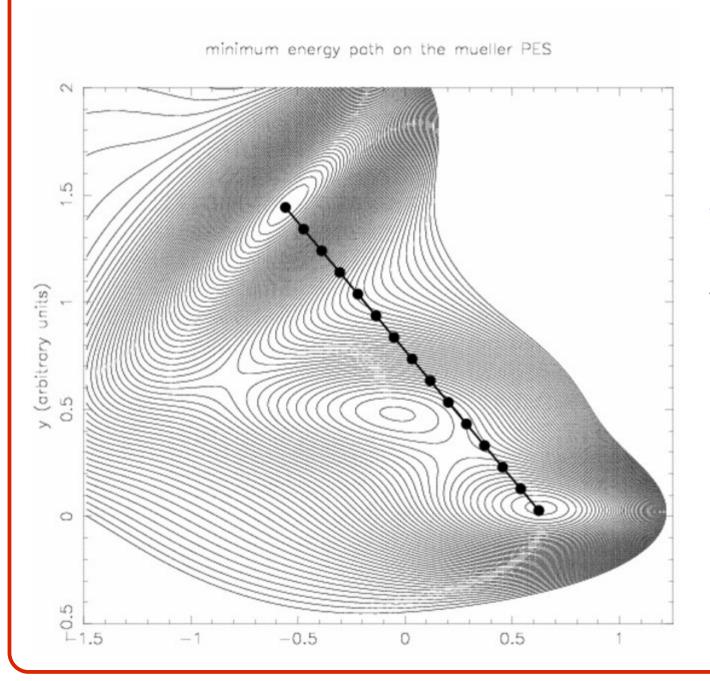
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- Each image feels forces due to <u>external potential + springs</u>.
- → NEB idea [1,2]: elastic forces are projected along the path and external forces are projected orthogonally to the path.
- → Projections are defined by the path's tangent: the tangent definition plays a crucial role.

[1] G.Mills and H.Jonsson, Phys. Rev. Lett. 72, 1124 (1994)

[2] G.Henkelman and H.Jonsson, J. Chem. Phys. 133, 9978 (2000)

NEB on the Mueller PES



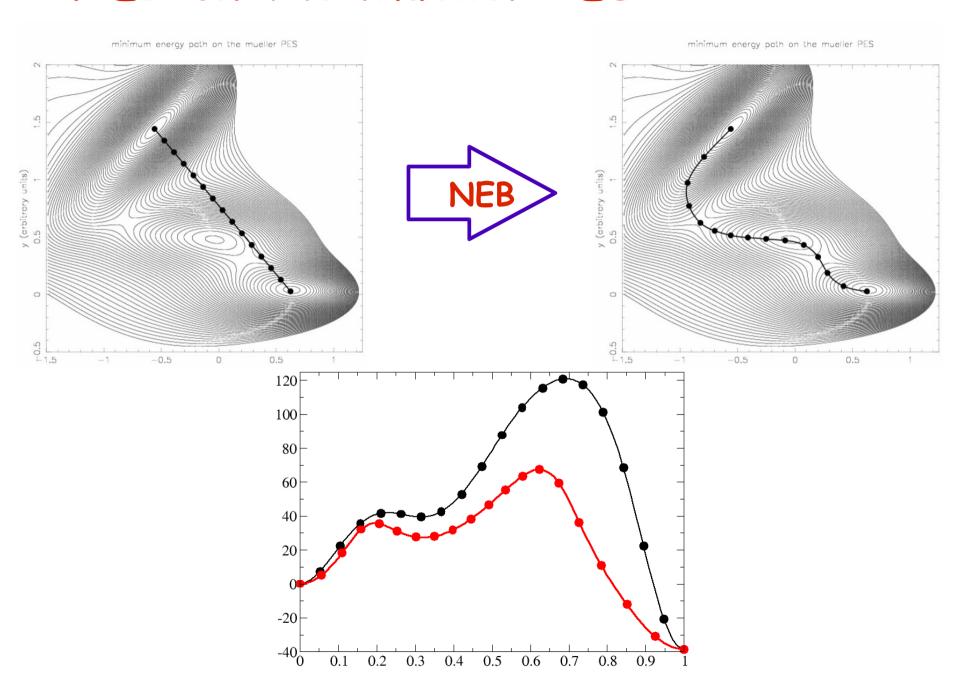


Initial guess: linear interpolation between the two end points.

$$\vec{x}_i = \vec{x}_0 + \frac{i}{N} \left(\vec{x}_N - \vec{x}_0 \right)$$

NEB on the Mueller PES





Improvements



1) Higher resolution around the saddle point: variable elastic constants.

2) Accurate identification of saddle point: climbing image.

Variable elastic constants



- \rightarrow The user specifies the minimum (k_min) and the maximum (k_max) values of the elastic constants.
- → Spring constants can be chosen so that springs are stiffer where the potential energy is higher ($k_{max} > k_{min}$): higher resolution around saddle points.
- The value of the elastic constant for each image x_i is obtained by interpolating between k_{\min} and k_{\max} :

$$k_i = \frac{1}{2} \left(k_{max} + k_{min} - (k_{max} - k_{min}) \cos \left(\pi \frac{(V(x_i) - V_{min})}{(V_{max} - V_{min})} \right) \right)$$

Climbing image



A given image (x_i) can be made "to climb" up-hill the PES once the climbing direction is specified.

In the CI scheme the direction is given by the path's tangent.

$$F(x_{i_{max}}) = -\nabla V(x_{i_{max}}) + 2\tau_{i_{max}} (\tau_{i_{max}}) \nabla V(x_{i_{max}})$$

- → The image can be automatically chosen during the optimisation as the one with the highest energy (CI_scheme="auto").
- → One or more images can be forced to climb (CI_scheme="manual").
- Climbing Image should be used after some optimisation steps.





A detailed explanation of all the keywords can be found in the file Doc/INPUT_PW.

```
&CONTROL
 calculation = "neb" <= mandatory</pre>
                              optional (0)
 nstep
&IONS
 num of images
                              mandatory
 CI scheme
                              optional (no-CI)
                              optional (quick-min)
 opt scheme
                          <=
                              optional (1.0)
 ds
                          <=
  first last opt
                              optional (.FALSE.)
                          <=
                              optional (0.1)
 k max
                          <=
                              optional (0.1)
 k min
                          <=
                              optional (0.05)
 path thr
                          <=
```

NEB: input variables



A detailed explanation of all the keywords can be found in the file Doc/INPUT_PW.

```
first image
                                                                 mandatory
   X \ 0.0 \ 0.0 \ 0.0 \ \{ \text{ if } pos(1) \text{ if } pos(2) \text{ if } pos(3) \}
                   { if pos(1) if pos(2) if pos(3) }
   Y 0.5 0.0
               0.0
                     { if pos(1) if pos(2) if pos(3) }
   Z 0.0 0.2
               0.2
intermediate image 1
                                                             <= optional</pre>
   X 0.0 0.0
               0.0
   Y 0.9 0.0 0.0
   z_{0,0} 0.2
               0.2
intermediate image ...
                                                             <= optional
   x 0.0 0.0 0.0
          0.0 0.0
   Y 0.9
          0.2 0.2
   Z 0.0
last image
                                                                 mandatory
   x 0.0 0.0
               0.0
   Y 0.7 0.0 0.0
   Z 0.0 0.5 0.2
```

NEB: output



Files in the working directory (./):

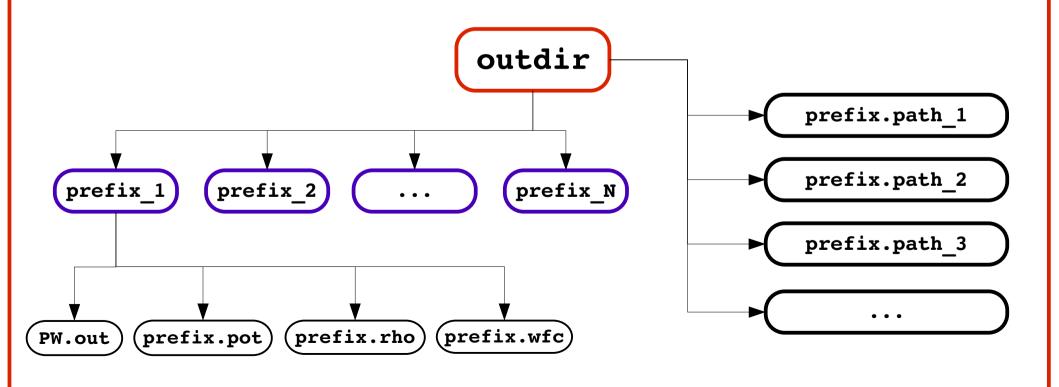
- prefix.path <= file containing data required to restart a NEB calculation</pre>
- → prefix.axsf <= file containing the path in xcrysden format
- → prefix.xyz <= file containing the path in xyz format
- → prefix.dat <= file containing the reaction coordinate, the energy and the error of each image
- → prefix.int <= file containing a cubic interpolation for the energy profile

NEB: output



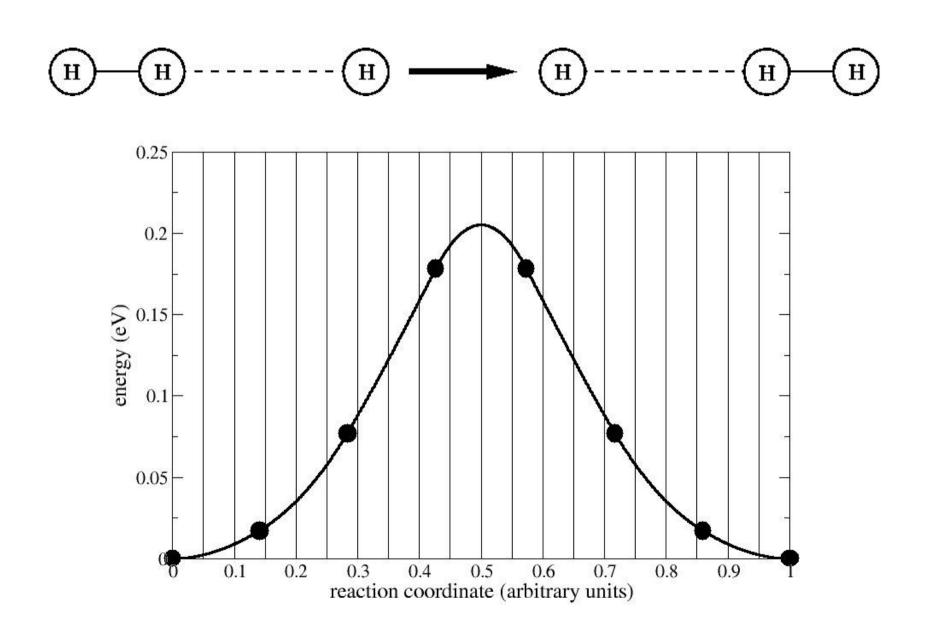
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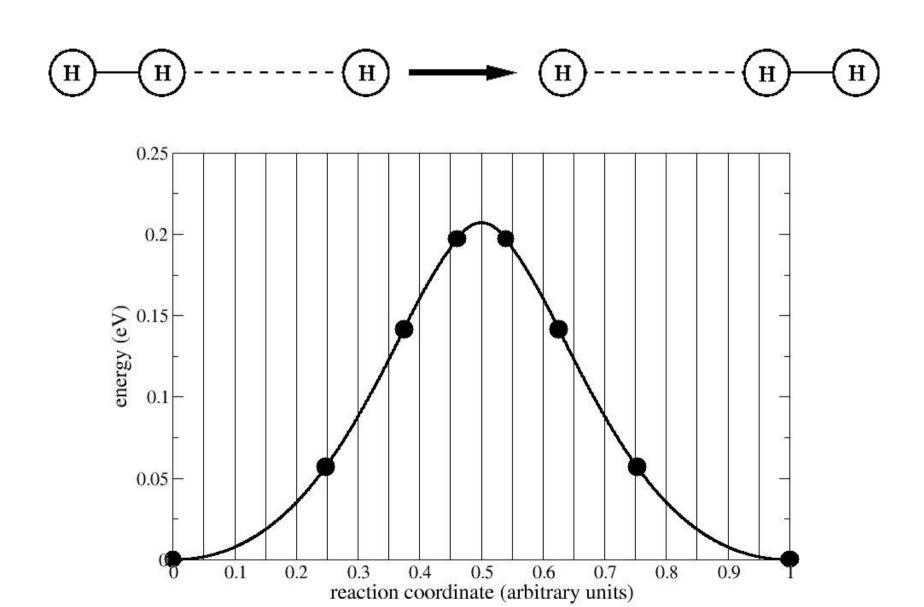
Example 17: collinear proton transfer plain NEB





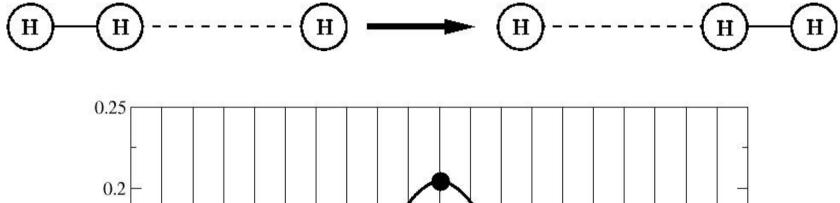
Example 17: collinear proton transfer variable elastic constants

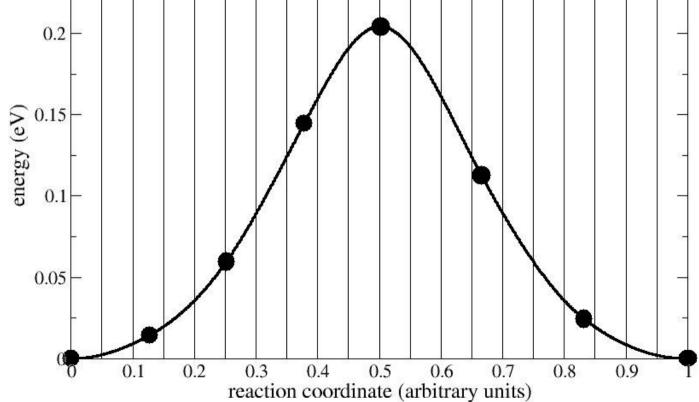




Example 17: collinear proton transfer climbing image (manual on image 5)







Notes on parallel execution



PWscf has two levels of parallelisation plus one specific for NEB:

1) R&G: wave-functions are distributed among the CPUs so that each CPU works on a subset of plane-waves. The same is done on real-space grid-points. By <u>default</u> all the CPUs are used for this <u>parallelisation</u> <u>scheme</u>. Example with 8 CPUs:

```
prompt> mpirun -np 8 pw.x -in input_file > output_file
```

2) k-points: k-points (when present) are distributed among pools of CPUs. Each pool can contain one or more CPUs. In this latter case R&G parallelisation is used within the pool. This scheme is selected by specifying in the command line the required number of k-points pools. Example with 8 CPUs and 2 k-points pools:

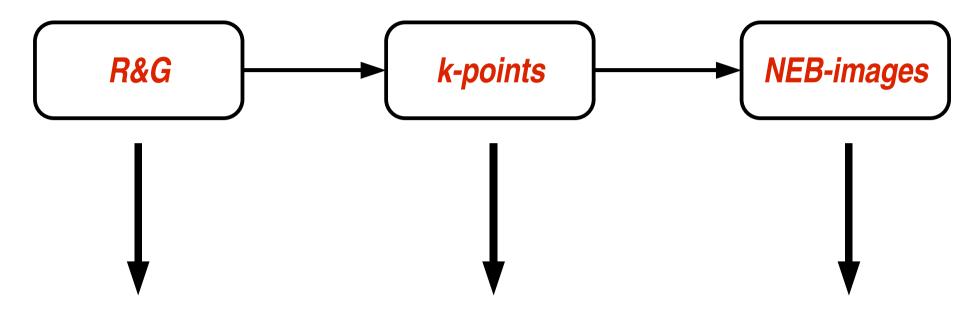
```
prompt> mpirun -np 8 pw.x -npool 2 -in input_file > output_file
```

3) NEB-images: NEB images are distributed among pools of CPUs. Within each image R&G and k-points parallelisation schemes can also be used. This scheme is selected by specifying in the command line the required number of images pools. Example with 8 CPUs, 2 k-points pools and 2 images pools:

```
prompt> mpirun -np 8 pw.x -npool 2 -nimage 2 -in input_file > output_file
```

Notes on parallel execution





- High scalability of the memory usage.
- × High intra-pool communication.
- Good work-load balance among CPUs.
- Best choice. Bad scaling when number of plane-waves per CPU is very small.
- × NO scalability of the memory usage.
- **X** Low inter-pool communication.
- Work-load among pools can result to be unbalanced.
- X Good when npool is a whole divisor of the number of kpoints.

- **x** NO scalability of the memory usage.
- **x** Extremely low inter-pool communication.
- Work-load among pools can result to be unbalanced.
- X Good for paths with several images.