



**Cinvestav**

# Random methods to predict novel crystalline structures

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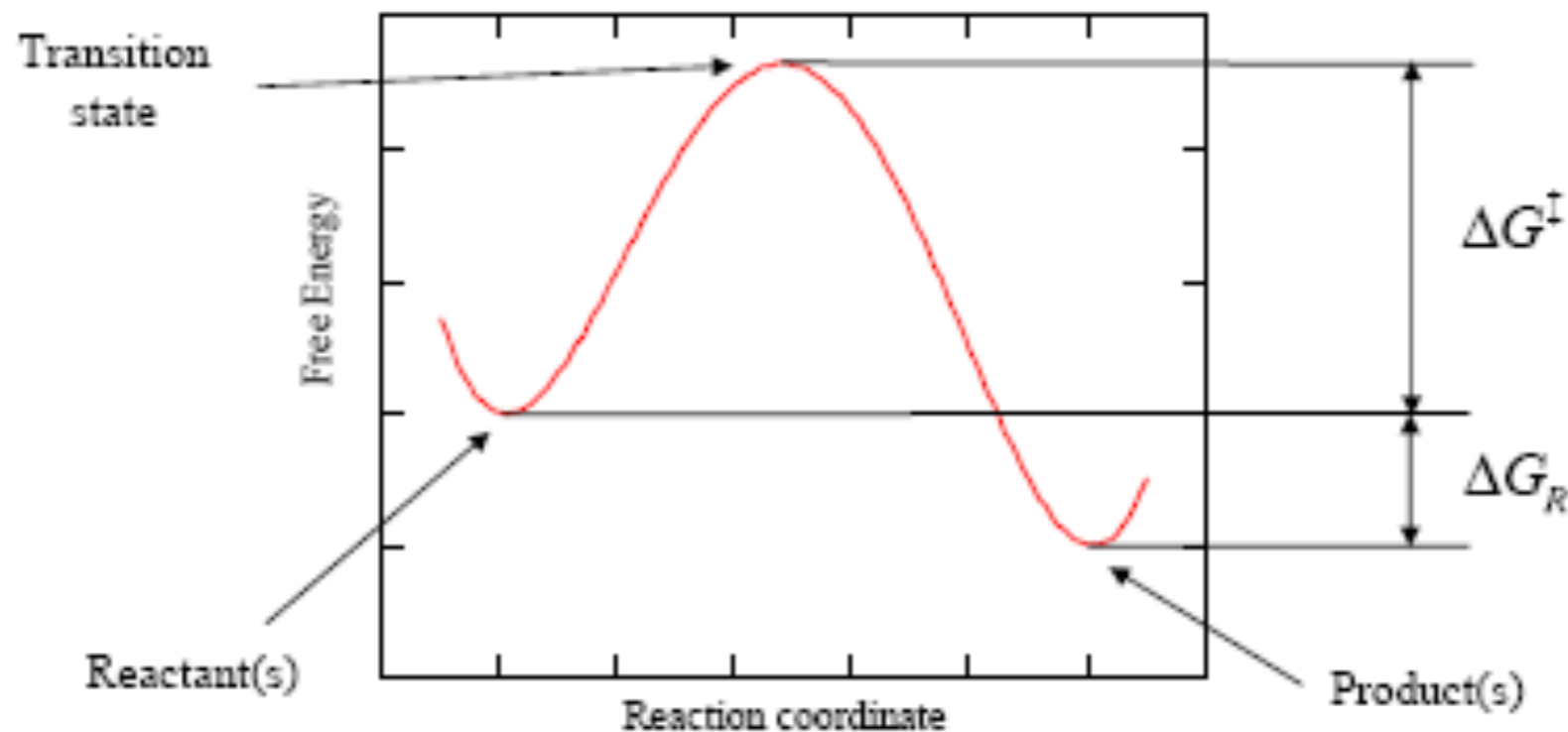
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# Today's Menu

- String method
  - \* **Reminder: basics and implementation**
  - \* **Abinit variables, testing and uses**
- Random Approach: Genetic Algorithm
  - \* **Definition. Rules, fitness function, etc**
  - \* **Implementation in Abinit**
- Generalizations for GA
- Conclusions and perspectives

# String Method

- Method to Study rare events by finding a path between two given states. To find energy barrier and transition state.
- Given a potential, you can find a minimum energy path
- Processes that are more energetic than entropic



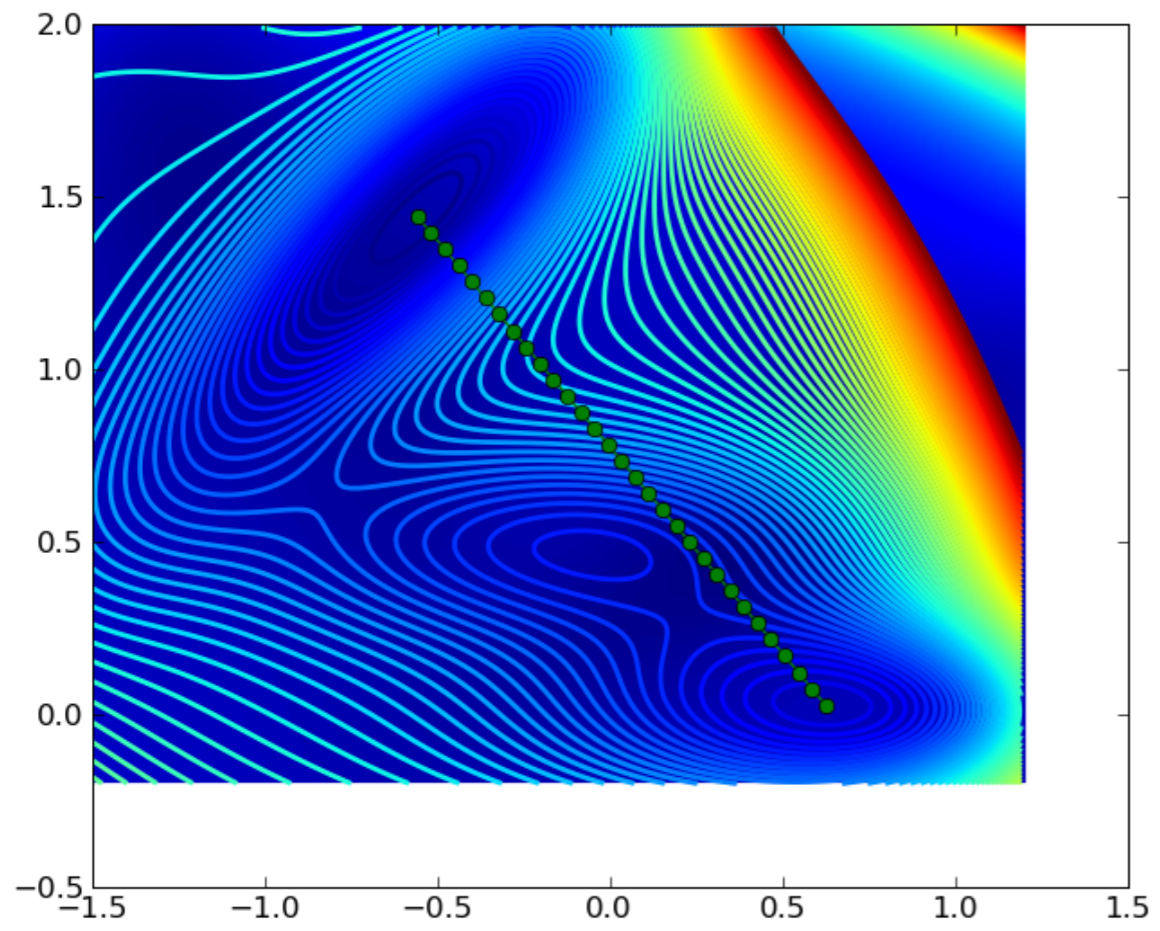
# Finding reactions mechanisms

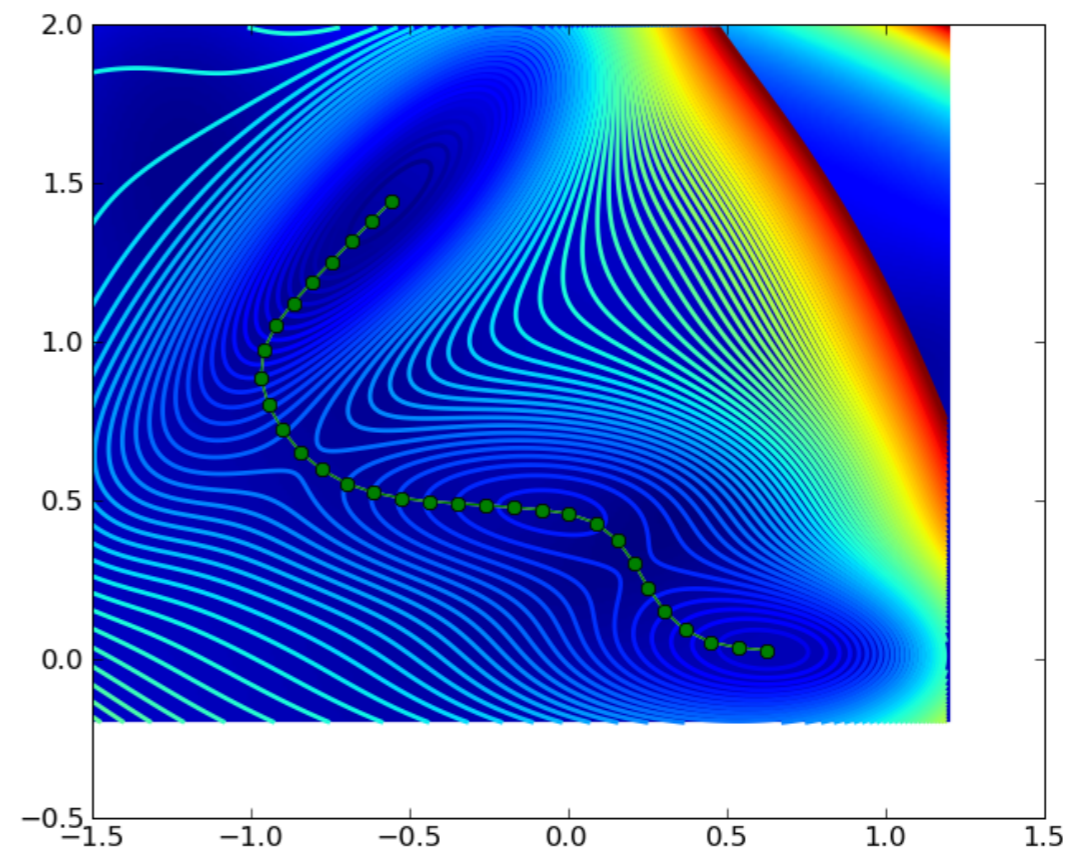
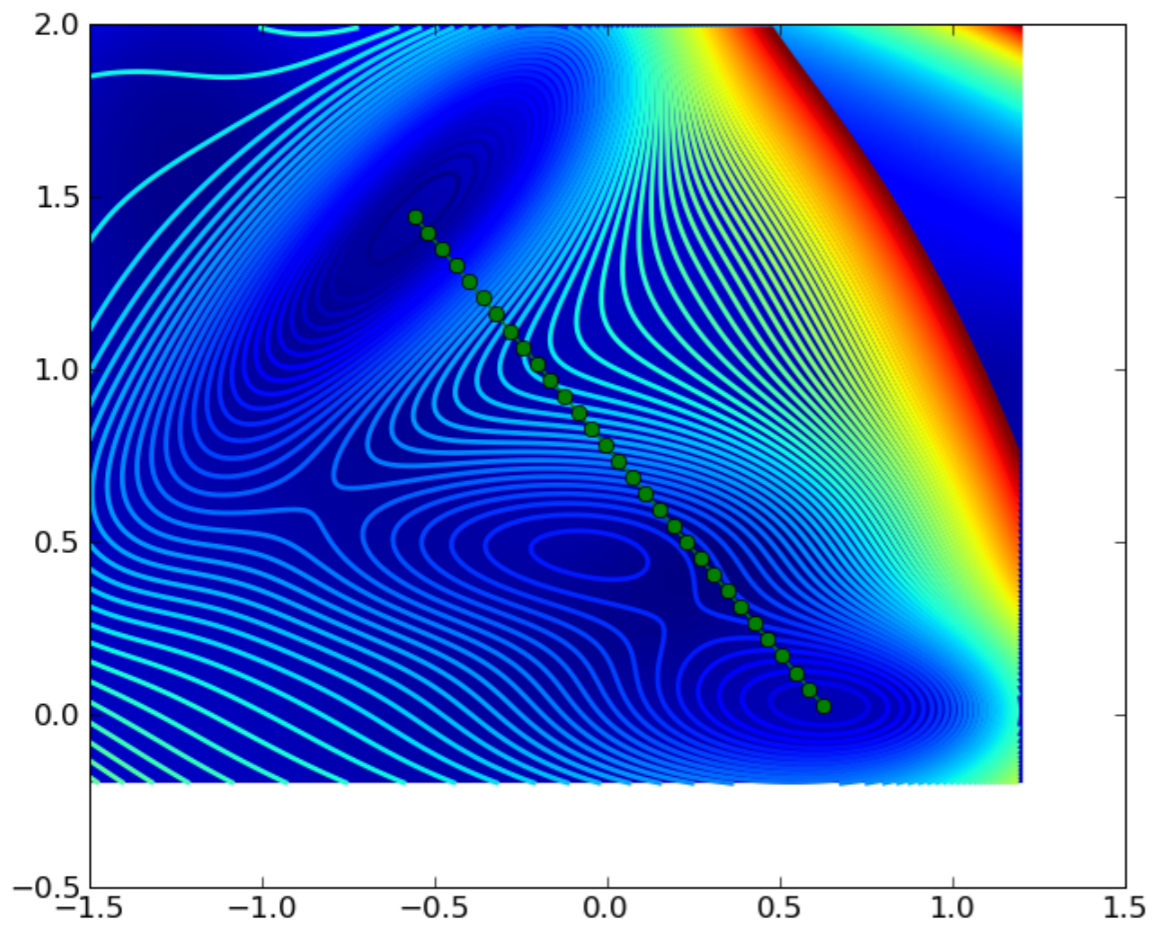
- Minimize a reaction path (a string) by evolving it according to the equation

$$\frac{d\varphi_i}{dt} = -\nabla V(\varphi_i)$$

- To improve the resolution and decrease the changes of changing “too much”, the images positions are “reparametrized” along the path, such that Euclidean distance between them is almost constant.
- *NEB is a particular case of SM*

Weinan *et al.* J. Chem. Phys. 126, 164103 (2007)





# Abinit instructions

- ▶ *ntimimage* 15
- ▶ *nimage* 10
- ▶ *imgmv* 2
- ▶ *chksymbreak* 0
- ▶ *tolimg* 1.d-5
- ▶ *dynimage* 0 8\*1 0
- ▶ *xcart*           3\*0d0           3\*0.766911
- ▶ *xcart\_lastimg* 3\*0d0   2.233089 2\*0.766911

# Genetic Algorithm: survival of the fitness

- First proposed by Holland in “*Adaption in Natural and Artificial Systems*” (1992).
- Organisms reproduce themselves in offsprings which are similar but not equal, due to randomness (**mutation**) and sexual reproduction (**crossing**), which gives characteristics from their parents.
- Finally, the best adapted to the “environment” will survive to every generation, the population will become “better” and “better”
- This idea gives a robust search and optimal algorithm for complicated minima problems.
- A population is like a chromosome, which is conformed by genes.



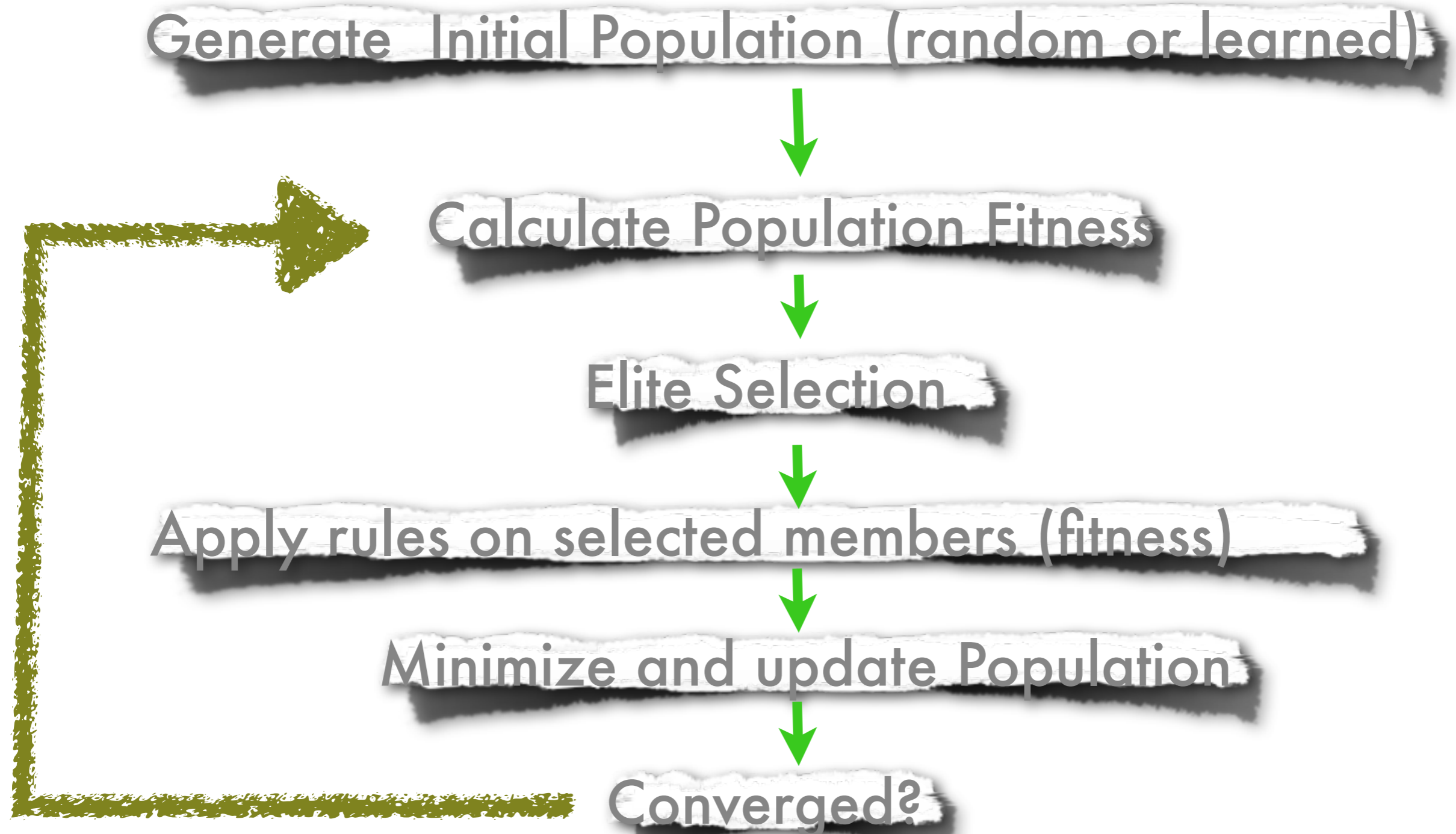
# Basic parts

- Initial configurations. Usually random with no bias. Some input from “nice” configurations is welcome.
- Representation: **binary or real**
- Fitness function: function to be minimized and assigned to every population member.
- Selection. **Stochastic based on Fitness.**
- Rules: crossover, mutation, slicing, etc
- Stagnation: problems biasing the search
- Population number: Usually fixed but it can also vary.

**Deaven and Ho PRL 75:288 (1995)**

# Genetic Algorithm: survival of the fitness

- A basic flow diagram is as follows:



# Representation, initialization and fitness

- Every member of the population is represented in strings. Positions of atoms form a  $3N$  vector.
- Every member is randomly initialized but also good initial configurations can help. (Lego pieces!).
- **How “fit”** is the criteria used for reproduction preventing early good structures to dominate.
- We use total energy (ideal enthalpy).

- Several Weights

$$P_i = \frac{E_i - E_{Min}}{E_{Max} - E_{Min}}$$

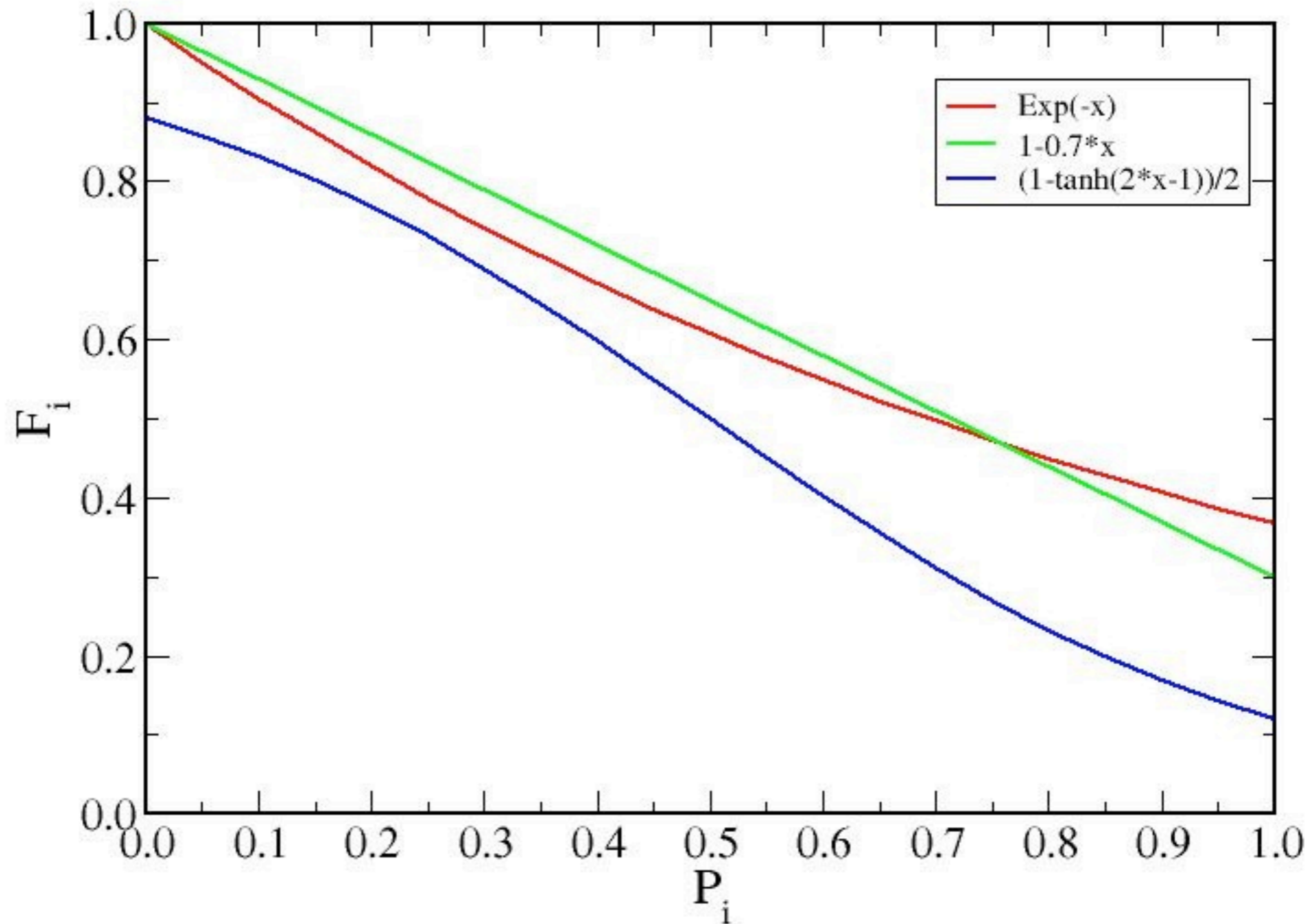
$$F_i = \text{Exp}(-3P_i) \quad F_i = 1 - 0.7 * P_i \quad F_i = \frac{1 - \tanh(2P_i - 1)}{2}$$

**R.L. Johnston, Dalton Trans. page 4193 (2003)**

# Representation, initialization and fitness

- Every n atoms for
- Every n configurations
- How “fitness” is defined
- We use
- Several

$$F_i = E x$$



**R.L. Johnston, Nature Trans. page 4175 (2005)**

# Selection

- Members of the population are selected for reproduction or mutation according to its fitness
- Prevent trapping of local minimum (high selectivity).
- Use of Elitism. The “super-population”.
- Use probability for selection  $\phi_i = \frac{F_i}{\sum F_i}$  with a cumulative probability between *ith* and *(i+1)th*
- There are other selection methods like the tournament (two members are selected and the one with the lowest fitness is chosen)

# Rules of GA

## Two point crossover



1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
---	---	---	---	---	---	---	---	---	----	----	----	----	----	----	----	----	----	----	----

1	2	3	4	9	8	7	6	5	10	11	12	13	14	15	16	17	18	19	20
---	---	---	---	---	---	---	---	---	----	----	----	----	----	----	----	----	----	----	----

## Random Crossing

Parents

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
a	b	c	d	f	g	h	i	j	k	l	m	n	o	p	q	e	s	t	u



Children

a	2	3	d	5	6	h	8	j	10	11	m	13	14	p	16	17	s	19	20
1	b	c	4	f	g	7	i	9	k	l	12	n	o	15	q	e	s	t	u

## N-point crossover

Parents

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
a	b	c	d	f	g	h	i	j	k	l	m	n	o	p	q	e	s	t	u



Children

1	2	3	4	5	6	7	8	9	10	11	12	n	o	p	q	e	s	t	u
a	b	c	d	f	g	h	i	j	k	l	m	13	14	15	16	17	18	19	20

## Mutation (coordinates or cell parameters, volume)

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
---	---	---	---	---	---	---	---	---	----	----	----	----	----	----	----	----	----	----	----



1	2	3	4	5	9	7	8	9	10	14	12	13	12	15	16	17	18	16	20
---	---	---	---	---	---	---	---	---	----	----	----	----	----	----	----	----	----	----	----

# Arithmetic Average

Parents

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
21	22	23	24	25	26	27	28	29	30	31	32	33	34	35	36	37	38	39	40



Child

11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29	30
----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----	----

# Geometric Average $\sqrt{a_i \cdot b_i}$

Parents

1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
21	22	23	24	25	26	27	28	29	30	31	32	33	34	35	36	37	38	39	40



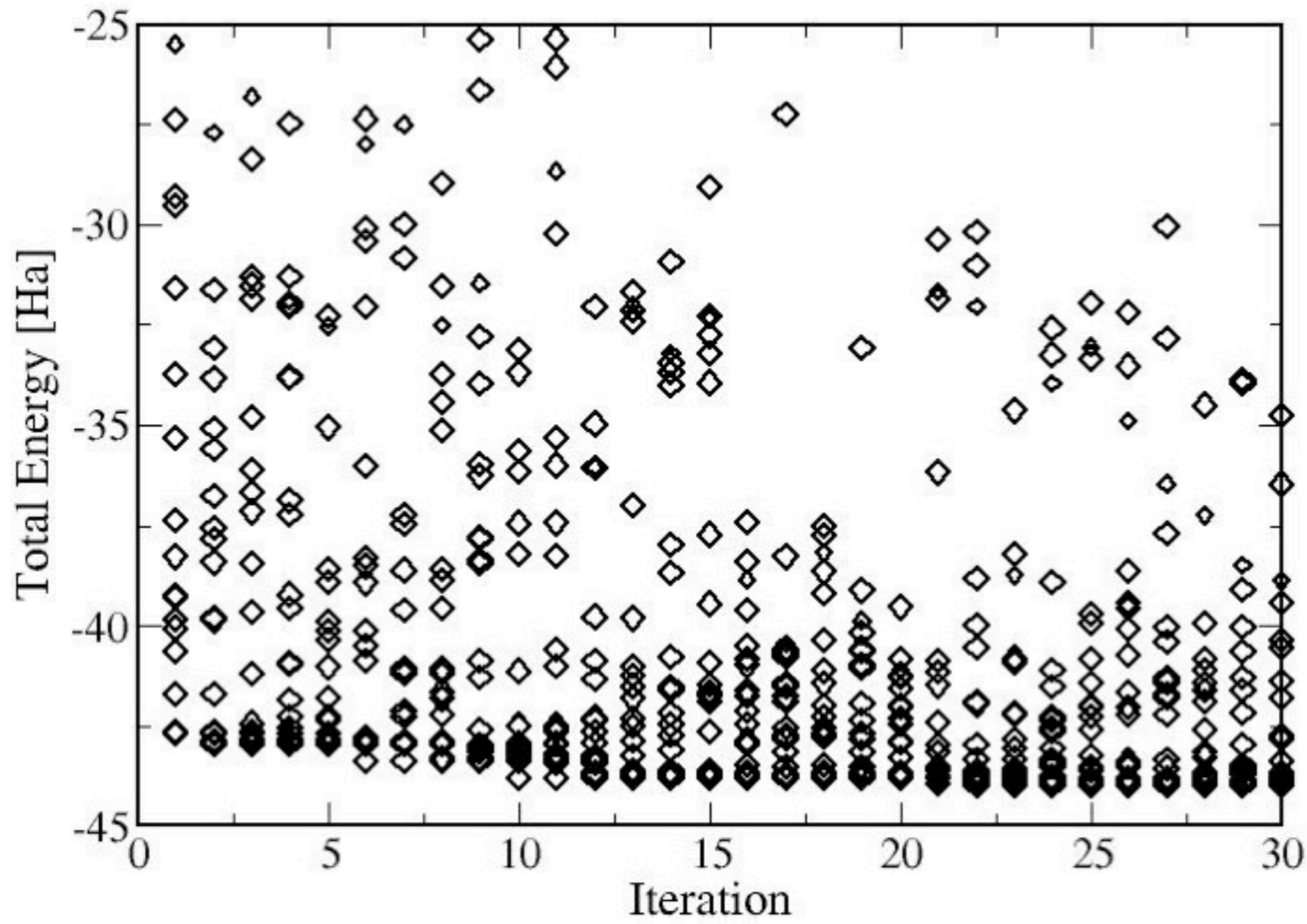
4.58	6.63	8.31	9.79	11.18	12.49	13.75	14.96	16.15	17.32	18.47	19.60	20.71	21.81	22.91	24.00	25.08	26.15	27.22	28.28
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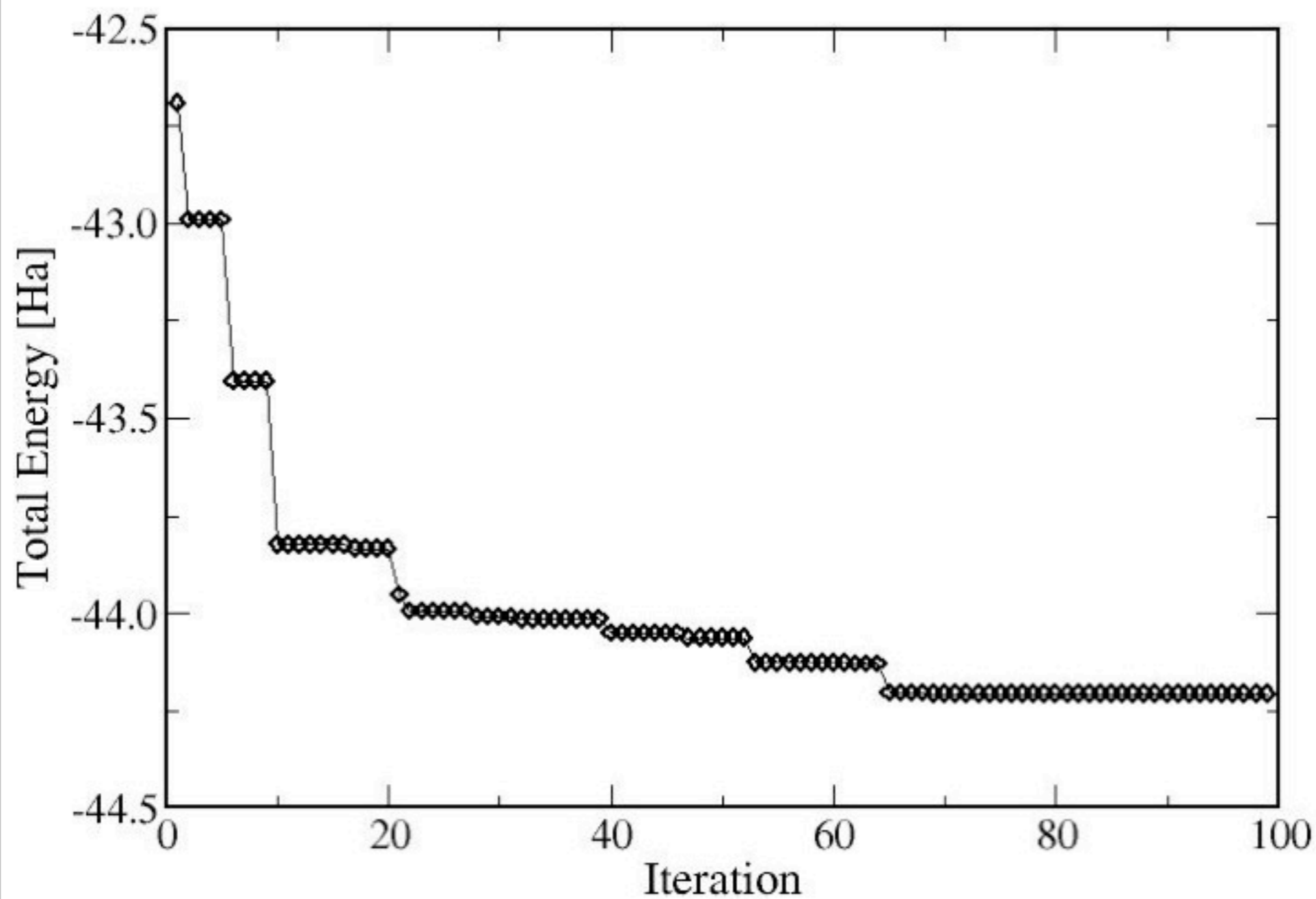


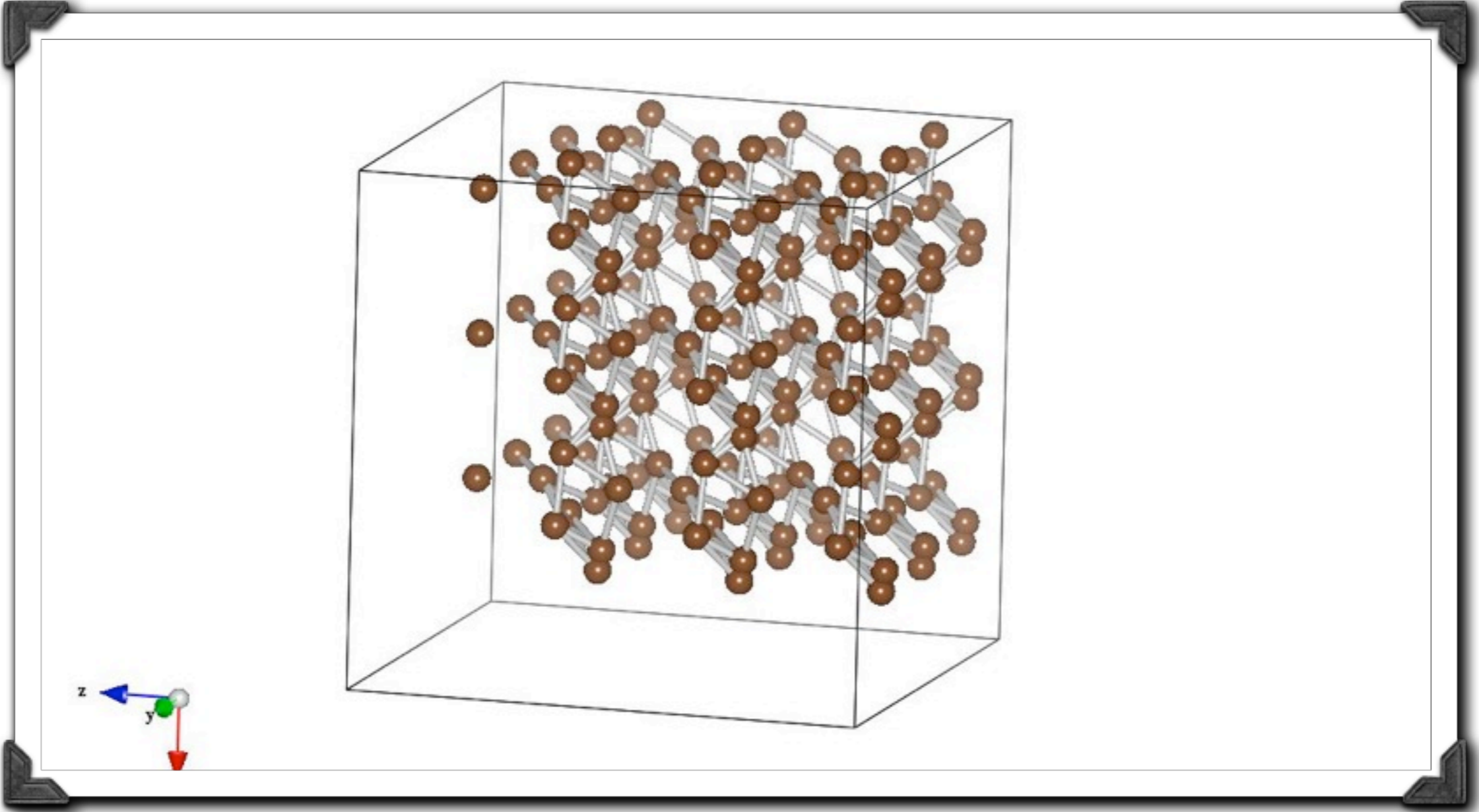
# Additional issues: changes and constraints

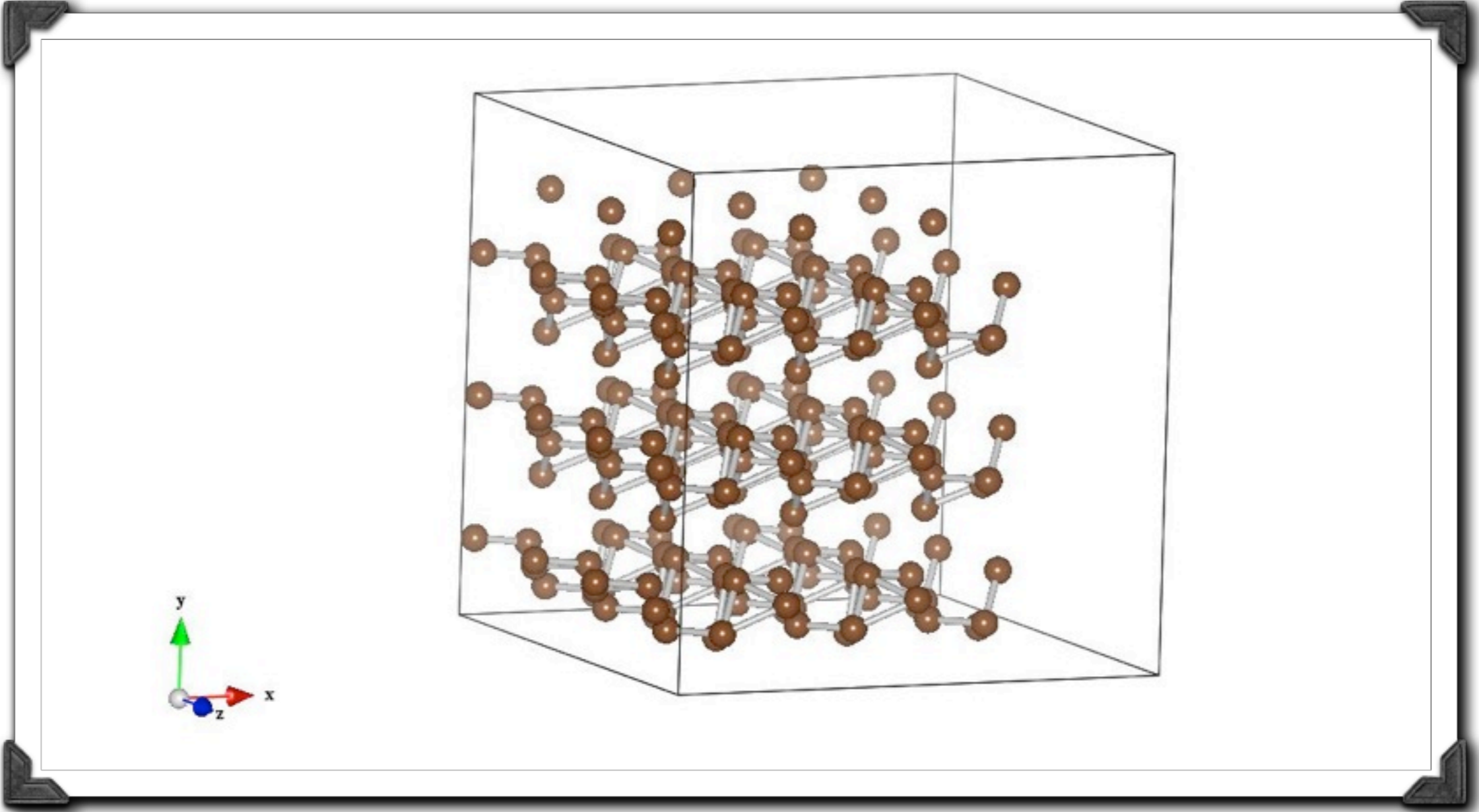
- Random change in cell volume
- Check atomic distances after a new child has been created (close atoms less than a given distance or same position)
- Check configurational distance.
  - ✓ Find distance from origin
  - ✓ rearrange distances from large to small ( $d_1 < d_2 < d_3 < \dots < d_n$ )
  - ✓ Find distance between two configurations

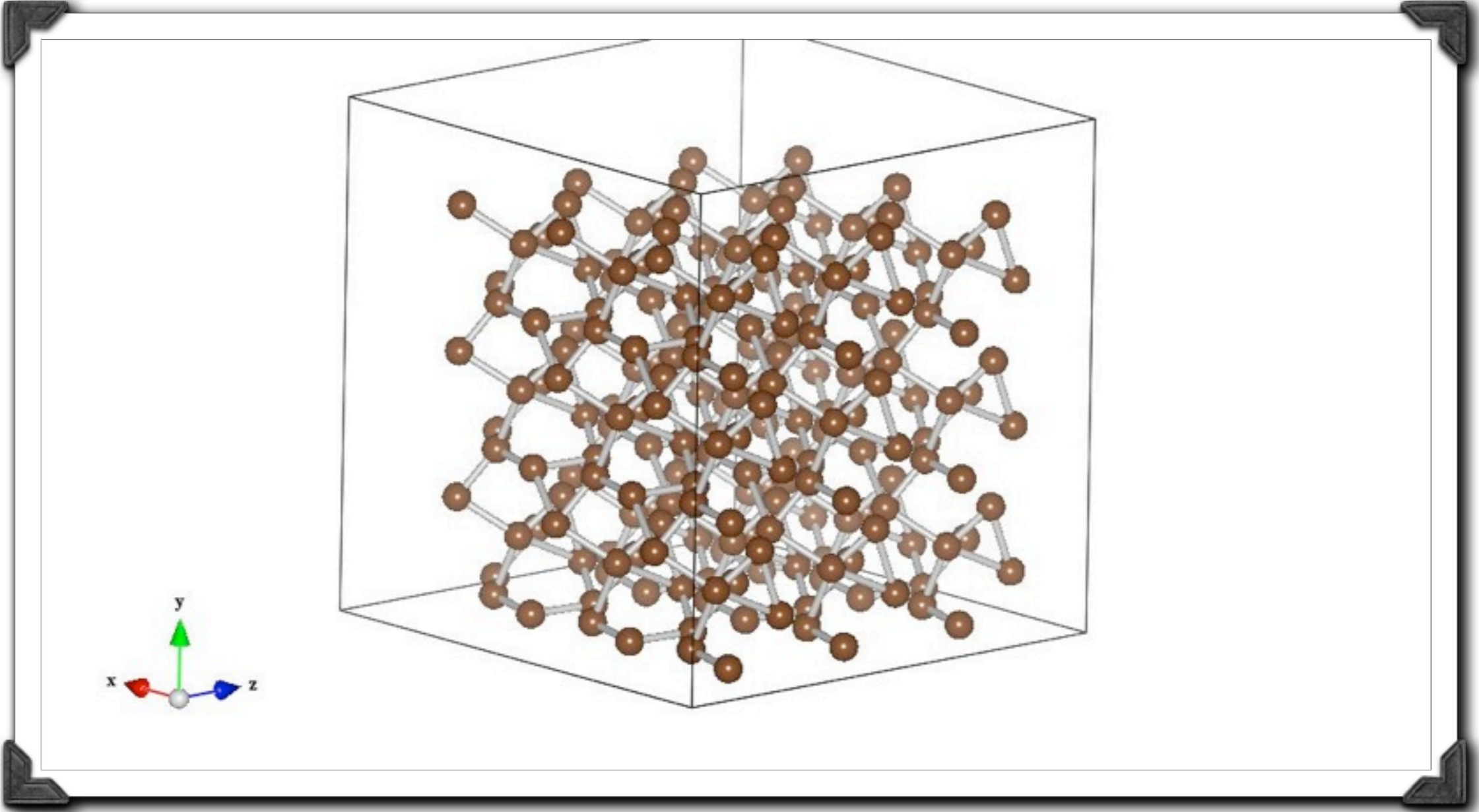
$$\sum_{i=1}^N |d_i^\alpha - d_i^\beta|$$







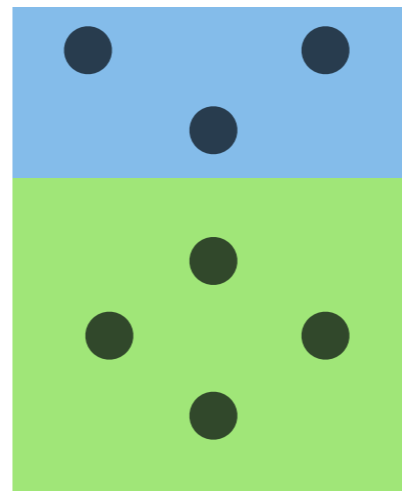
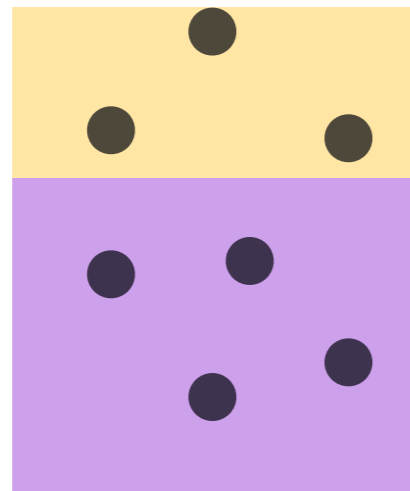




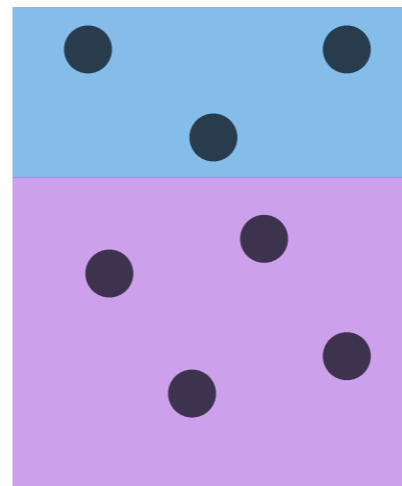
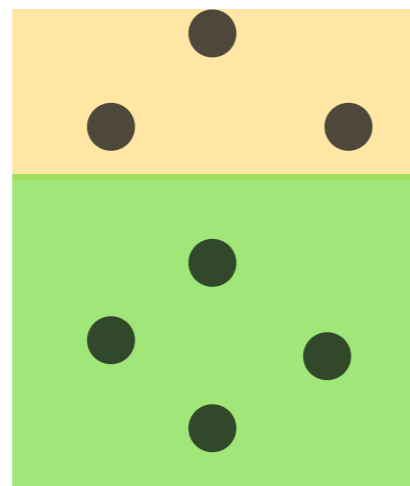
# Still.. not enough

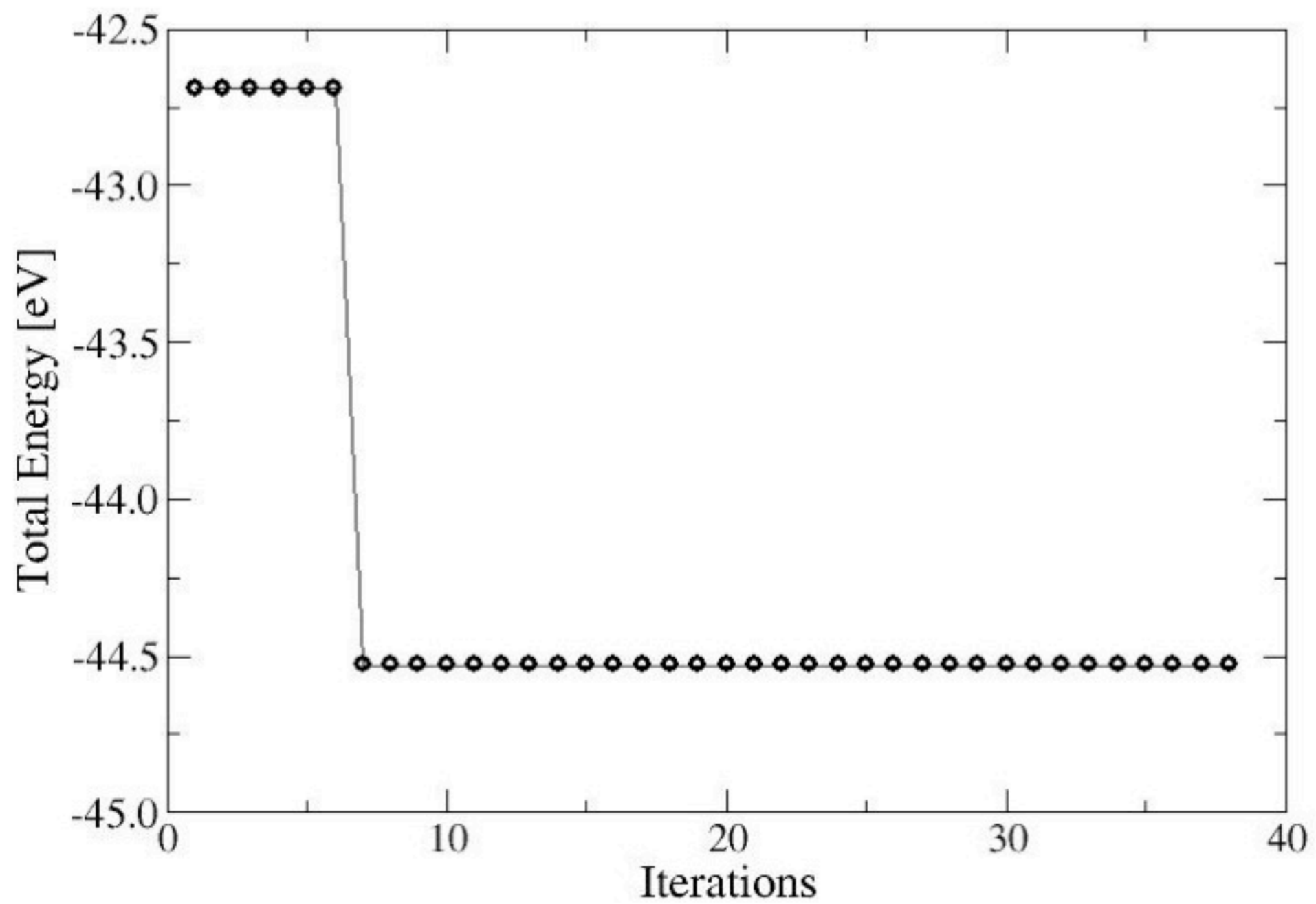
- Slicing method (in the unitary cell)

Parents

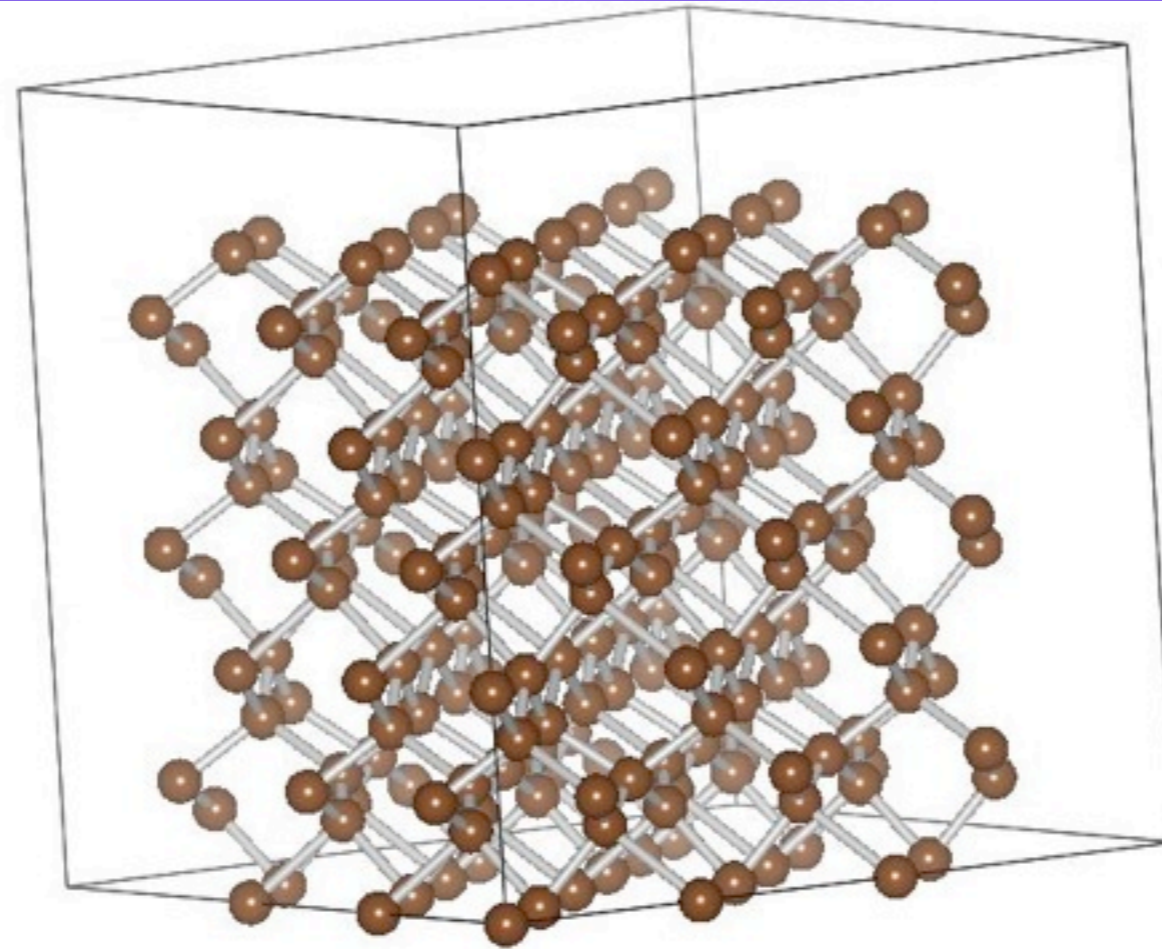
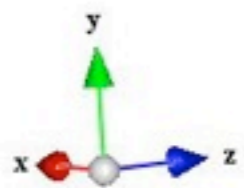


Children









# Input example GA

ntimimage 100  
optcell 0  
dilatmx 1.3  
nimage 20  
chksymbreak 0  
random\_atpos 2  
imgmov 4

acell 3\*6.0d0  
natom 8  
nstep 100  
ntypat 1  
typat 8\*1  
znucl 6  
kptopt 1  
ngkpt 3 3 3

# Future and status (Marc's bar!)

- Include cell deformation. (optcell 2+ dilatmx) Almost!
- Use the Enthalpy. Almost!
- Parallelization. Almost!
- Lego approach. Typical structures. Maybe from  $g(\mathbf{r})$
- Generalization for several species (some rules needs to be modified)
- Include different fitness functions
- Etching technique.  $N$  is not constant. (Landman)
- Improve searching: Bee algorithm or graph theory