

# REVISIÓN DE MODELOS Y ALGORITMOS DE PREDICCIÓN DE REDES CRISTALINAS

Trabajo de grado de Ingeniería Física  
Iván David Buitrago Torres  
Dirigido por Ph.D. Elisabeth Restrepo

PCM Computational Applications

# Justification

- Knowledge and understanding of the methods
- Prediction of structures unknown to the date
- Meta-Stable Structures (Optimization of physical properties)
- Cost reduction

# Objectives

- General:
- Make a review of the methods that exists and are being developed for the prediction of crystal structures.
- Specifics:
- Carry out a bibliographical search of the methods developed for the prediction of crystal structures.
- Explain conceptually and systematically the process used in each method

# Crystal Structure Prediction

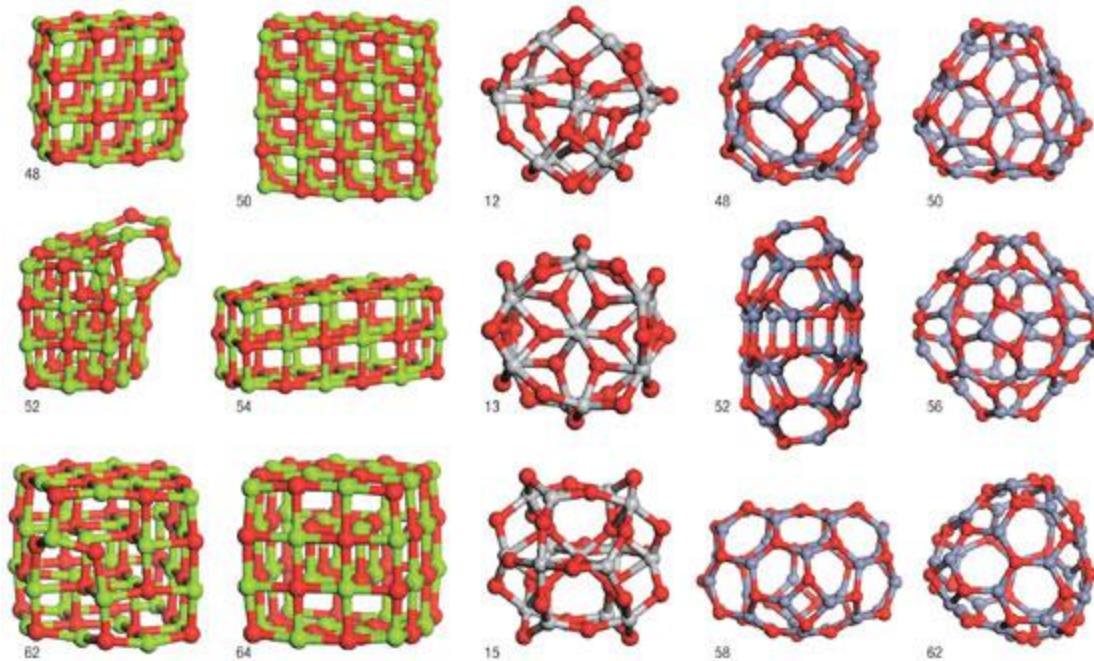
- Prediction  $\neq$  Simulation

Prediction: Ab Initio Calculations in order to obtain structures.

Simulation: Experimental Structural Data in order to obtain physical properties.

# Crystal Structure Prediction

- Jhon Maddox, 1988: "One of the continuing scandals in the physical sciences is that it remains in general impossible to predict the structure of even simplest crystalline solids from a knowledge of their chemical composition."



Taken from Crystal structure prediction from first principles, Scott M. Woodley & Richard Catlow

# Crystal Structure Prediction

- Crystal structure prediction by evolutionary algorithms and direct search methods (Diploma thesis) Bc. Oto Kohulák.  
Univerzita Komenského v Bratislave:

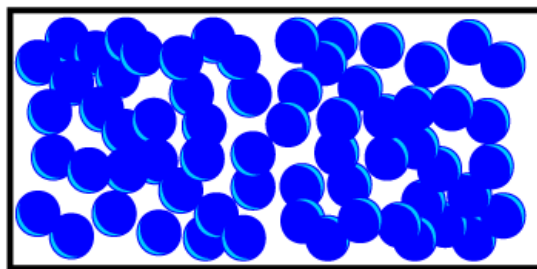
Thermodynamics: Functions of free energy: Gibbs, Enthalpy, Helmholtz, etc...

Minimum Energy = Stable Structure = Thermodynamical Equilibrium

# Crystal Structure Prediction

- Why not one by one?
- We can try to reach minimum by fixing every position but...
- $10^{23}$  particles per  $cm^2$

1 mole of sodium



has  $6.023 \times 10^{23}$  atoms

equipmentexplained.com

Taken from [http://chemistryhungergames.weebly.com/uploads/8/8/4/9/8849208/190572\\_orig.gif](http://chemistryhungergames.weebly.com/uploads/8/8/4/9/8849208/190572_orig.gif)

# Crystal Structure Prediction

- Then what?...
- Establish energy function, number of particles, and chemical/thermodynamical parameters.
- Use energy term as objective function.
- Apply Algorithms.



# Crystal Structure Prediction

- Constrained and unconstrained Gibbs free energy minimization in reactive systems using genetic algorithm, Adrián Bonilla-Petriciolet.
- At constant temperature T and pressure P, for c multi component and  $\pi$  multi phase system:

$$G = \sum_{j=1}^{\pi} \sum_{i=1}^c \eta_{ij} \mu_{ij}$$

$\eta$  = number of moles of component  $i$  in phase  $j$   
 $\mu$  = chemical potential of component  $i$  in phase  $j$

- «variables are based on the transformation of physical compositions, restrict the solution space to the compositions that satisfy stoichiometry requirements, and also reduce the dimension of the composition space by the number of independent reactions satisfying the Gibbs phase rule.»

$$F = C - P + 2$$

# Crystal Structure Prediction

- Crystal structure prediction by evolutionary algorithms and direct search methods (Diploma thesis), Bc. Oto Kohulák
- «To have sufficient precision for our calculations, we cannot use classical potentials, on the other hand full quantum mechanical calculations with multielectron wave functions are too time consuming. So we must perform quantum calculations in clever way.»

$$H = - \sum_e \frac{\hbar^2}{2m_e} \Delta_{\mathbf{r}_e} - \sum_i \frac{\hbar^2}{2m_i} \Delta_{\mathbf{r}_i} + \frac{1}{2} \sum_{e \neq e'} \frac{e^2}{|\mathbf{r}_e - \mathbf{r}_{e'}|} + \frac{e^2}{2} \sum_{i \neq i'} \frac{Z_i Z_{i'}}{|\mathbf{r}_i - \mathbf{r}_{i'}|} + \frac{e^2}{2} \sum_{e,i} \frac{Z_i}{|\mathbf{r}_e - \mathbf{r}_i|},$$

$$n(\mathbf{r}) = \sum_i \int_{\sigma} g_i |\psi_i(\mathbf{r}_{\sigma})|^2,$$

# Global Optimization Algorithms

● *Global optimization algorithms, Thomas Weise*

*The optimization is a branch of applied mathematics*

$$f: X \rightarrow Y, f \in F$$

Classification of optimization methods:

By operation mode:

- Deterministics
- Stochastic
- Heuristic
- Metaheuristic

By its properties:

- Online
- Offline

# Some definitions

● Solution candidates/phenotypes:

$$\mathbf{x} \in X$$

Solution space  $S$ :

$$X^* \subseteq S \subseteq X$$

Optimal Set:

$$x^* \in X^*$$

Search space / Genome:

$$g \in G$$

Gene: Distinguishable unities in genotypes

# Some definitions

## Search Operations:

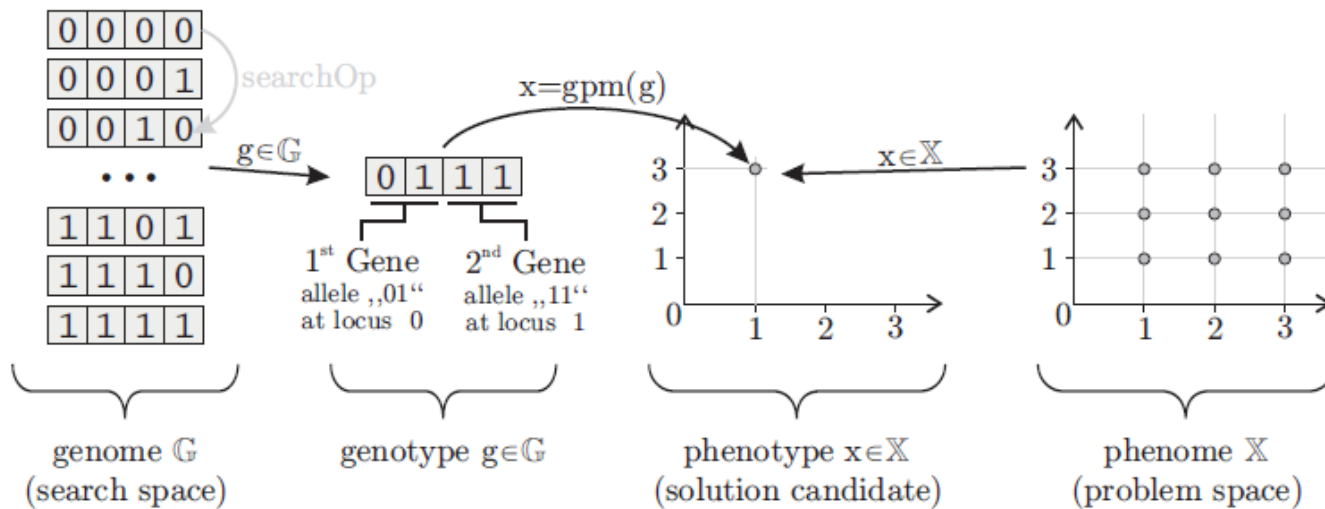
searchOp allow to solve a given problem in a defined set with different arities

Genotype/Phenotype Mapping: Phenomena to Math interpretation

$$gpm: G \rightarrow X$$

$$\forall g \in G \exists x \in X: gpm(g) = x$$

An optimization problem is well defined if  $X$ ,  $F$ ,  $G$ ,  $Op$  and  $gpm$  are known



# Some definitions

● Objective space: Codomain of all objective functions

$$F = \{f_i: X \rightarrow Y_i: 0 < i < n, Y_i \in R\} \rightarrow Y = Y_1 x Y_2 \dots x Y_n$$

Fitness: Measure of «good» or «bad» (Only have meaning during the process)

$$v(x) \in V: X \rightarrow V, V \in R^+$$

Individual:

$$p.x = gpm(p.g)$$

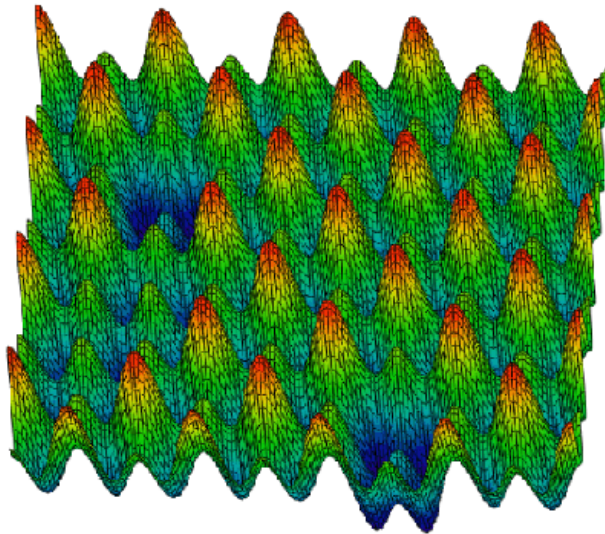
# Some definitions

Population:

$$Pop \in G \times X: \forall p(p.g, p.x) \in Pop \rightarrow p.x = gpm(p.g)$$

**Problem Landscape:** Maps  $X$  to a Probability

$$\Phi: X \times N \rightarrow [0,1] \in R^+$$



Taken from

[http://news.princeton.edu/uploads/296/image/model\\_landscape.gif](http://news.princeton.edu/uploads/296/image/model_landscape.gif) 10/03/2017

# Some definitions

● Optimization Algorithm: Transformation

Will find  $x^*$  if:

- searchOp is partially ordered
- gpm is surjective

$$(X, F, SearchOp, G, gpm) \rightarrow \Phi$$

$$\exists x_i^* \in X \rightarrow \lim_{t \rightarrow \infty} \Phi(x_i^*, t) = 1$$



# Multiobjective functions

• Weighted sums:

$$g(x) = \sum_i \omega_i f_i(x)$$

Pareto optimization: Domination

$$\begin{aligned} x_1 \vdash x_2 \iff & \forall i: 0 < i < n \rightarrow \omega_i f_i(x_1) \leq \omega_i f_i(x_2) \\ & \wedge \exists j: j < n \rightarrow \omega_j f_j(x_1) < \omega_j f_j(x_2) \end{aligned}$$

# Multiobjective functions

● Pareto optimal: Element which is not dominated by any other.

$$x^* \in X^* \leftrightarrow \nexists x \in X: x \vdash x^*$$

Prevalence comparator function: Generates Optimal Set  $X^*$ , compares elements of  $X$

$$(x_1 \succ x_2) \text{ if } cmp_F(x_1, x_2) < 0$$

$x_1$  prevails over  $x_2$

$$x^* \in X^* \leftrightarrow \nexists x \in X: x = x^* \wedge x \vdash x^*$$

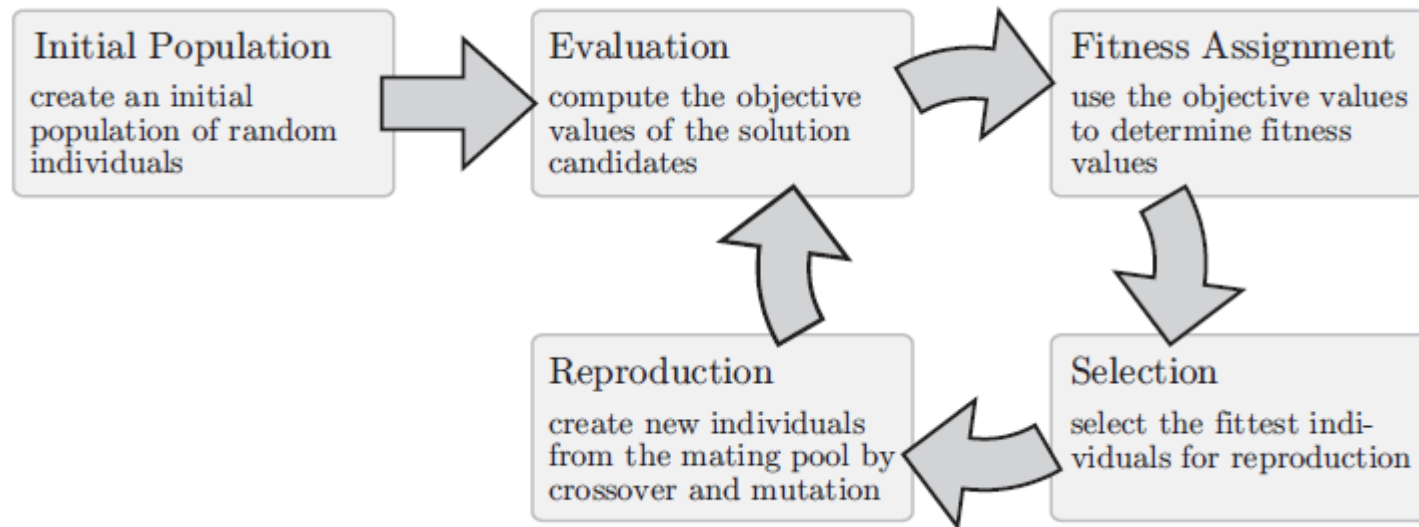
# Evolutionary Algorithms

- Based in 10 statements made by Charles Darwin in his 1859 book «On the origin of the species»
  1. Big fertility -> More offspring
  2. Population is constant
  3. Food is limited but stable
  4. Individuals fight for food -> Fight for survival
  5. Not 2 identical individuals in sexual reproductions
  6. Variations affect fitness hence survival
  7. Some variations are hereditary
  8. Less fitness, less offspring
  9. Individual survive and reproduce, transmit features
  10. Species change along time

# Evolutionary Algorithms

- Search space is all DNA chains.
- Fitness is measured depending of objective function.

# Evolutionary Algorithm Scheme



# Evolutionary Algorithm Scheme

- Initial population (Pop) are individuals  $p$  with genome  $p.g$
- Each individual  $p.x$  is evaluated in  $f$
- Comparison between evaluations with  $cmpf$  set fitness
- In reproduction phase  $p.g$  genomes are combined randomly through a  $searchOp$  and offspring is combined with Pop to form next generation.
- Termination criterion is evaluated

# Evolutionary Algorithm Implementation

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**Algorithm 2.1:**  $X^* \leftarrow \text{simpleEA}(\text{cmp}_F, ps)$

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**Input:**  $\text{cmp}_F$ : the comparator function which allows us to compare the utility of two solution candidates

**Input:**  $ps$ : the population size

**Data:**  $t$ : the generation counter

**Data:**  $Pop$ : the population

**Data:**  $Mate$ : the mating pool

**Data:**  $v$ : the fitness function resulting from the fitness assigning process

**Output:**  $X^*$ : the set of the best elements found

```
1 begin|
2    $t \leftarrow 0$ 
3    $Pop \leftarrow \text{createPop}(ps)$ 
4   while  $\neg \text{terminationCriterion}()$  do
5      $v \leftarrow \text{assignFitness}(Pop, \text{cmp}_F)$ 
6      $Mate \leftarrow \text{select}(Pop, v, ps)$ 
7      $t \leftarrow t + 1$ 
8      $Pop \leftarrow \text{reproducePop}(Mate)$ 
9   return  $\text{extractPhenotypes}(\text{extractOptimalSet}(Pop))$ 
10 end
```

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# Implementation

- USPEX (Universal Structure Predictor: Evolutionary Xtallography)

Unexpected Stable Stoichiometries of Sodium Chlorides.

Weiwei Zhang, Artem Oganov.

## Unexpected Stable Stoichiometries of Sodium Chlorides

Weiwei Zhang,<sup>1,2\*†</sup> Artem R. Oganov,<sup>2,3,4\*†</sup> Alexander F. Goncharov,<sup>5,6</sup> Qiang Zhu,<sup>2</sup> Salah Eddine Boulfelfel,<sup>2</sup> Andriy O. Lyakhov,<sup>2</sup> Elissaios Stavrou,<sup>5</sup> Maddury Somayazulu,<sup>5</sup> Vitali B. Prakapenka,<sup>7</sup> Zuzana Konôpková<sup>8</sup>

Sodium chloride (NaCl), or rocksalt, is well characterized at ambient pressure. As a result of the large electronegativity difference between Na and Cl atoms, it has highly ionic chemical bonding (with 1:1 stoichiometry dictated by charge balance) and B1-type crystal structure. By combining theoretical predictions and diamond anvil cell experiments, we found that new materials with different stoichiometries emerge at high pressures. Compounds such as Na<sub>3</sub>Cl, Na<sub>2</sub>Cl, Na<sub>3</sub>Cl<sub>2</sub>, NaCl<sub>3</sub>, and NaCl<sub>7</sub> are theoretically stable and have unusual bonding and electronic properties. To test this prediction, we synthesized cubic and orthorhombic NaCl<sub>3</sub> and two-dimensional metallic tetragonal Na<sub>3</sub>Cl. These experiments establish that compounds violating chemical intuition can be thermodynamically stable even in simple systems at nonambient conditions.

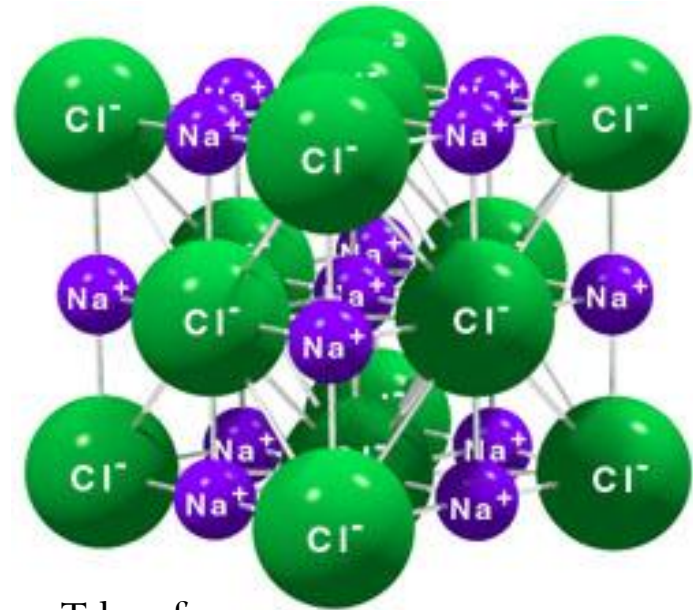


# NaCl

- FCC Structure
- Stable
- Lower Energy State



Taken from  
<http://www.mspbs.gov.py/dvent/wp-content/uploads/2016/02/Sal.png>

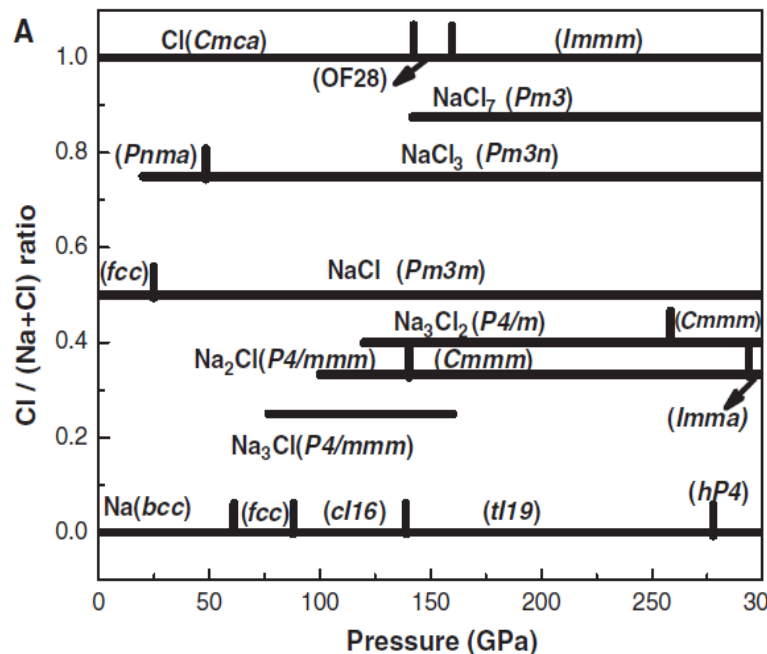


Taken from  
<http://moorechemistry.weebly.com/solvation-process.html>

# NaCl

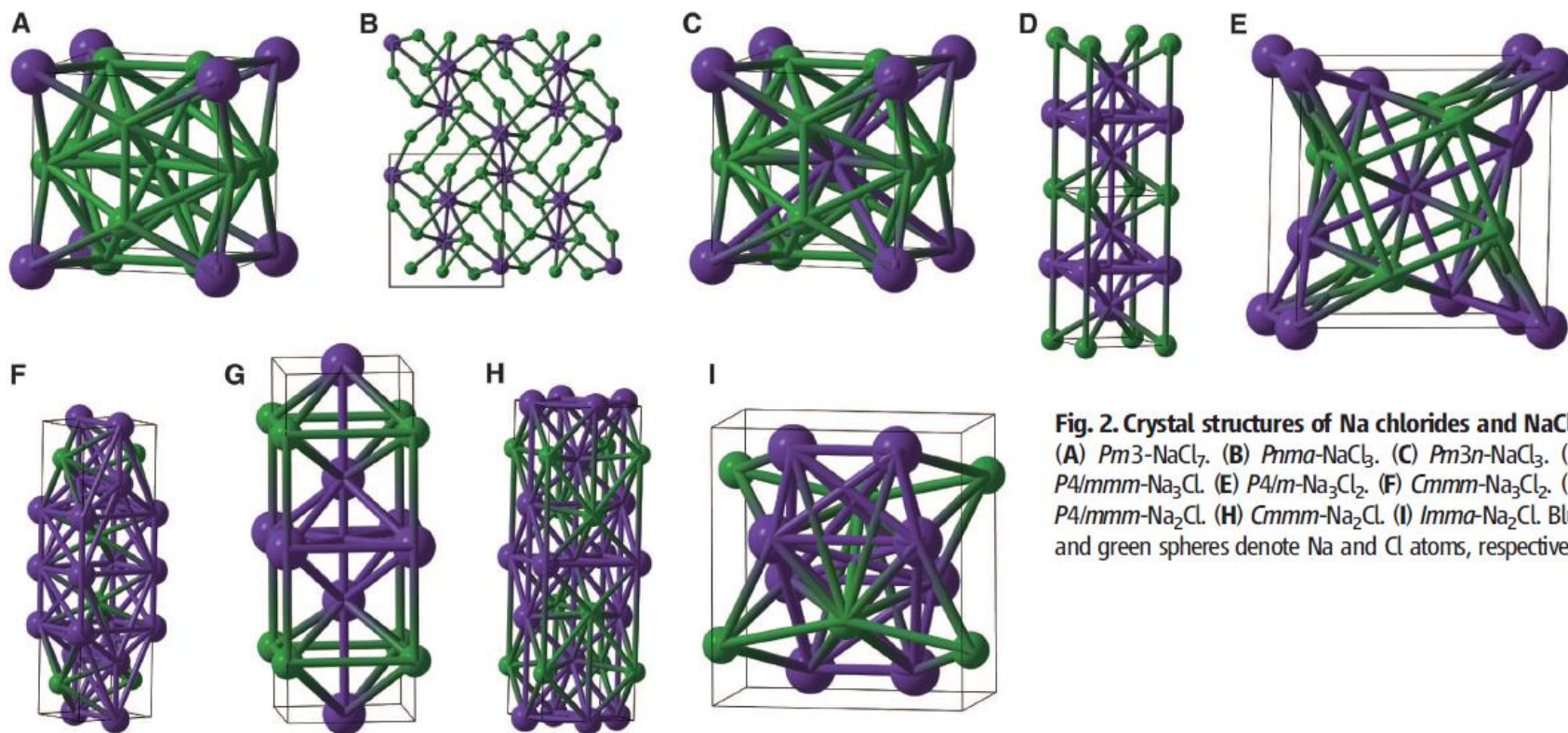
- At high temperatures and high pressures, apportation of energy will be bigger than inner energy of the system.

$$G = U + pV - TS$$



Unexpected Stable  
Stoichiometries of  
Sodium Chlorides.  
Weiwei Zhang, Artem  
Oganov.

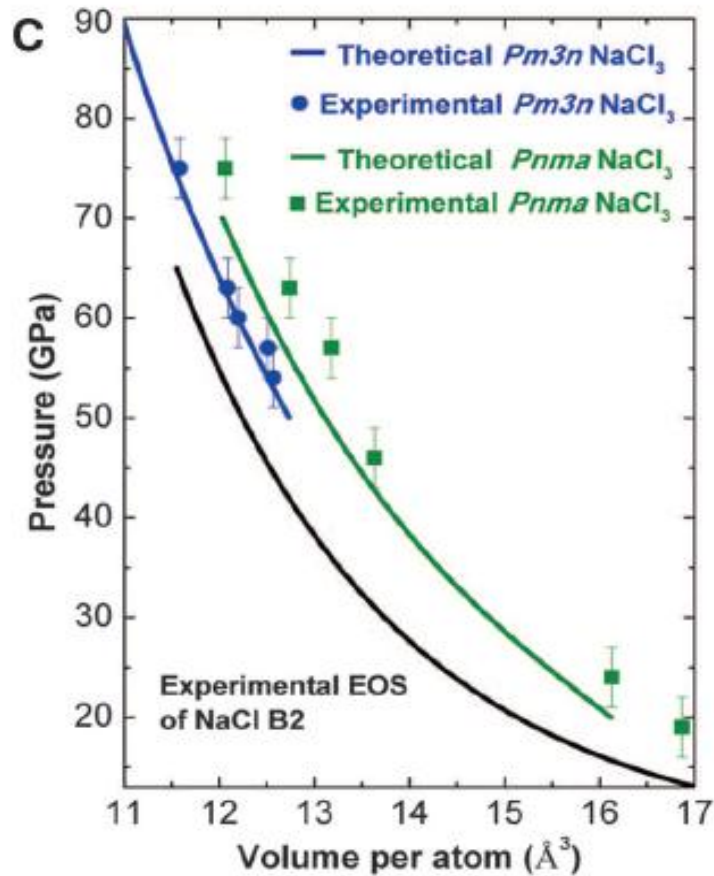
# NaCl



**Fig. 2. Crystal structures of Na chlorides and  $\text{NaCl}_7$ .** (A)  $Pm\bar{3}$ - $\text{NaCl}_7$ . (B)  $Pnma$ - $\text{NaCl}_3$ . (C)  $Pm\bar{3}n$ - $\text{NaCl}_3$ . (D)  $P4/mmm$ - $\text{Na}_3\text{Cl}$ . (E)  $P4/m$ - $\text{Na}_3\text{Cl}_2$ . (F)  $Cmmm$ - $\text{Na}_3\text{Cl}_2$ . (G)  $P4/mmm$ - $\text{Na}_2\text{Cl}$ . (H)  $Cmmm$ - $\text{Na}_2\text{Cl}$ . (I)  $Imma$ - $\text{Na}_2\text{Cl}$ . Blue and green spheres denote Na and Cl atoms, respectively.

Unexpected Stable Stoichiometries of Sodium Chlorides. Weiwei Zhang, Artem Oganov.

# NaCl



Unexpected  
Stable  
Stoichiometries of  
Sodium  
Chlorides.  
Weiwei Zhang,  
Artem Oganov.

# What is next?

- Determine the conceptual bases and the algorithmic structure of another global optimization methods like Simulated Annealing, Particle Swarm Optimization, etc.
- Study the physical variables that describe the behaviour of nanoparticle systems, to establish the global application of these algorithms to different systems.
- Review results obtained with these algorithms.
- Program some of these algorithm at short range for empiric tests.

THANK YOU!