

Generalized Pattern Search Methods for a Structure Determination Problem



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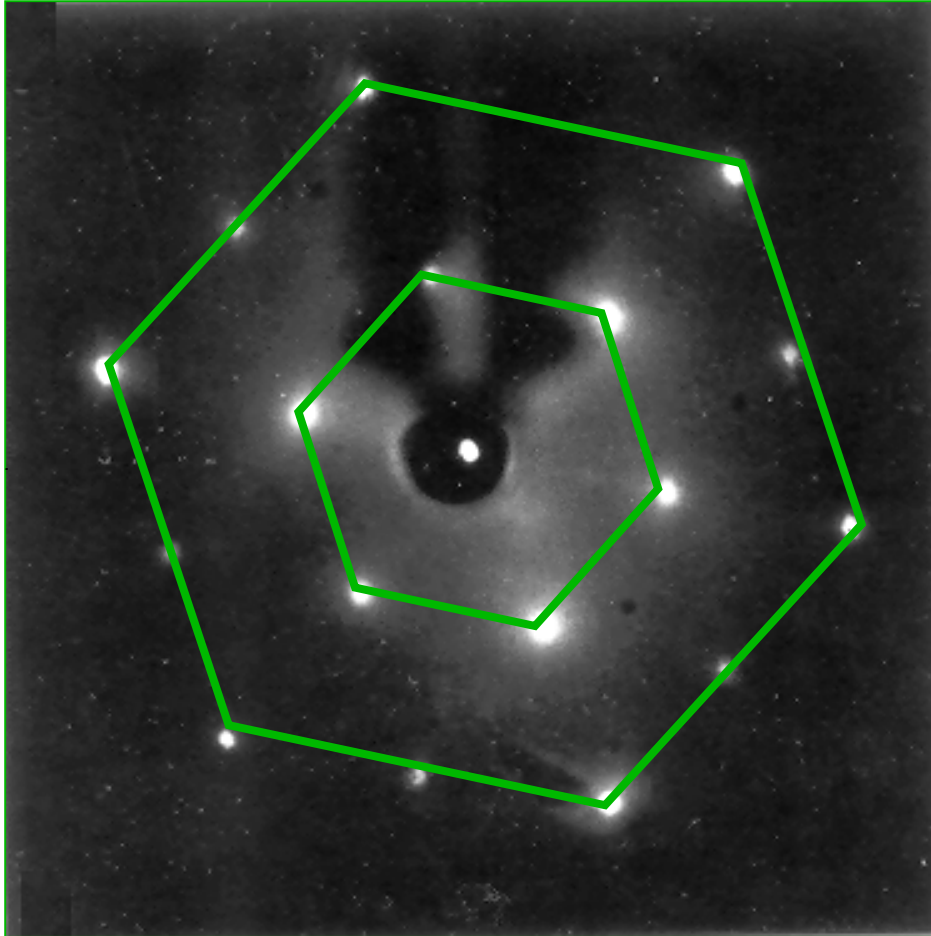
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C O M P U T A T I O N A L R E S E A R C H D I V I S I O N



Low-Energy Electron Diffraction (LEED)



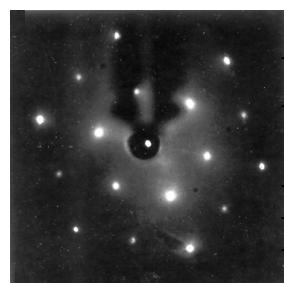
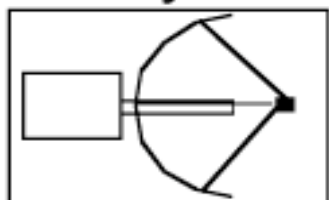
Low-energy electron diffraction pattern due to monolayer of ethynidyne attached to a rhodium (111) surface

- ❖ Goal is to determine surface structure through low energy electron diffraction (LEED)
- ❖ Inverse problem consists of minimizing the error between experiment and theory
- ❖ Combination of local/global optimization
- ❖ Contains both continuous and categorical variables
 - Atomic coordinates
 - Ni, Li
- ❖ Function not smooth; no derivatives available

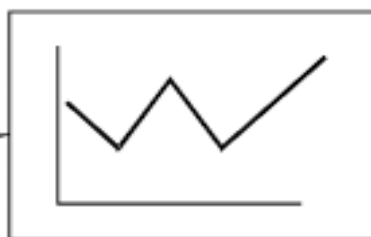
Low Energy Electron Diffraction

Experiment

LEED system



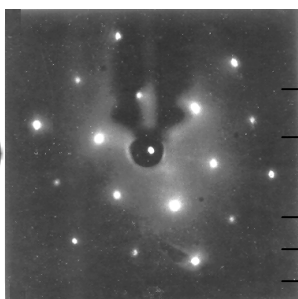
I-V spectra



Theory

(x,y,z) input parameters

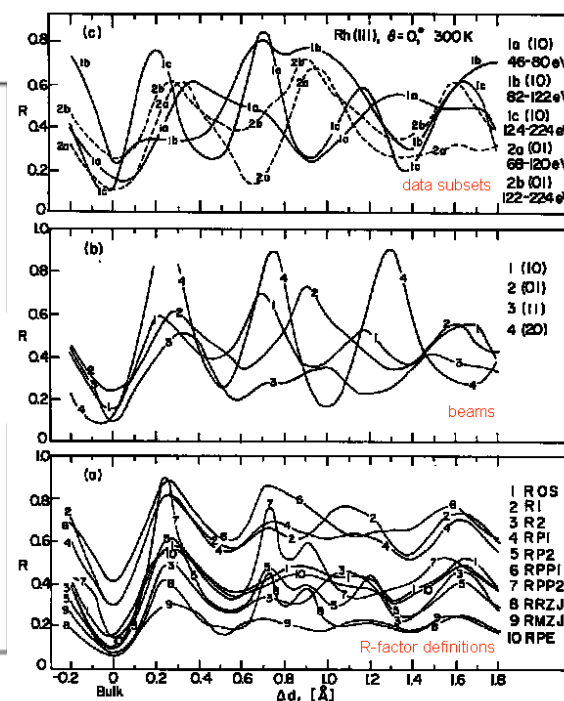
- 1) (-1.33, -0.08, 2.51)
- 2) (0.33, 0.00, 0.00)
- 3) (1.89, 1.22, 3.51)



I-V spectra



R-Factors



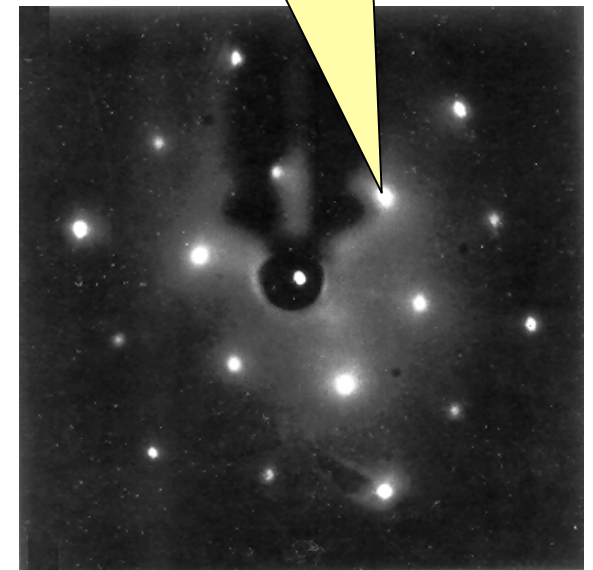
Pendry R-factor

$$R = \frac{\sum_g \int (Y_{gth} - Y_{gexp})^2 dE}{\sum_g \int (Y_{gth}^2 + Y_{gexp}^2) dE},$$

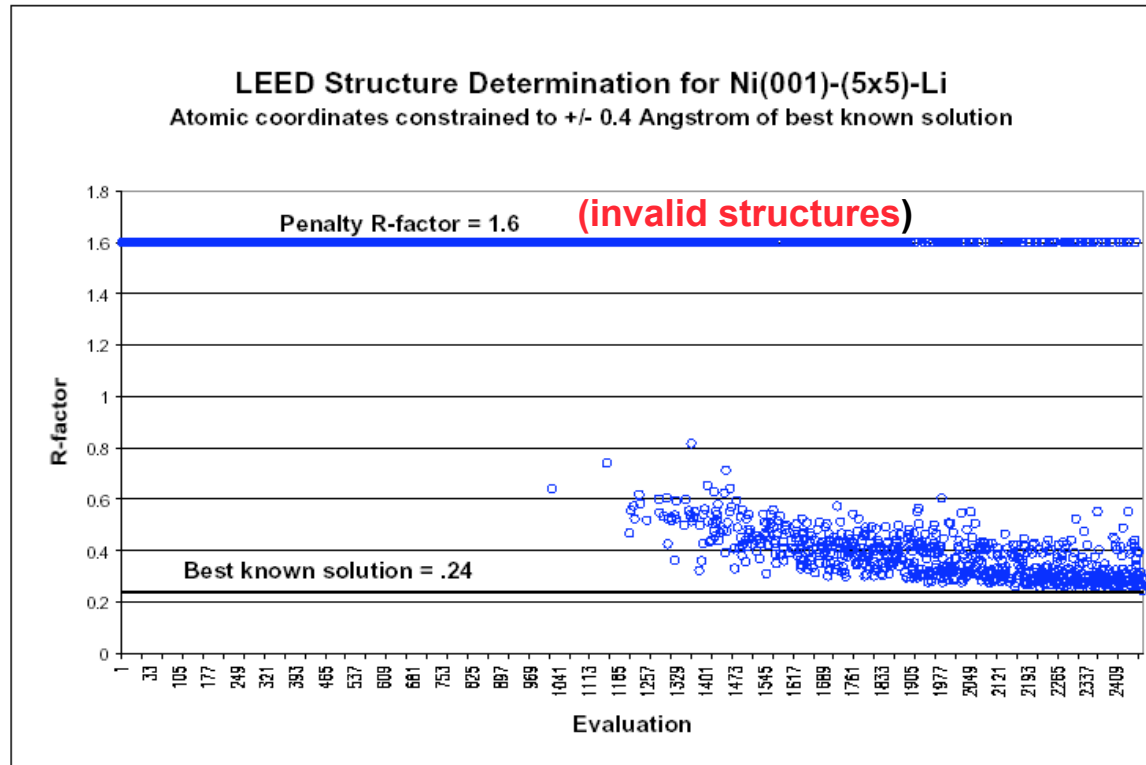
$$Y(E) = L^{-1} / (L^{-2} + V_{oi}^2),$$

$$L(E) = I'(x, y, z) / I(x, y, z)$$

I = Intensity



Previous Work

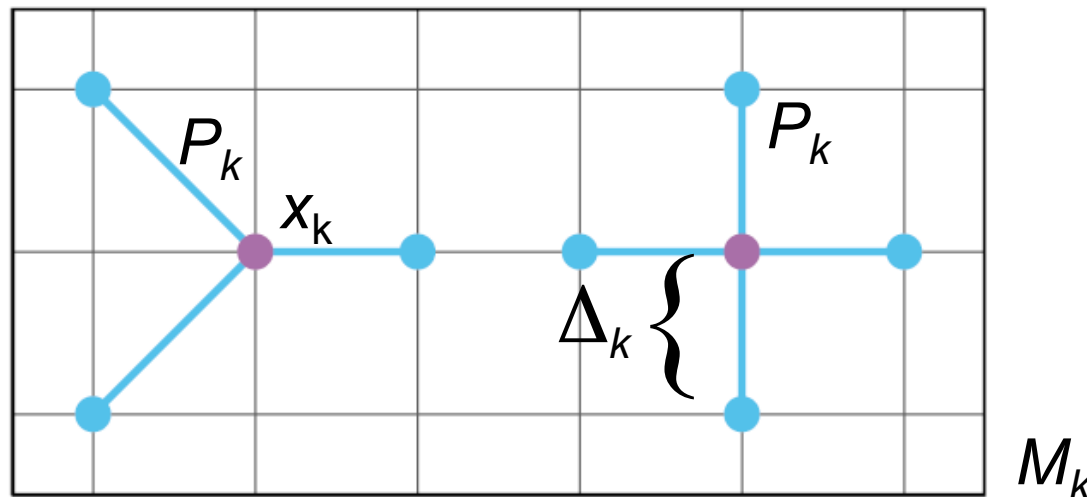


- ❖ Previous work used genetic algorithms to solve the optimization method.
- ❖ Large number of invalid structures generated (more on this later).
- ❖ Overall, a solution was found - *after adding sufficient constraints*.

1. *Global Optimization in LEED Structure Determination Using Genetic Algorithms*, **R. Döll and M.A. Van Hove**, Surf. Sci. **355**, L393-8 (1996).
2. *A Scalable Genetic Algorithm Package for Global Optimization Problems with Expensive Objective Functions*, **G. S. Stone**, M.S. dissertation, Computer Science Dept., San Francisco State University, 1998.

Brief overview of pattern search methods

- ❖ Pattern search methods, Torczon, Lewis & Torczon, Lewis, Kolda, Torczon (2004), etc.
- ❖ Extension to mixed variable problems by Audet and Dennis (2000).
- ❖ Case of nonlinear constraints studied in Abramson's PhD dissertation (2002).
- ❖ Good convergence properties
- ❖ Good software available - APPSPACK (Kolda), OPT++ (Hough, Meza, Williams), NOMADm (Abramson)



Generalized Pattern Search Framework

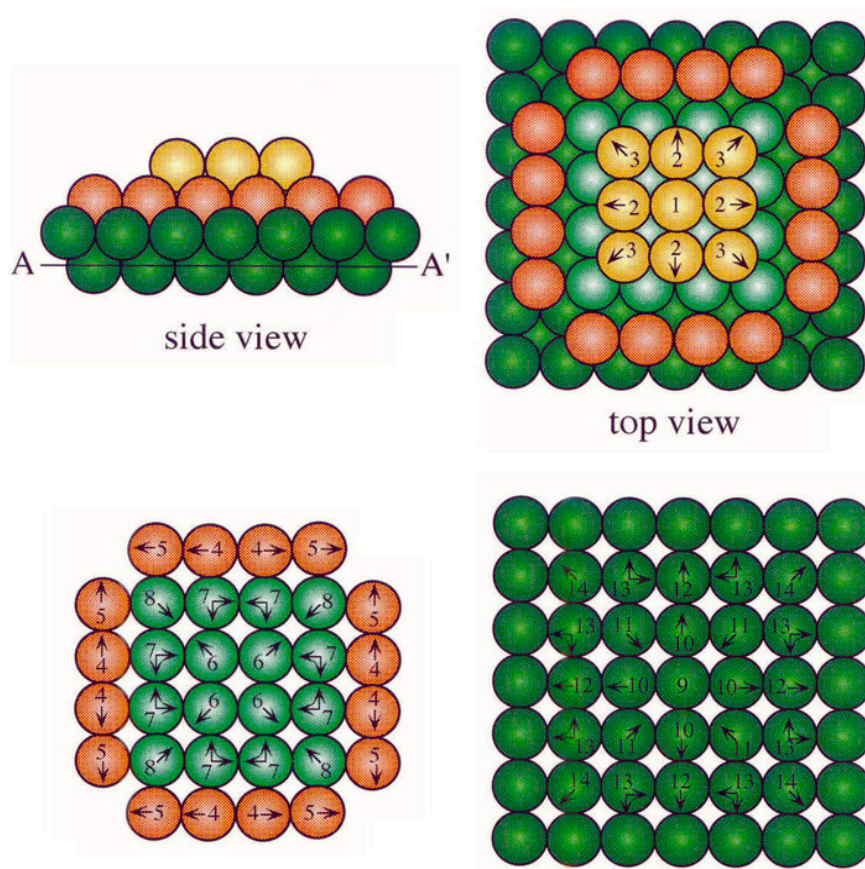
1. Initialization: Given Δ_0 , x_0 , M_0 , P_0
 2. For $k = 0, 1, \dots$
 - a) SEARCH: Evaluate f on a finite subset of trial points on the mesh M_k
 - b) POLL: Evaluate f on the frame P_k
 3. If successful - mesh expansion:
 - a) $x_{k+1} = x_k + \Delta_k d_k$
 4. Otherwise contract mesh
- Global phase can include user heuristics or surrogate functions
- Local phase more rigid, but necessary to ensure convergence

NOMADm

- ❖ Variables can be continuous, discrete, or categorical
- ❖ General constraints (bound, linear, nonlinear)
 - Nonlinear constraints can be handled by either filter method or MADS-based approach for constructing poll directions
- ❖ Objective and constraint functions can be discontinuous, extended-value, or nonsmooth.
- ❖ Available at:
<http://en.afil.edu/ENC/Faculty/MAbramson/NOMADm.html>

Test problem

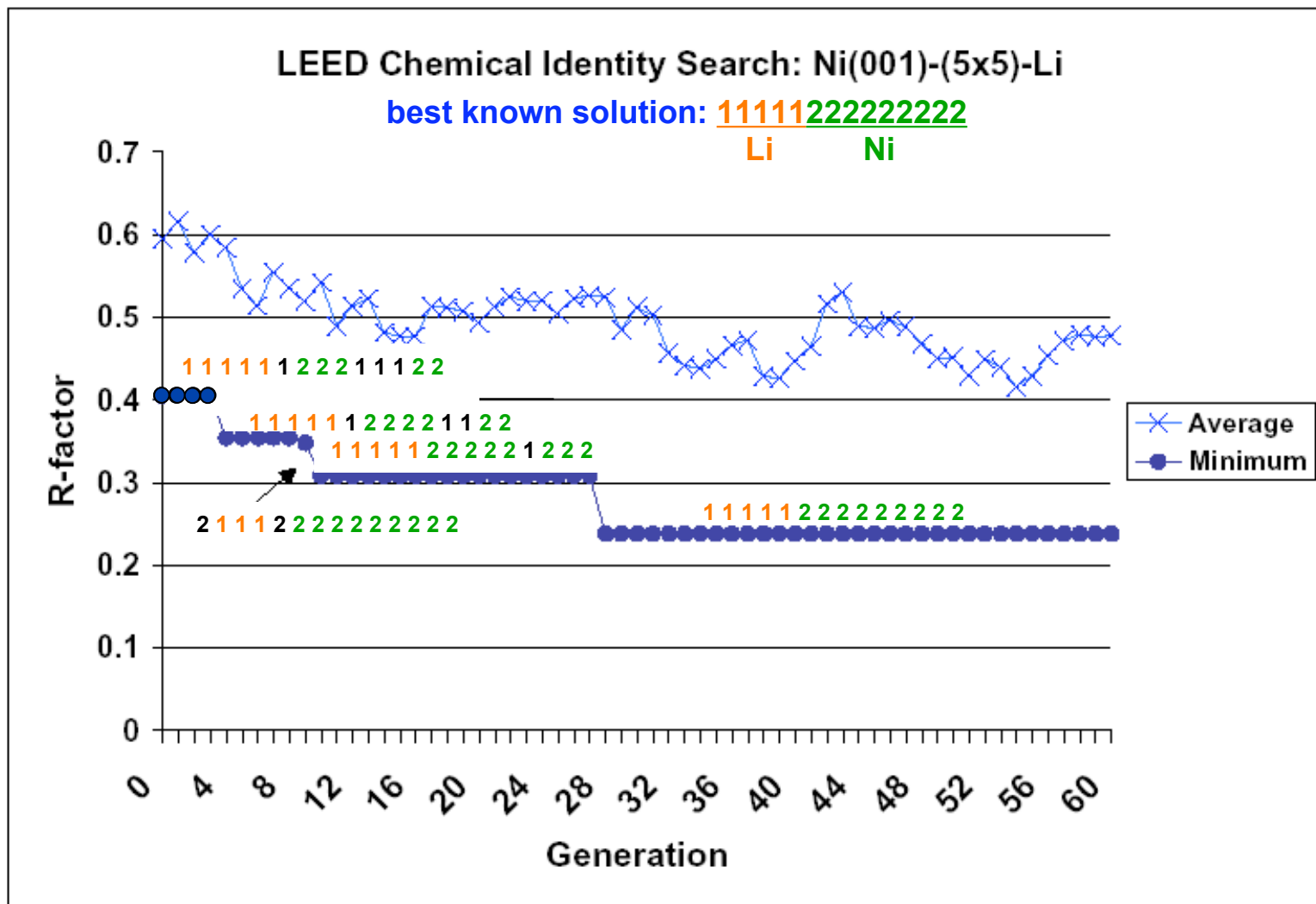
Ni(100)-(5x5)-Li



- ❖ Model contains three layers of atoms
- ❖ Using symmetry considerations we can reduce the problem to 14 atoms
 - 14 categorical variables
 - 42 continuous variables
- ❖ Positions of atoms constrained to lie within a box
- ❖ Best known previous solution had R-factor = .24

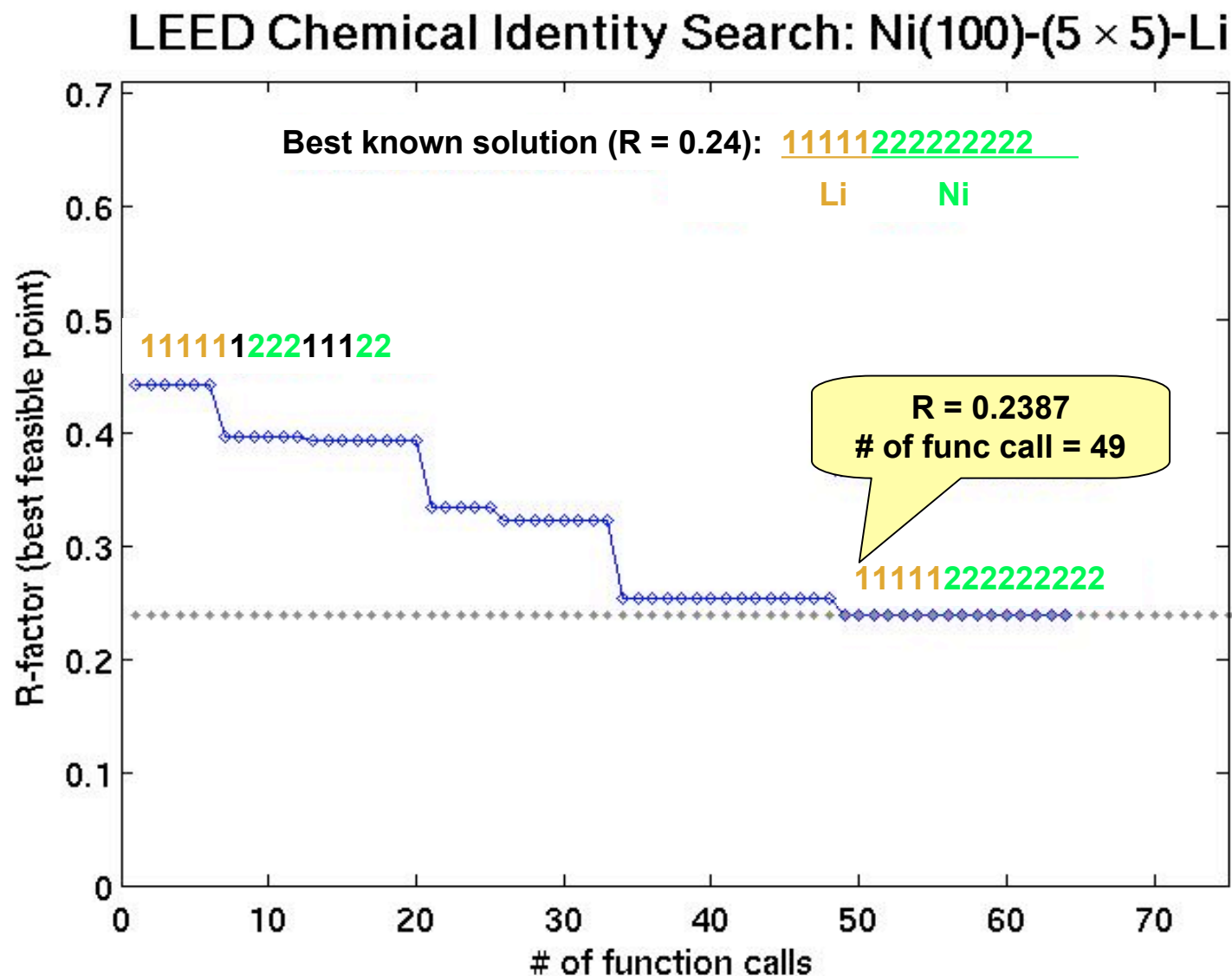
Model 31 from set of TLEED model problems

GA results - categorical variable search with fixed atomic positions

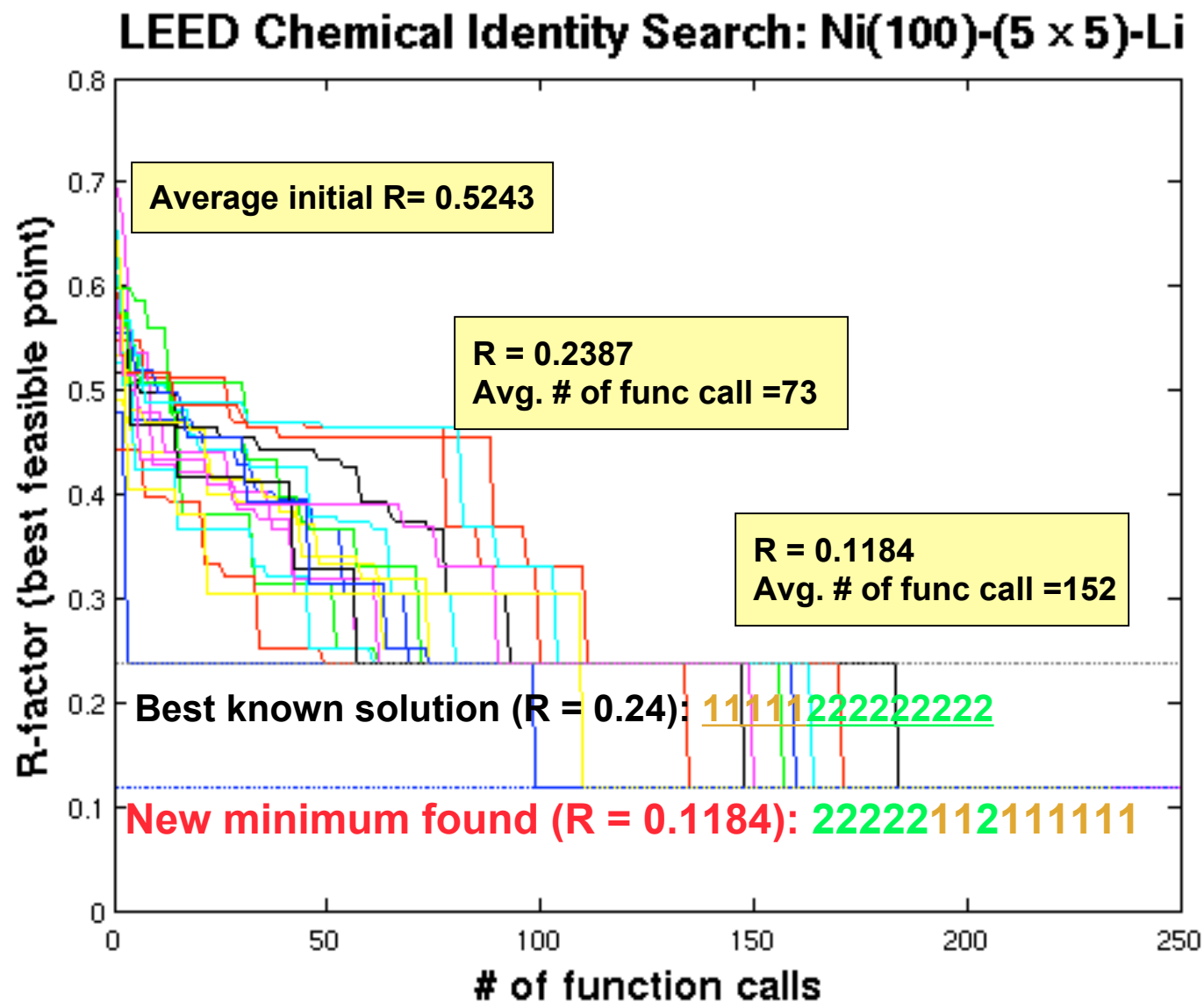


Remark: population size = 10 / Generation

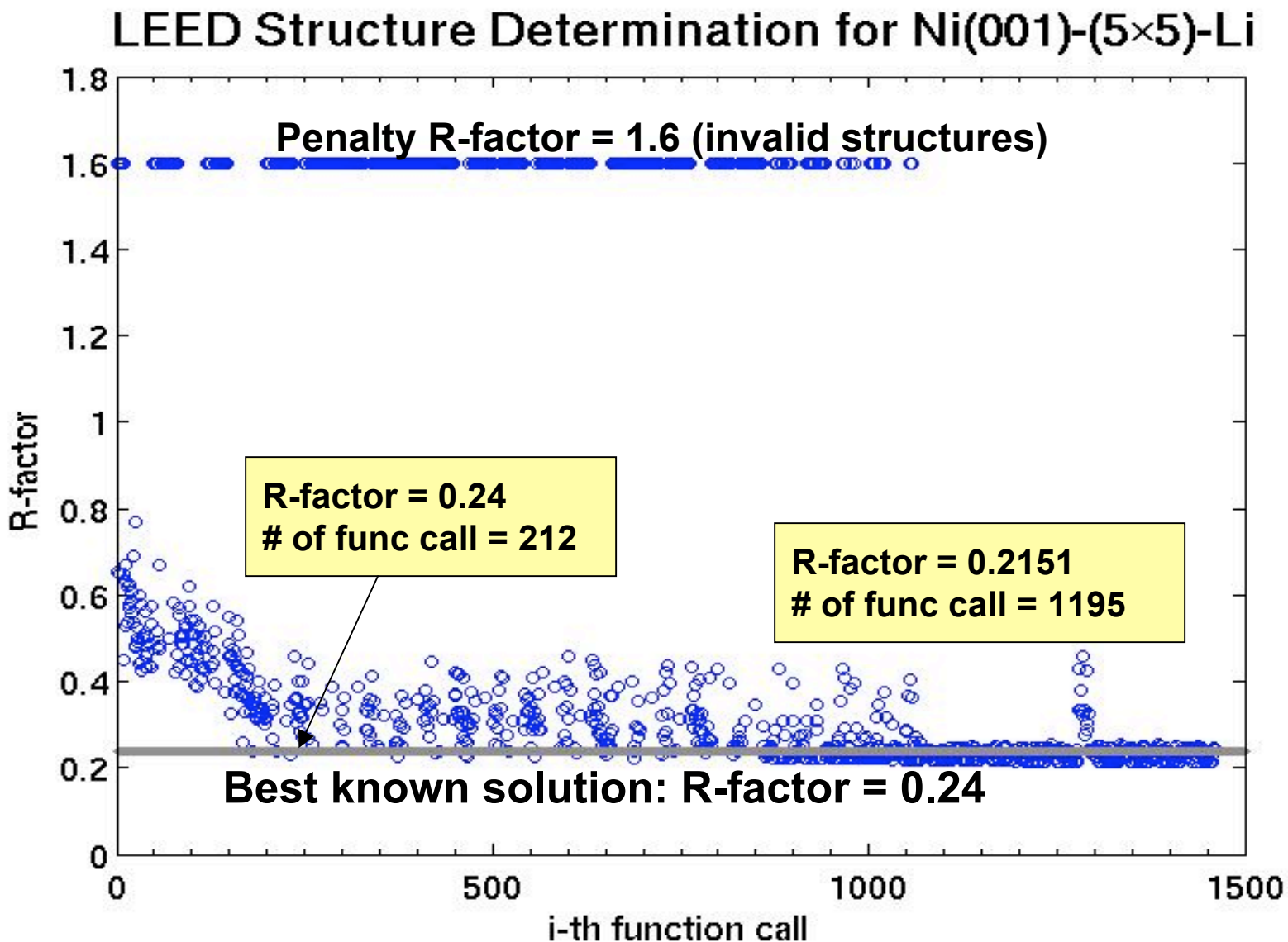
NOMAD results for categorical variables with fixed atomic positions



NOMAD results for 20 trials using LHS + GSS



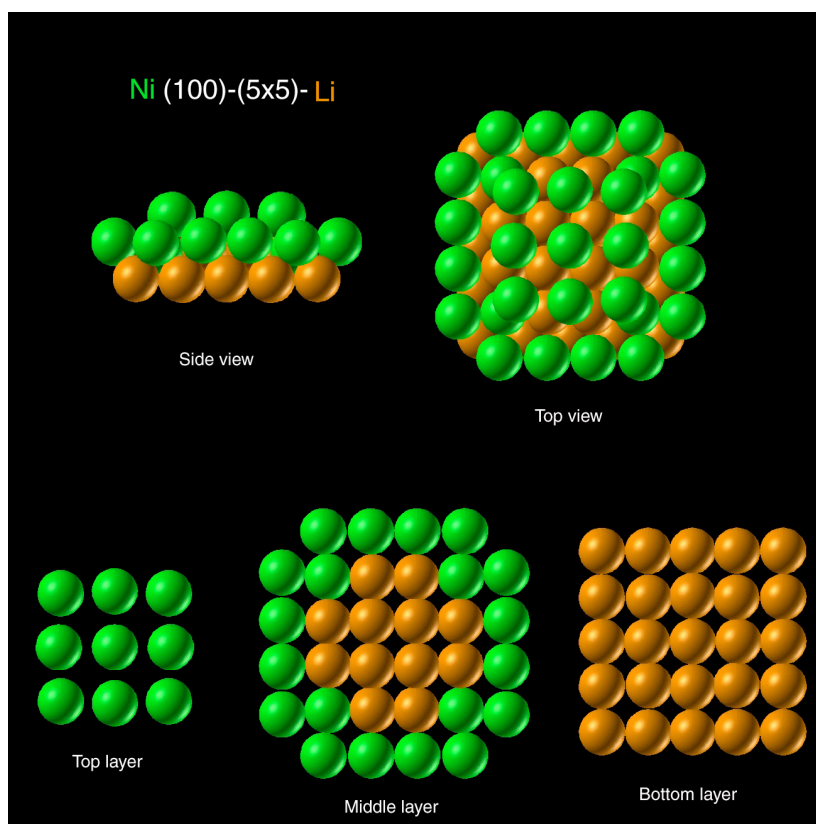
Minimization with respect to both types of variables removes coordinate constraints



LEED Chemical Identity Search: Ni (100)-(5x5)-Li

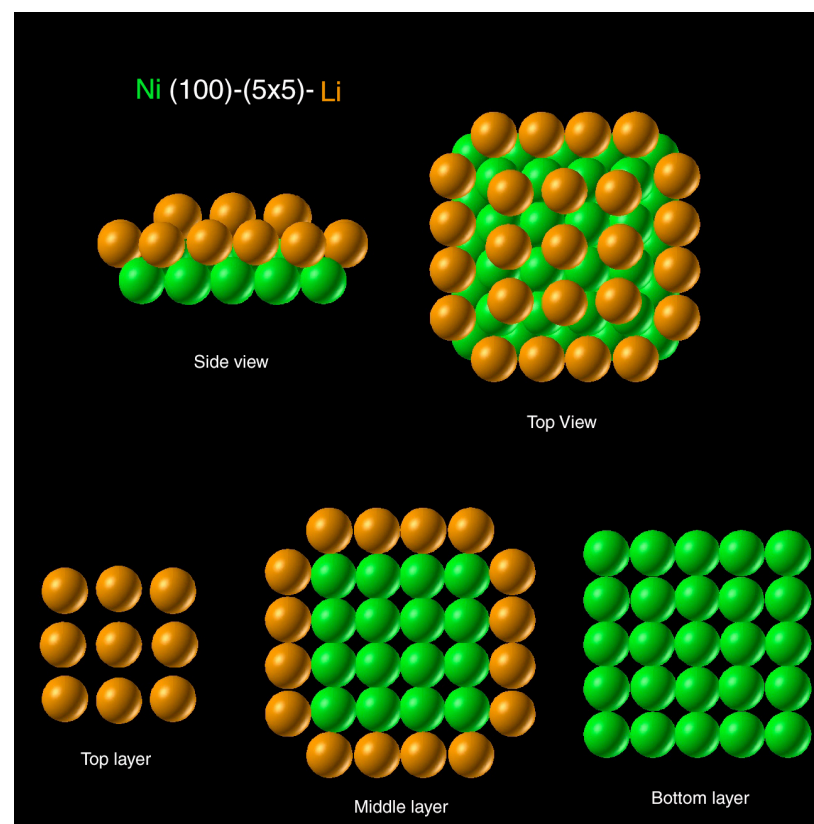
New structure found

$R = 0.1184$



Previous best known solution

$R = 0.24$



Conclusions

- ❖ Generalized pattern search methods for mixed variable problems were successful in solving the surface structure determination problem
 - On average NOMAD took 60 function evaluations versus 280 for previous solution (GA)
 - Improved solutions from previous best known solutions found in all cases
 - Generation of far fewer invalid structures
- ❖ Algorithm appears to be fairly robust, with a better structure found in all 20 trial points
- ❖ Ability to minimize with respect to both categorical and continuous variables a critical advantage for these types of problems

Acknowledgements

- ❖ Chao Yang
- ❖ Lin-Wang Wang
- ❖ Xavier Cartoxa
- ❖ Andrew Canning
- ❖ Byoungnak Lee

Questions