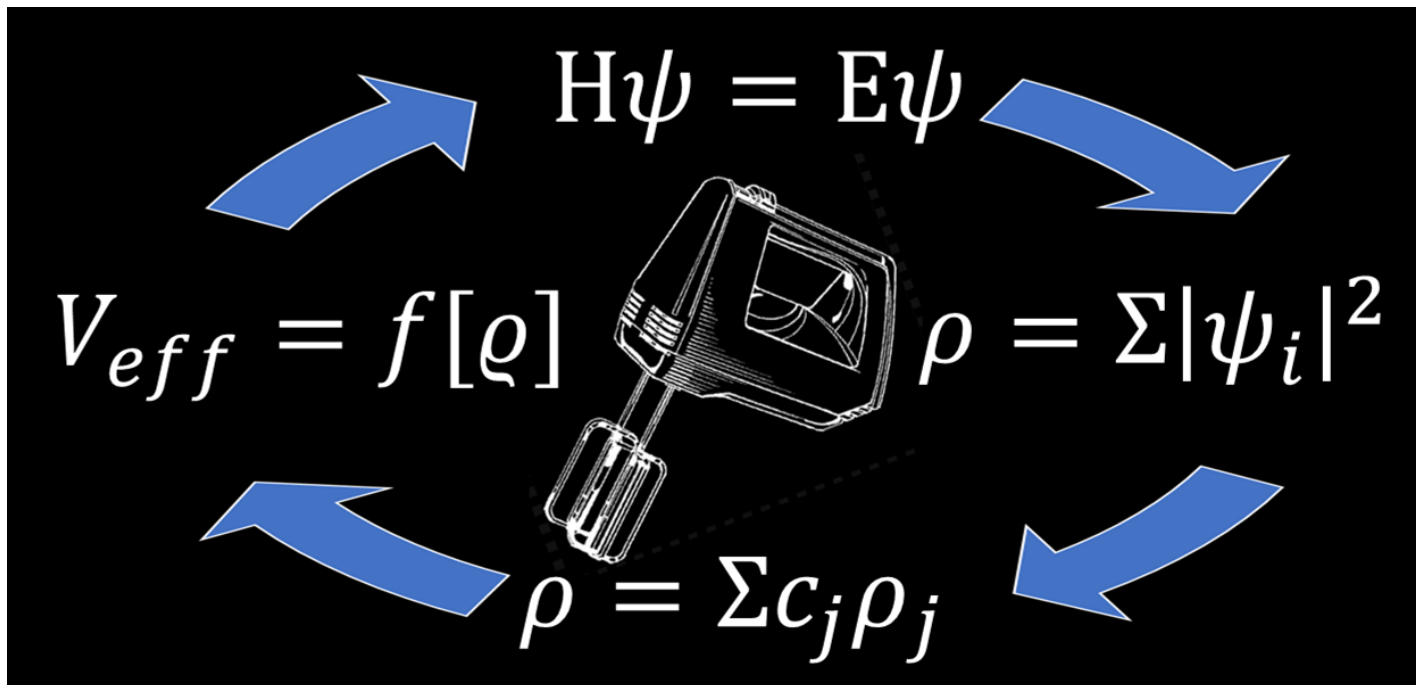


# *Mixing for Dummies*

L. D. Marks

Department of Mat Sci & Eng

Northwestern University



# Acknowledgements

**Peter Blaha**  
**TU Wien**




**Russel Luke**  
**U. Göttingen**



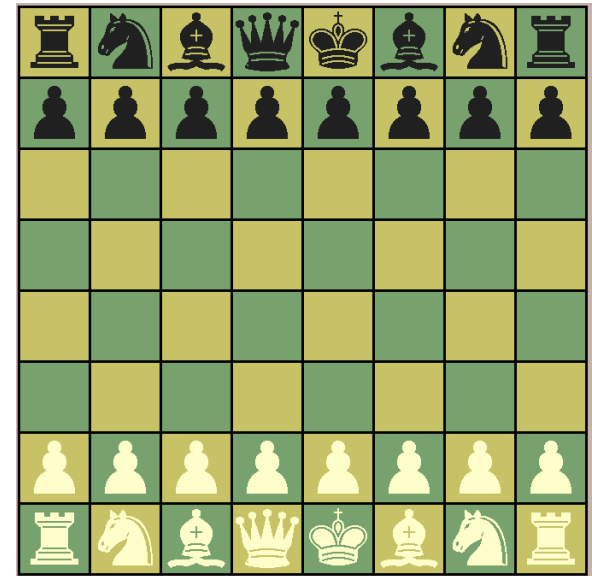
National Science Foundation  
WHERE DISCOVERIES BEGIN

# Overview

- 
- Role of mixing and force minimization
  - Nuts and bolts in Wien2k
  - Underlying math
    - Simple form
    - What is really going on
    - Predictive mixing
  - What to do when you mess up

# Overview

- Role of mixing and force minimization
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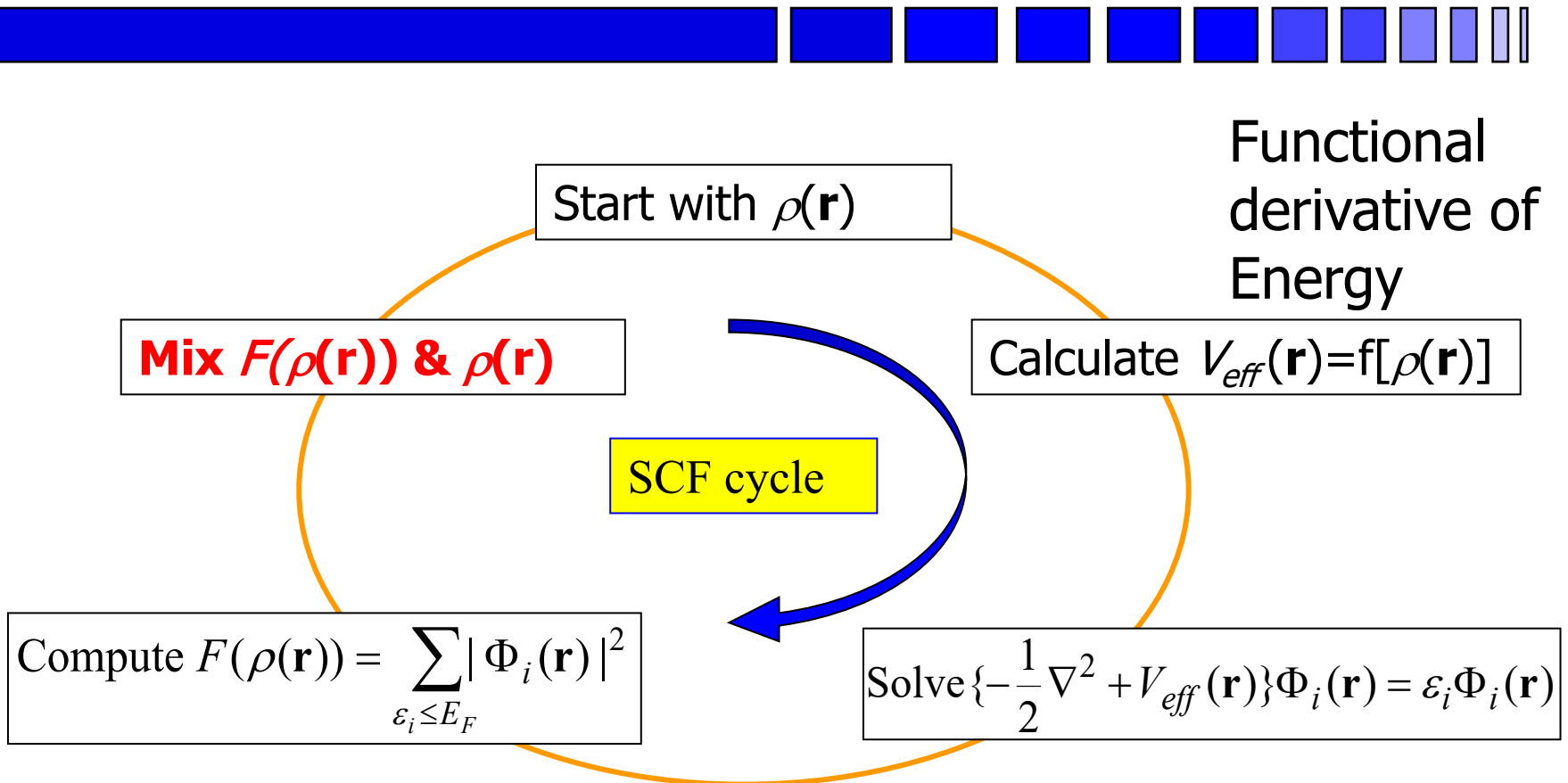


# What is DFT?

## □ Density Functional Theory


- Solve the quantum mechanics for tens to thousands of atoms
- Fundamental to many areas of current chemistry, materials science and physics
- One of the major killer of electrons on supercomputers

# Fixed Point cycle




*Solves a variational energy by a fixed-point method, implicit minimization*

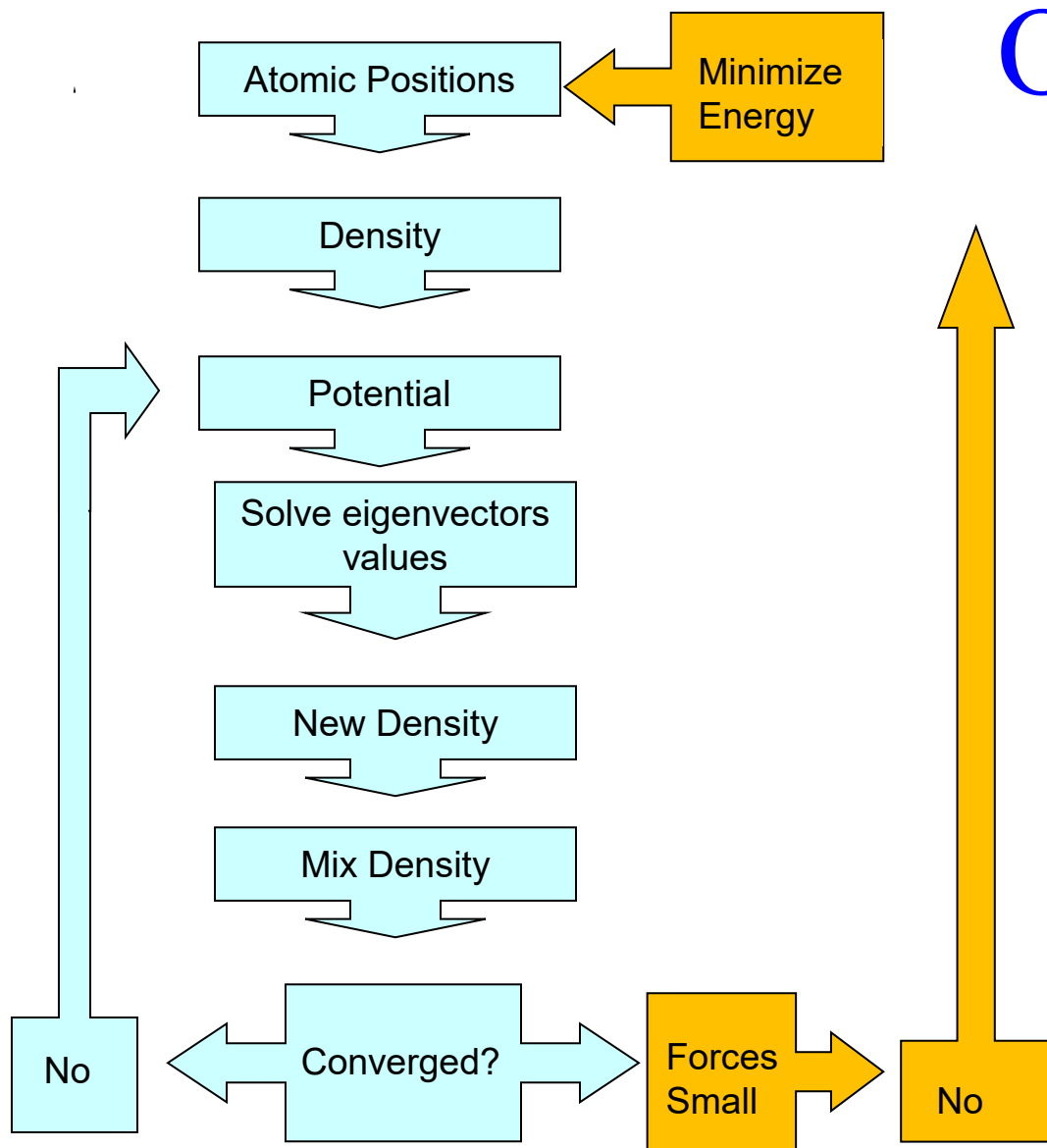
# General Structure

- 
- Density contains  $10^3$  to  $10^6$  components
  - No gradients plausible
  - Fortunately the eigenvectors/values are much, much smaller (10-1000)
  - Often converges in 20-40 iterations
  - Iff well posed, the stationary point is a variational minimum of the energy, so has good properties

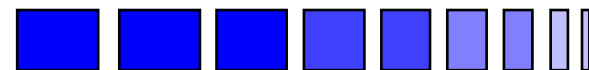
# What has to be computed

- 
1. The atomic positions need to be minimized
  2. A self-consistent density has to be found
    - Conventional is to do nested loops, 2 inside 1
    - An alternative is to solve the two at the same time (MSR1a)



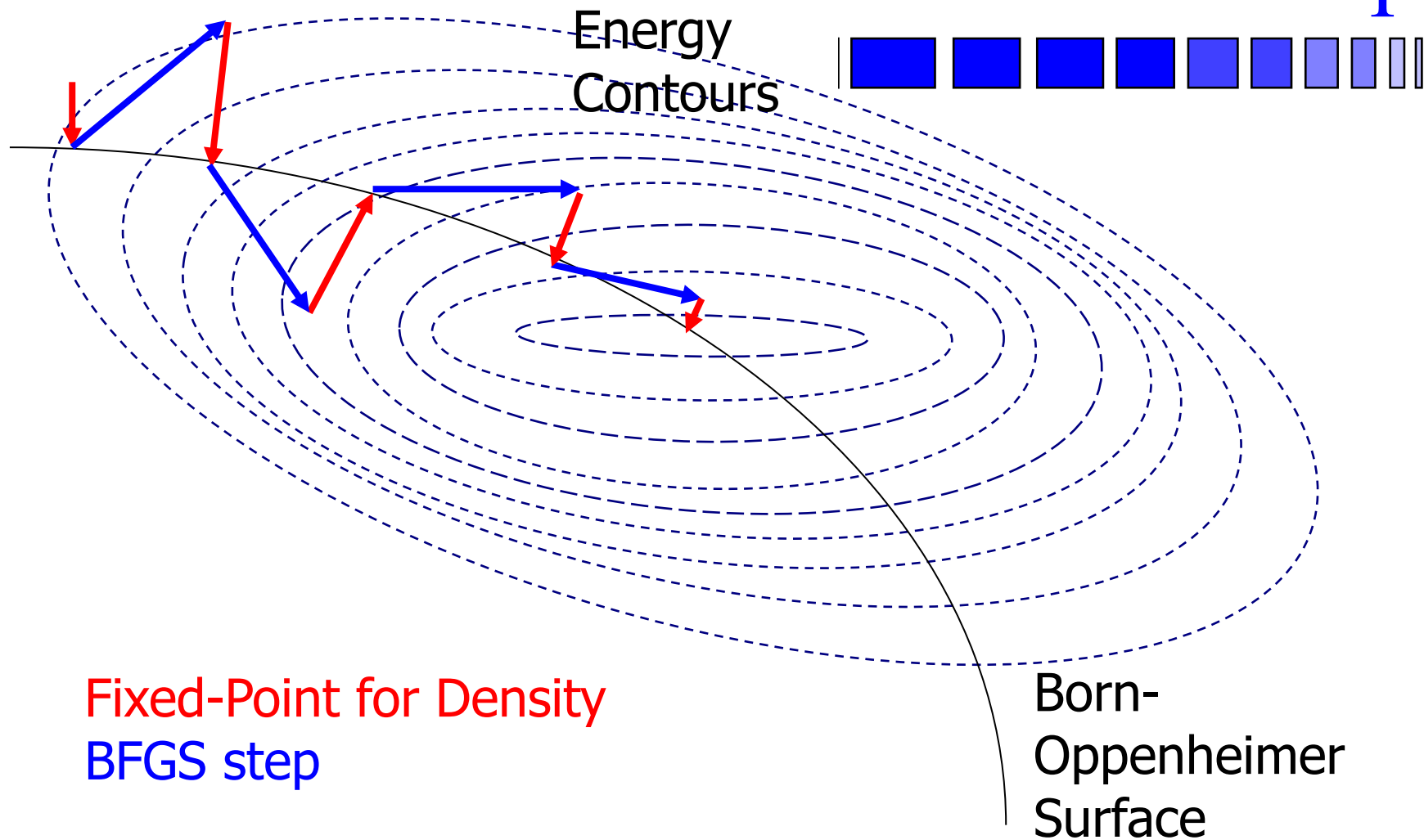


# Old (PORT)



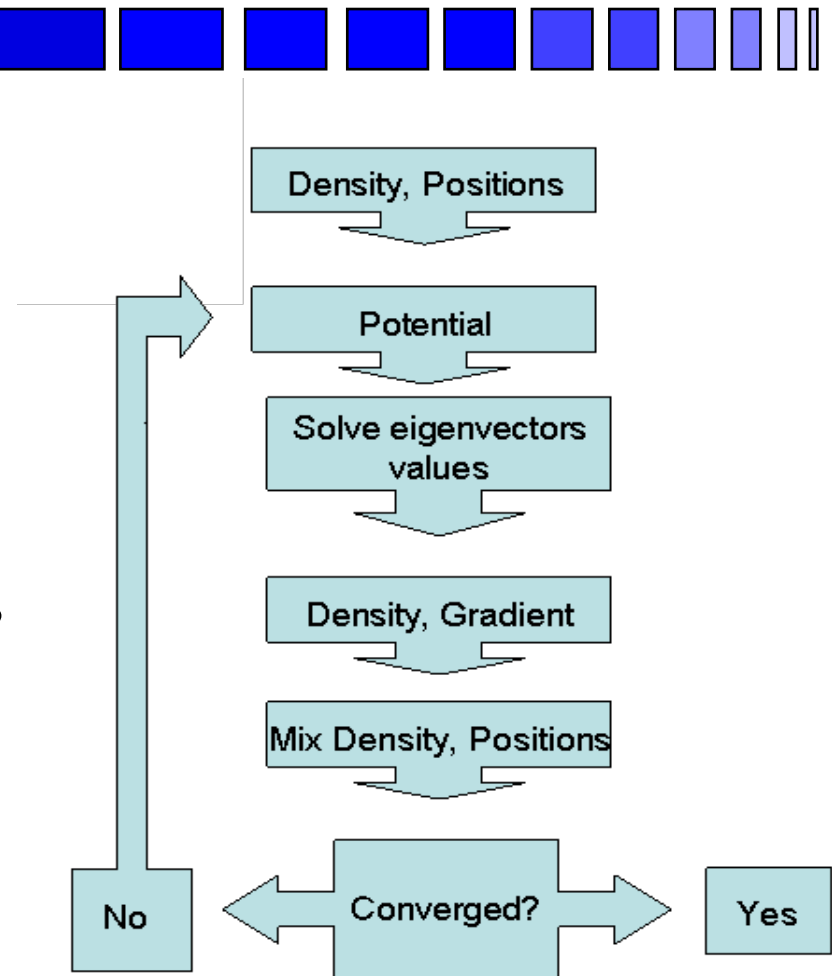
- Inner loop to obtain fixed-point for given atom positions
- Outer loop to optimize atomic positions

# Double-Loop



# MSR1a Fused Loop

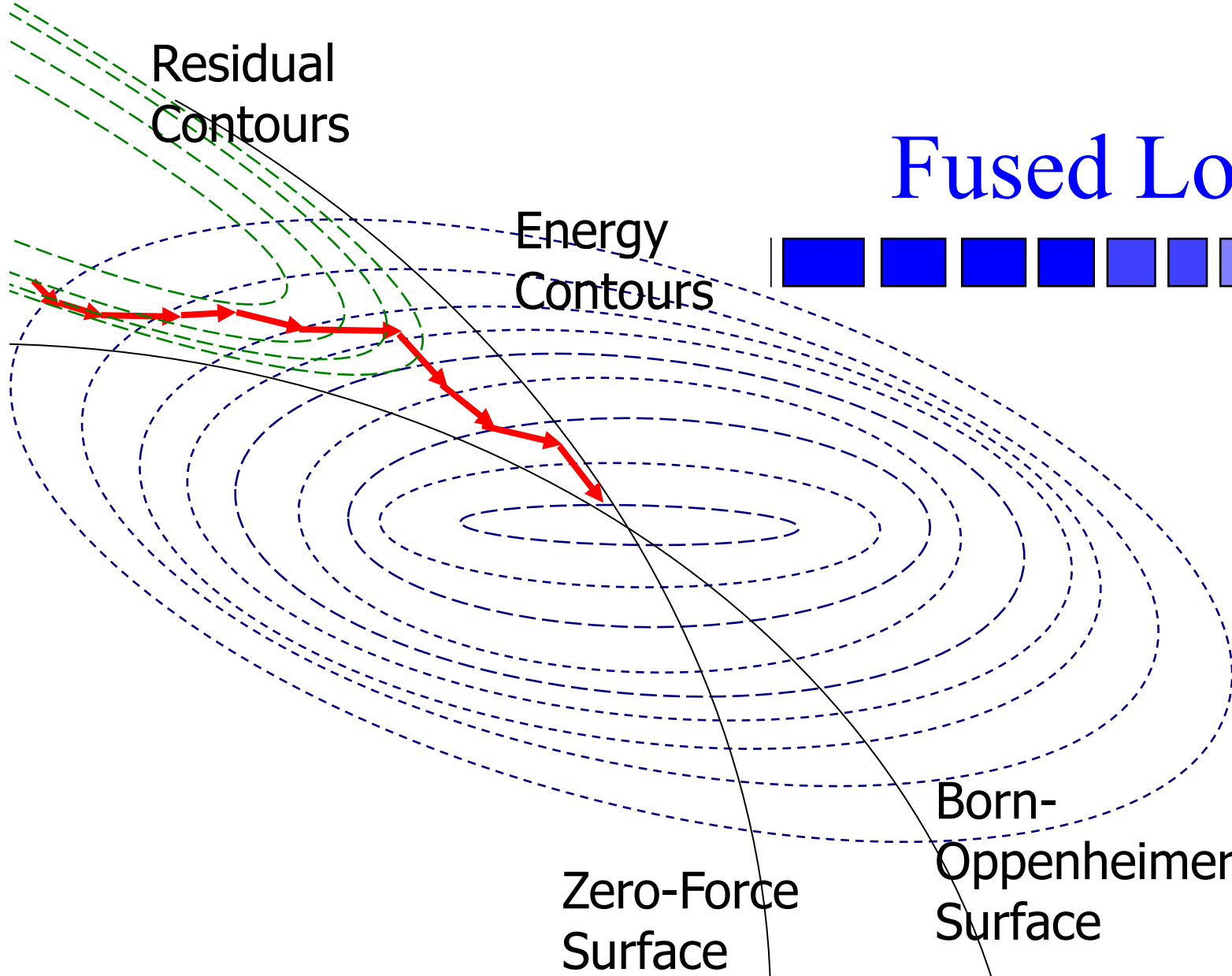
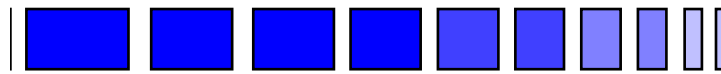
- Treat the density and atomic positions (as well as hybrid potentials etc as needed) *all* at the same time.
- No restrictions to “special” cases, general algorithm has to work for insulators, metals, semiconductors, surfaces, defects, hybrids....
- Few to no user adjustable parameters



Residual  
Contours

Energy  
Contours

Fused Loop

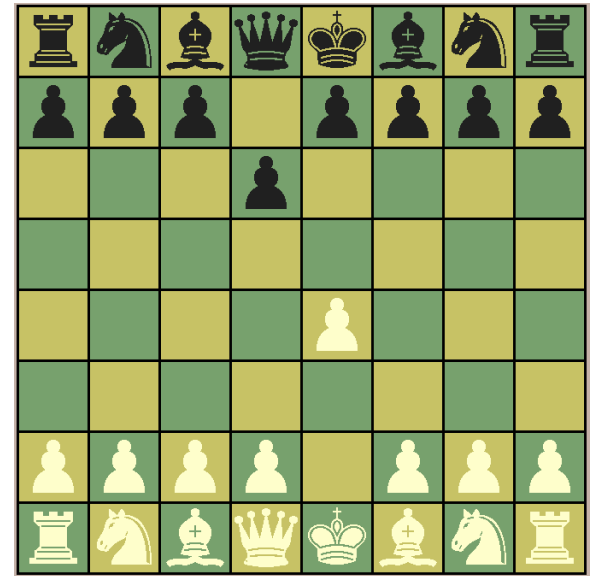


Zero-Force  
Surface


Born-  
Oppenheimer  
Surface

# Overview


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# Calculating a system

- 
- Choose the functional (LDA, PBE, WC)
  - Find the DFT equilibrium volume for the bulk, with similar RMT, RKMAX (may have to redo later)
  - Create the structure – supercell or other (e.g. Cryscon from Shape Software)
  - Look at it – you probably did something wrong!

# Converge density

- 
- Done with the “runXX\_lapw” scripts
  - In 99% of cases no need to change mixing parameters
  - First iteration might be problematic, could use “echo .001 > .pratt”
  - Most issues are due to bad models (e.g. cold fusion)

# Converge positions

- Done with `runXX_lapw` and “-min” option
- Again, in 99% of cases no need to change anything
- Sometimes it is better to slightly converge the density first
- Can converge better than density alone, but takes longer
- Be careful about iterative modes

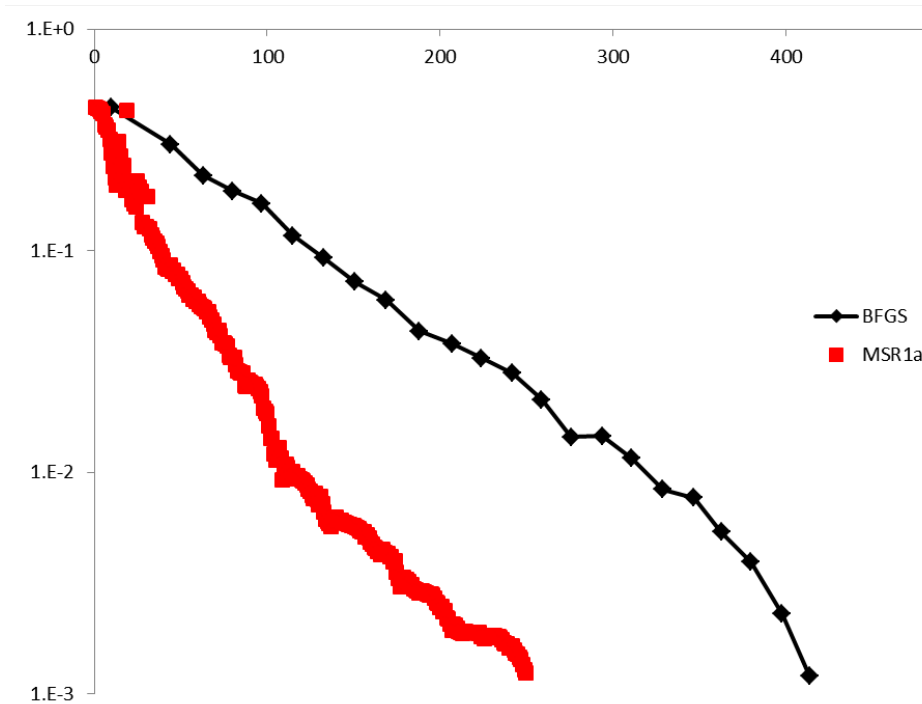


# Alternative for positions

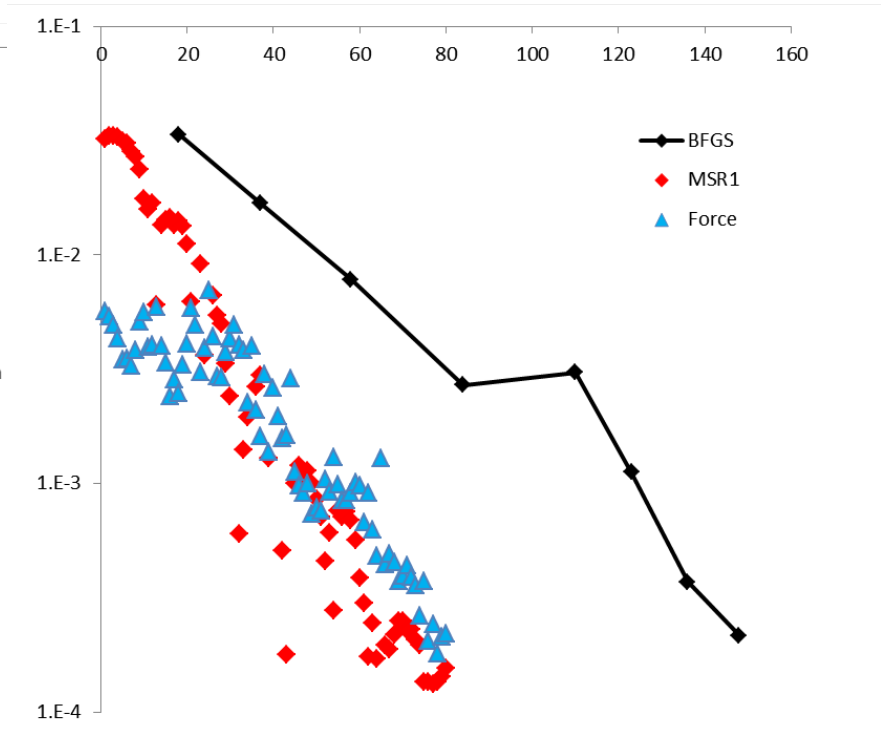
- Slower double loop (PORT), using `min_lapw`
- Uses gradients, and builds an approximation to the Hessian (2<sup>nd</sup> derivative matrix)
- If you are doing many almost identical calculations it can be faster due to the Hessian from prior calculations

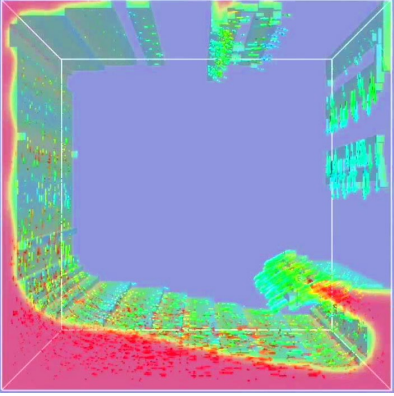
# Comparison

52 atoms, MgO (111)+H<sub>2</sub>O



108 atoms AlFe



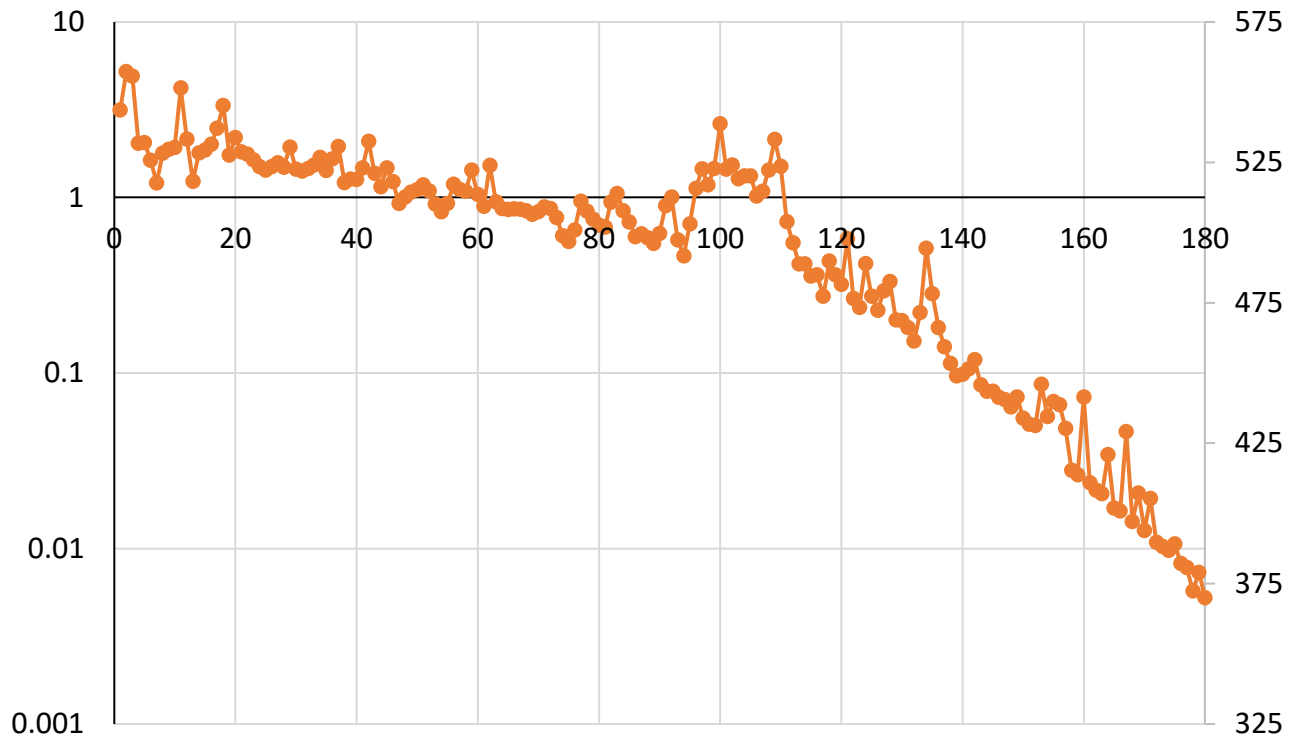


# Sloshing



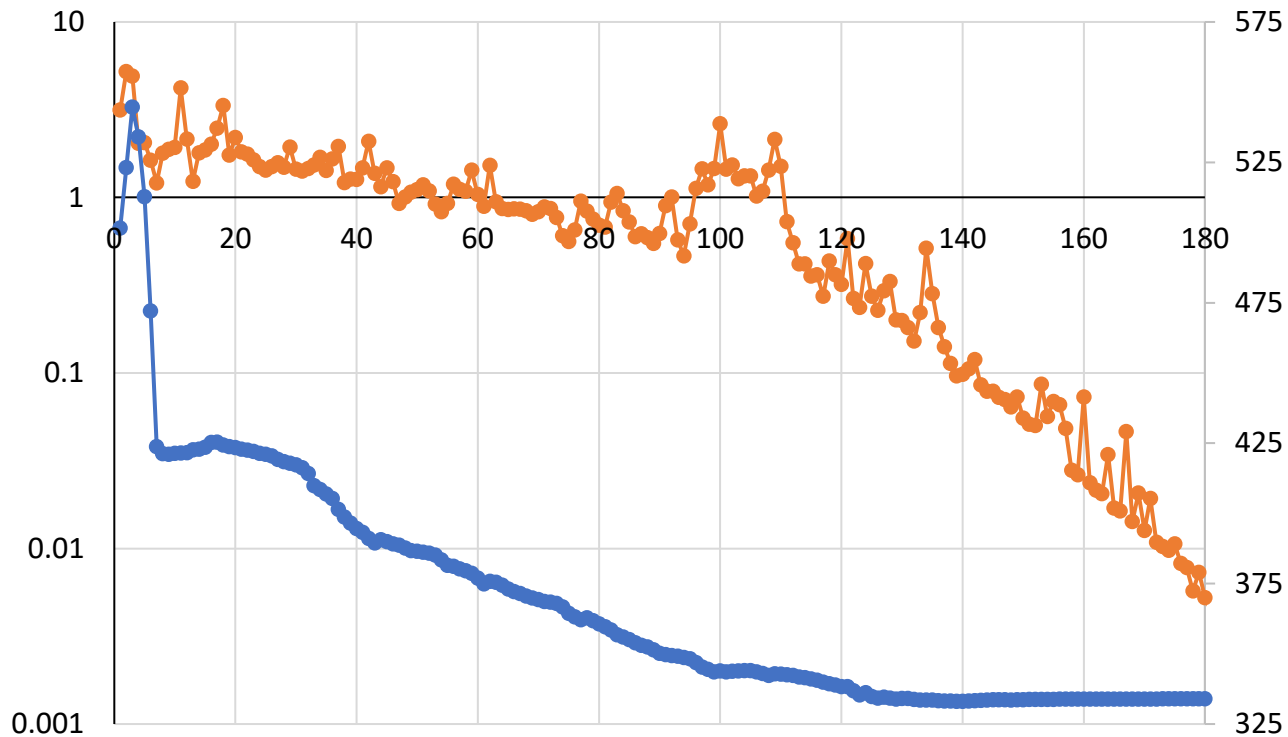
- Oscillations of the charge density
- Density moving from one end of a surface to another
- Appears as oscillations
- Due to too large steps (non-linear terms)
- Or....

# Not really oscillations



Density apparently oscillating

# Not really oscillations

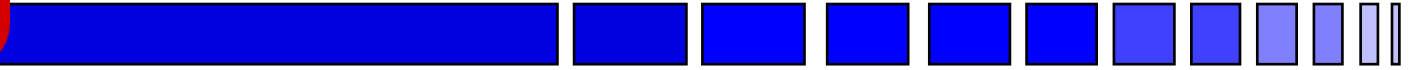


Density apparently oscillating

Total PW density in sphere converging



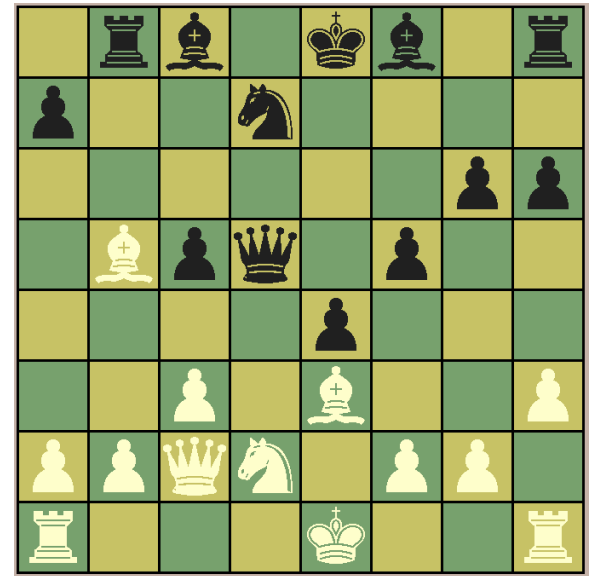
# Beware of Fake News



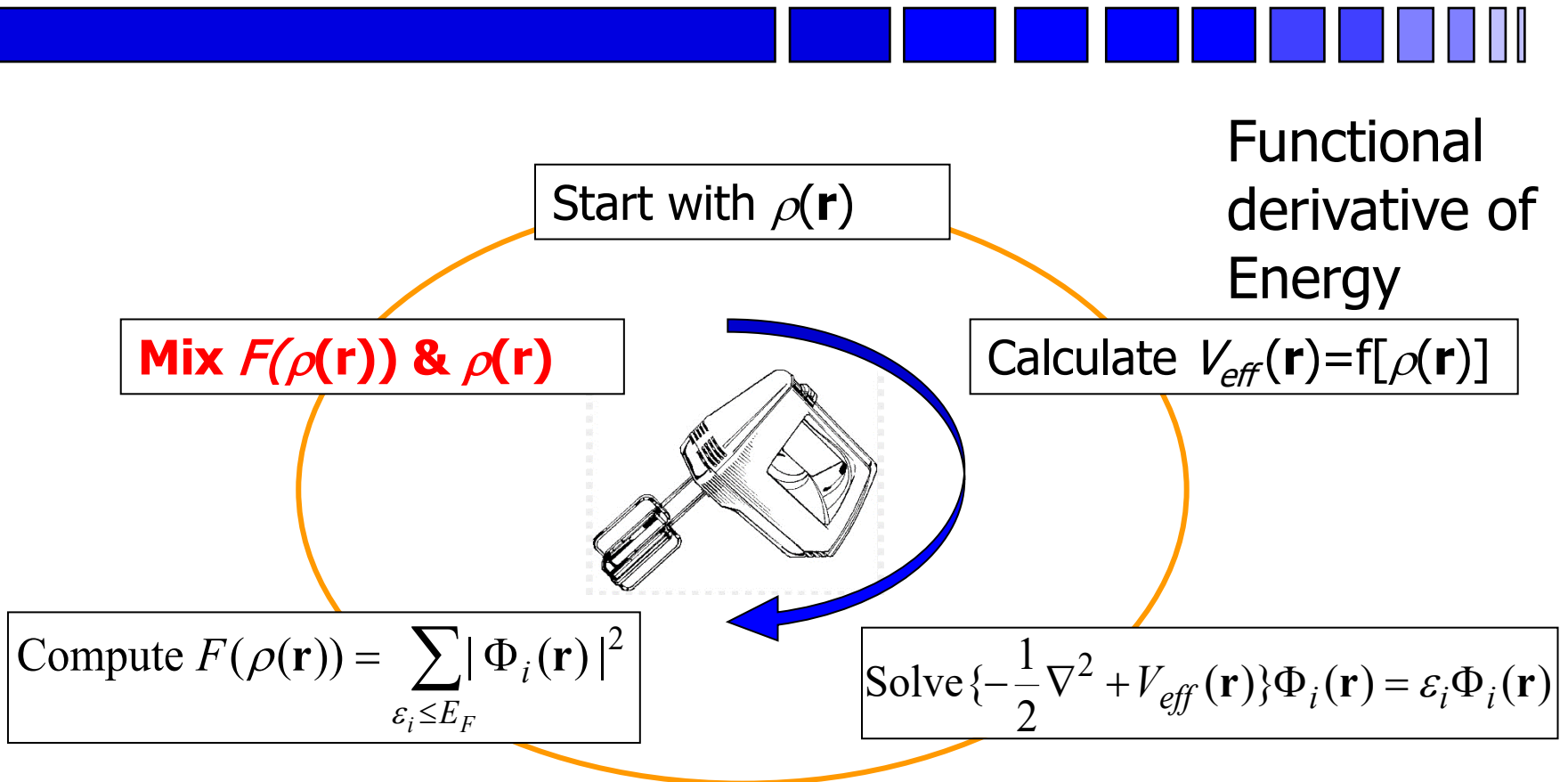
- “Reduce the mixing factor” wisdom is for other codes – don’t
- Fixing atoms probably wont make the calculation converge faster
- Bigger problems are worse
  - Density convergence scales as the number of bands and their width
  - Positions scale as the number of electron-phonon bands (dielectric eigenvalues)

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# Fixed Point cycle

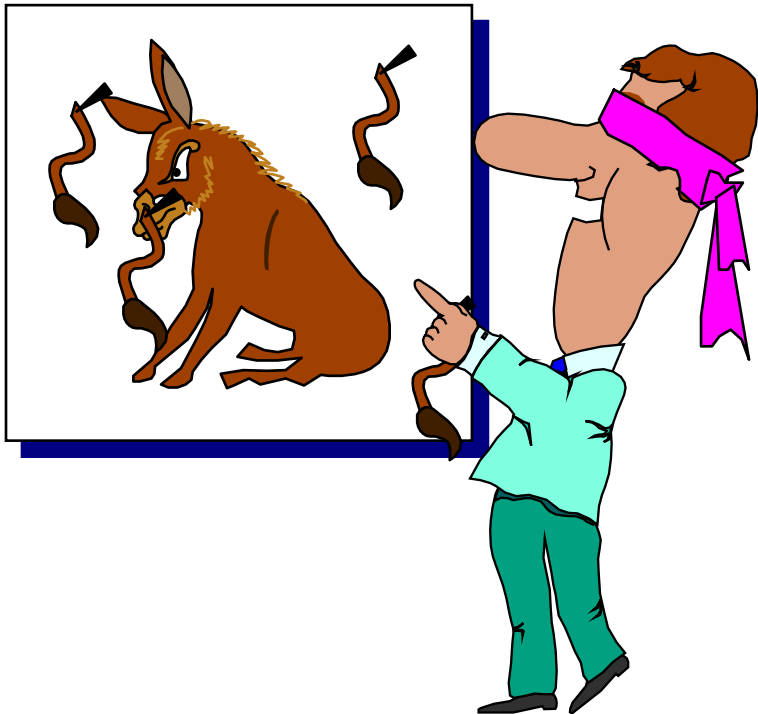


*Solves a variational energy by a fixed-point method, implicit minimization*

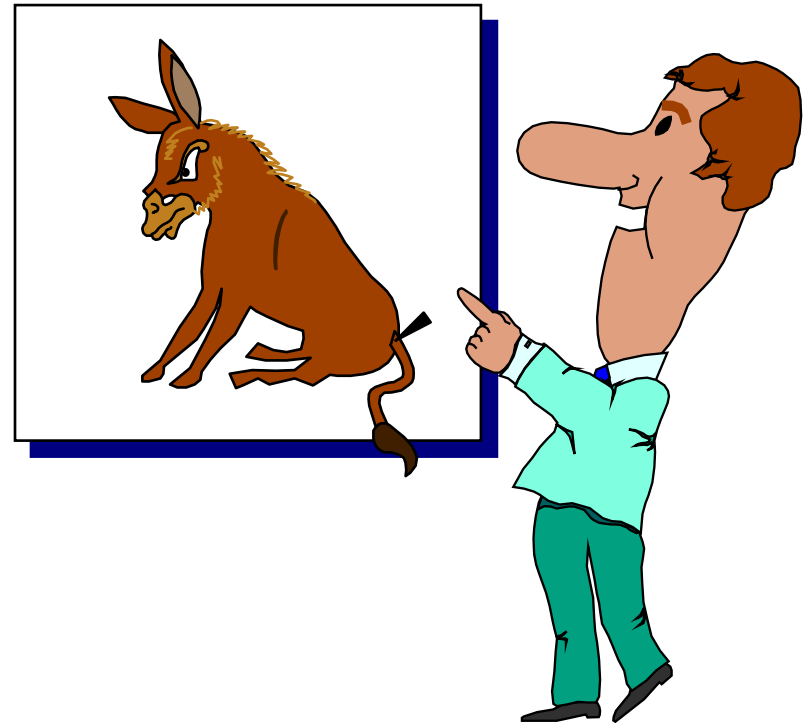


# Underlying principle


Current density



Use prior densities




# Information available

- 
- History of prior densities, and the density at the end of the SCF iteration
  - History of atomic positions, and the pseudo-forces
  - Use these to build an approximate mathematical model

# Historical perspective

- New point  $F(\rho_k)$
- Define a residue  $R_k = F(\rho_k) - \rho_k$
- Simplest is linear
$$\rho_{k+1} = \rho_k + \alpha R_k$$
- Slow, and what is  $\alpha$  ? (user choice)
- Pratt method often works if the user chooses right

# General form

- 
- $s_i = \rho_i - \rho_k ; y_i = R_i - R_k$
  - $R = R_k + B_k s_{k+1} + \cancel{s_{k+1} C_k s_{k+1}} + \dots$
  - Form a matrix for s & y values
  - The key question is how to construct the solution, for which there are more than a few issues.
  - Require that the reproduce prior steps

# Broyden Fixed-Point Methods



- Broyden's "Good Method"


$$B_{k+1} = B_k + \frac{(y_k - B_k s_k) s_k^T}{s_k^T s_k}$$

- Broyden's "Bad Method" ( $H=B^{-1}$ )

$$H_{k+1} = H_k + \frac{(s_k - H_k y_k) y_k^T}{y_k^T y_k}$$

C.G. Broyden, A Class of Methods for Solving Nonlinear Simultaneous Equations, *Mathematics of Computation*, 19 (1965) 577-593.

# Why Good & Bad?

- 
- Originally the “Good” method worked for Broyden (PRB...) and the “Bad” method did not
  - For many years the “Bad” method was never used.....except in mixing as DFT developers probably never read the literature
  - Subtle difference
    - Good Broyden: finds optimal density with current model (greedy algorithm)
    - Bad Broyden: finds most conservative density to minimize residue (least greedy algorithm)

# DFT history

- Initially people tried the “standard” Broyden method (1965) – it did not work
- Sequential Broyden, fair (Srivastava, 1984)
- Better, multisequant (Johnson 1988)

$$\rho_{k+1} = \rho_k + \sum \alpha_j R_j$$

- The coefficients from L2 (LS) solution.
- *Some LS are more equal than others.*  
*Different regularization & conditioning*

# General form

- Unpredicted step, greed  $\alpha_k$

$$u_k = R_k - B_k \rho_k$$

- Predicted step

$$p_k = B_k^{-1} R_k = H_k R_k$$

- General form

$$\rho_{k+1} = \rho_k + \alpha_k u_k + \beta_k p_k$$



# General form

- Unpredicted step, greed  $\alpha_k$

$$u_k = R_k - B_k \rho_k$$

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- General form

$$\rho_{k+1} = \rho_k + \alpha_k u_k + \beta_k p_k$$

# Issue 1: Unpredicted step

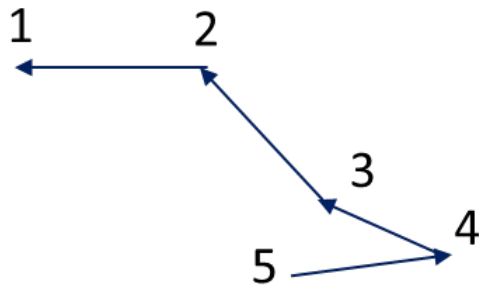
- No information available
- Must be controlled
- If too large, simplex gradients are unreliable
- Implicit trust, i.e. increase if improving “enough”
- It is hard to know what enough is



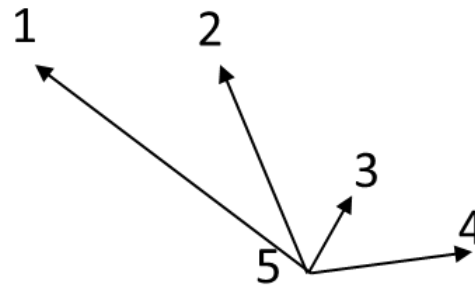


# Issue 2: Scaling

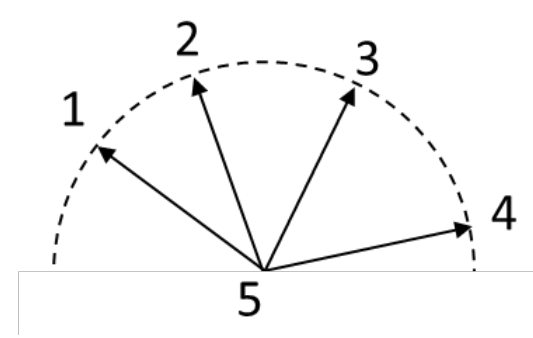
- How to treat the previous steps?
  - Sequential (Broyden) or multisequant (DIIS, Simplex)?
  - As is, with different magnitudes?
  - All my tests support a simplex gradient



Broyden



DIIS



Simplex

# General form

- Unpredicted step, greed  $\alpha_k$

$$u_k = R_k - B_k \rho_k$$


- Predicted step

$$p_k = B_k^{-1} R_k = H_k R_k$$

- General form

$$\rho_{k+1} = \rho_k + \alpha_k u_k + \beta_k p_k$$

# An issue for fixed-point problems

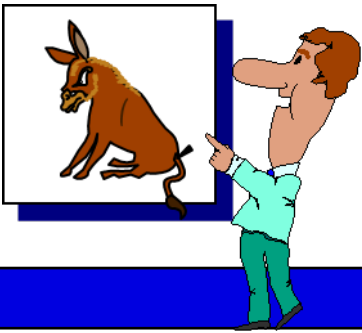
- 
- Most early methods are somewhat causal, appropriate for a minimization where the energy is decreasing
  - This does not need to be the case for a fixed-point problem as the gradient is not available, instead a residual
  - There is nothing a-priori to say that point 2 should be better than point 1
  - Hence....

# Multisecant Approach

- Consider a number of values:
- $S = (s_0, s_1, \dots, s_n)$  ;  $Y = (y_0, y_1, \dots, y_n)$
- Expand to a simultaneous solution:
- $BS = Y$  ; or  $HY = S$
- Minimum-Norm Solution (MSEC)

$$H_{k+1} = H_k + (S_k - H_k Y_k)(Y_k^T Y_k)^{-1} Y_k^T$$

Take  $H_k = I$



# What form?

- $H_k = S_k (T_k^T Y_k)^{-1} T_k^T$
- $T_k = (1 - \lambda)S_k + \lambda Y_k$
- $\lambda = 0$  is conventional “Good” Broyden, most greedy, may diverge, soft case
- $\lambda = 1 \cong$  DIIS, Anderson, “Bad” Broyden, least greedy, may stagnate (different scalings...), stiff case
- 8-16 memories (not critical)



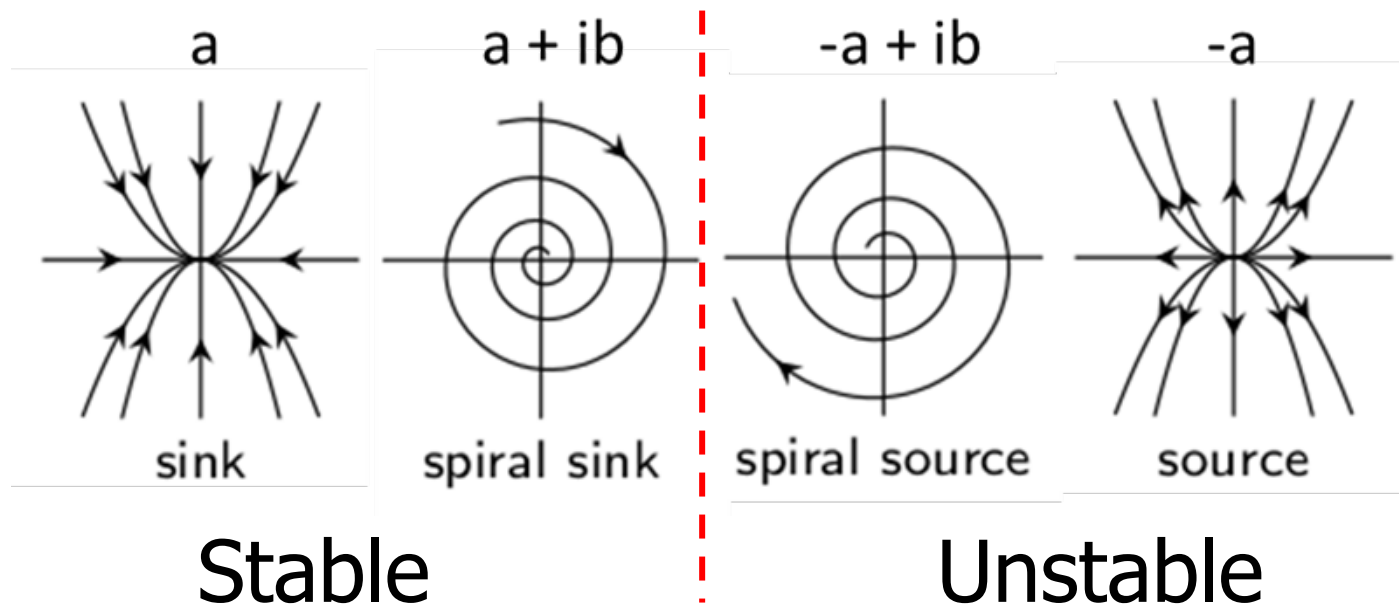
# What is a greedy algorithm?

- A greedy algorithm takes decisions on the basis of information at hand without worrying about the consequences. In many cases “greed is good”, but not always.
- Example: make 41c with 25c, 10c, 4c coins
- Optimum solution:  $25+4 \times 4$
- Greedy solution: start with 41c, use largest reduction
  - 25c                      Remainder 16
  - 10c                      Remainder 6
  - 4c                        Remainder 2




# Ansatz: inspired by optimization

- $T_k = \lambda S_k + Y_k$
- Search up for largest  $\lambda$  where  $T_k^T Y_k$  does not have source eigenvalues



# Issue 4: Predicted

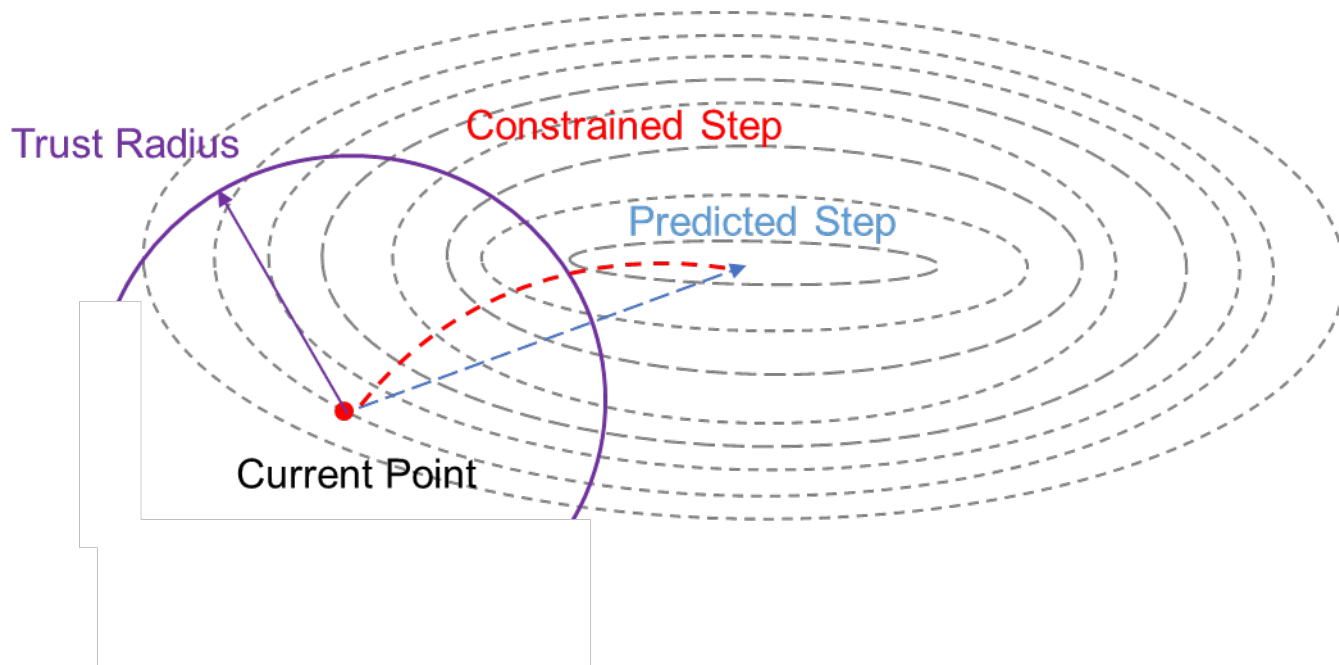

$$\square R = R_k + B_k s_{k+1} + s_{k+1} C_k s_{k+1} + \dots$$

- What about the higher-order term?
  - The linear model (predicted step is only valid for small enough steps)
  - What is small enough – needs a trust region
  - Surprisingly I don't think any code except Wien2k has trust regions for DFT



# Trust Region

- Total magnitude of step limited by a “Trust Region”; solve quadratic form to minimize the energy with this as a constraint






# Math issues

- There are cases with multiple fixed points
- There are cases where there is **no** downhill route from certain densities to the solution
- *At the solution* properties are good – but unpredictable otherwise
- Radius of convergence etc. changes with problem, details unclear
- Global minimum with atoms as well

# Numerical issues

- 
- Algorithms involve numerical integrations/differentiations – stability and conditioning errors
  - Some codes are more equal than others, e.g. break symmetry boundary conditions
  - Variables often effectively single-precision
  - Not a well researched topic (most codes have been written by physicists/chemists)

# Transparency issues

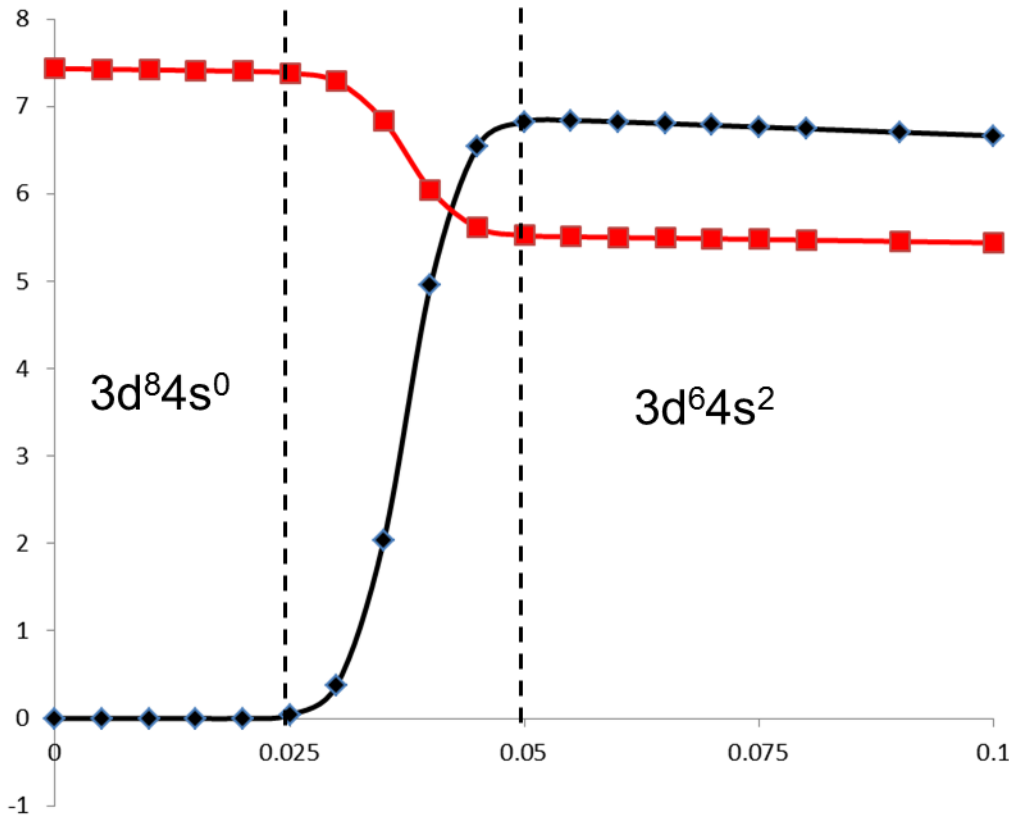
- All approaches require an inverse – how is rarely mentioned (and regularization)
- Codes can have rubber bands to hold them together (Wien2k did 15 years ago)
- They are not always written to be transparent
- Can have undisclosed variables (sometime not deliberate)

# Physics issues

- The fundamental character can change discontinuously, to illustrate
  - Problem starts as steam
  - Then condenses into water
  - Ends as ice
  - There might be a storm/lightning in the mix
- *This cannot be ignored in the math or algorithm.*

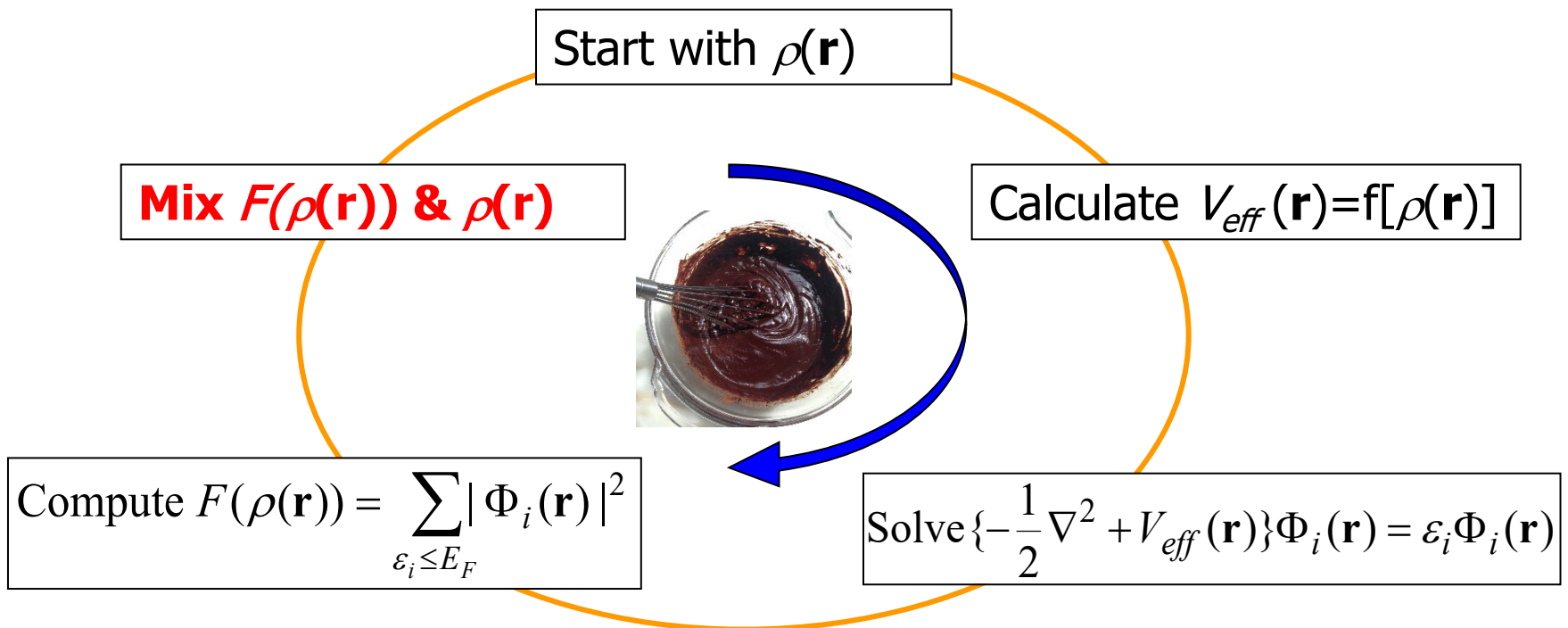


# Phase Transitions: Can change discontinuously



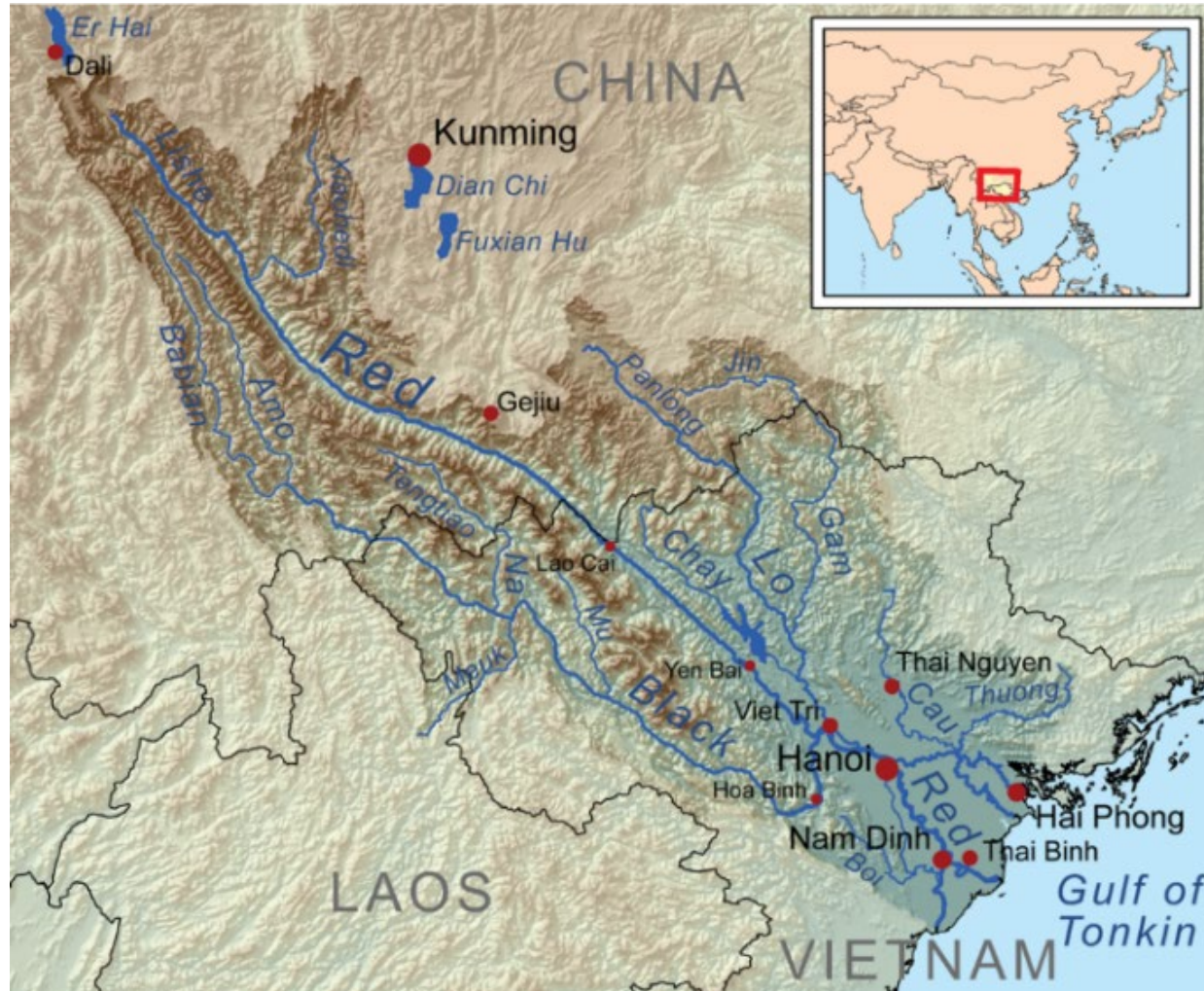
Electronic configuration of  $F(\rho)$  in the second step as a function of the size of the first Pratt step for an Fe atom, with the 4s occupancy within the muffin-tins in black (x10) and the 3d in red

# Analogy: water running downhill



*Solves a variational energy by a fixed-point method, implicit minimization*

# Flowing downhill can be simple



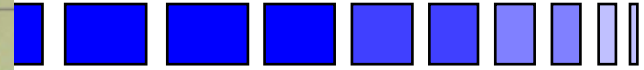
□ Red River



# Not so simple

- Mississippi
- Many wiggles, but no hard walls

# Problem Changes

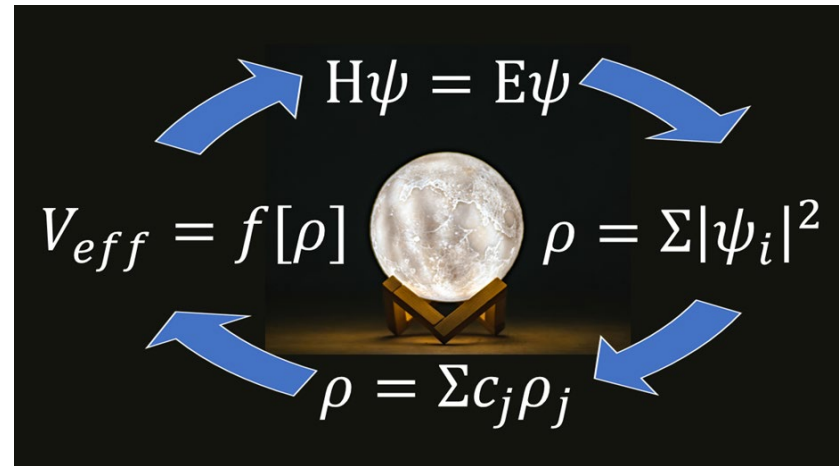


- Starts in the mountains
- Lake Powell
- Grand Canyon
- Down to sea...not!



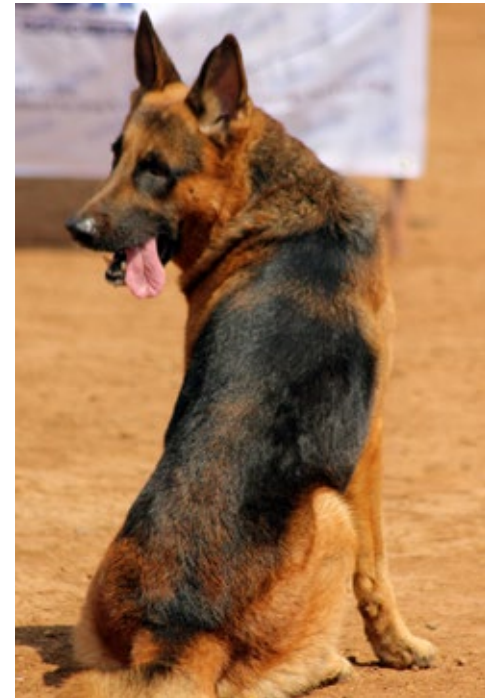
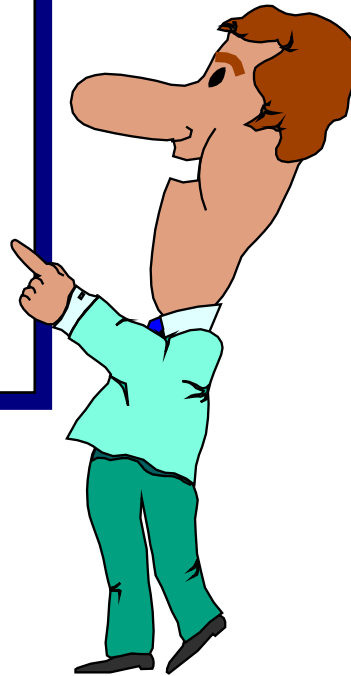
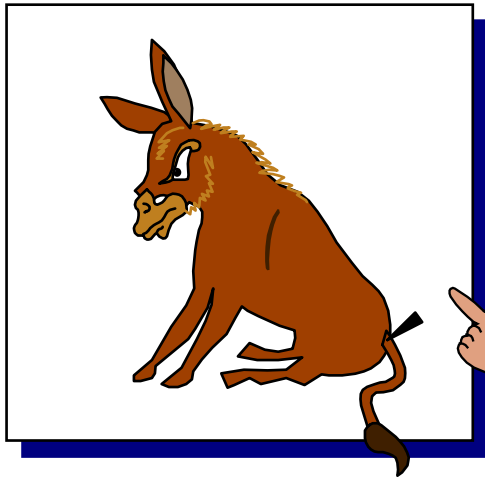
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# Predictive mixing (2021)

*Use best values for the last step*



DOI: 10.1021/acs.jctc.1c00630

# Reminder: general form

- Predicted step & residue

$$p_k = B_k^{-1} R_k = H_k R_k$$

$$R_k^P = B_k p_k$$

- Unpredicted step, residue greed  $\alpha_k$

$$u_k = R_k - B_k p_k = R_k^U$$

- General form

$$\rho_{k+1} = \rho_k + \alpha_k u_k + \beta_k p_k$$






# What we *should* have use

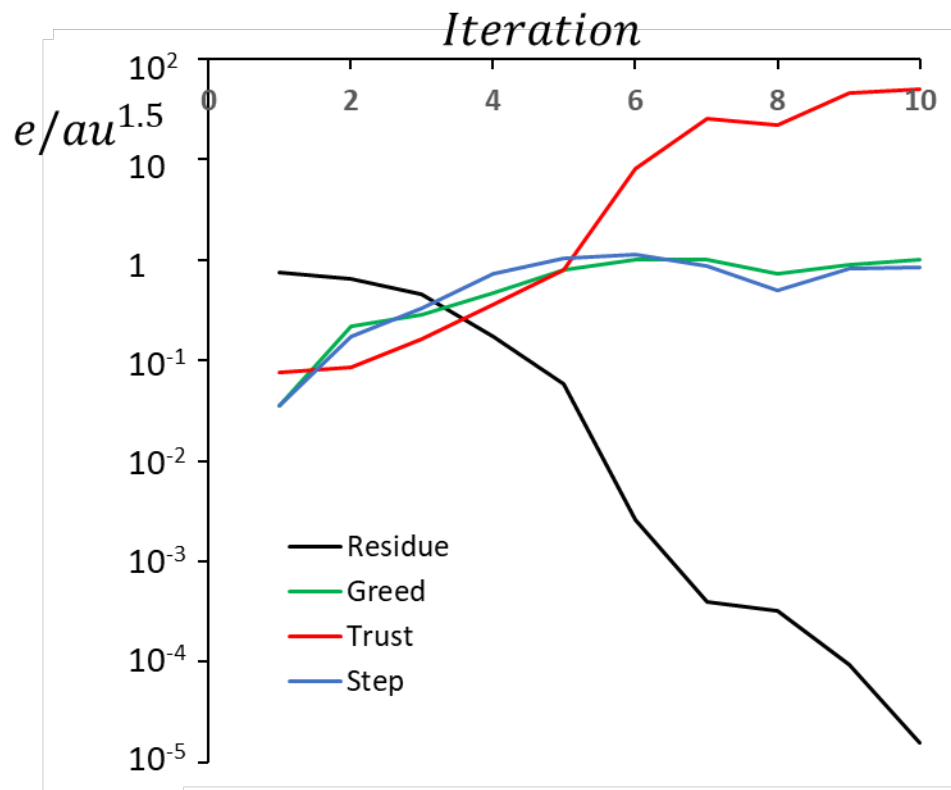


- Predicted step, find  $\beta'$  that minimizes  $\|R_{k-1}^P - \beta' B_k p_{k-1}\|$  then  $\beta_k = (\beta_{k-1} + \beta')/2$
- Unpredicted step find  $\alpha'$  that minimizes  $\|R_{k-1}^U - \alpha' B_k u_{k-1}\|$  then  $\alpha_k = (\alpha_{k-1} + \alpha')/2$
- For trust radius, use  $\|R_{k-1} - t' B_k s_k\|$  then  $t_k = (t_{k-1} + t')/2$
- General form, bounded by size
$$\rho_{k+1} = \rho_k + \alpha_k u_k + \beta_k p_k$$

# Standard additions

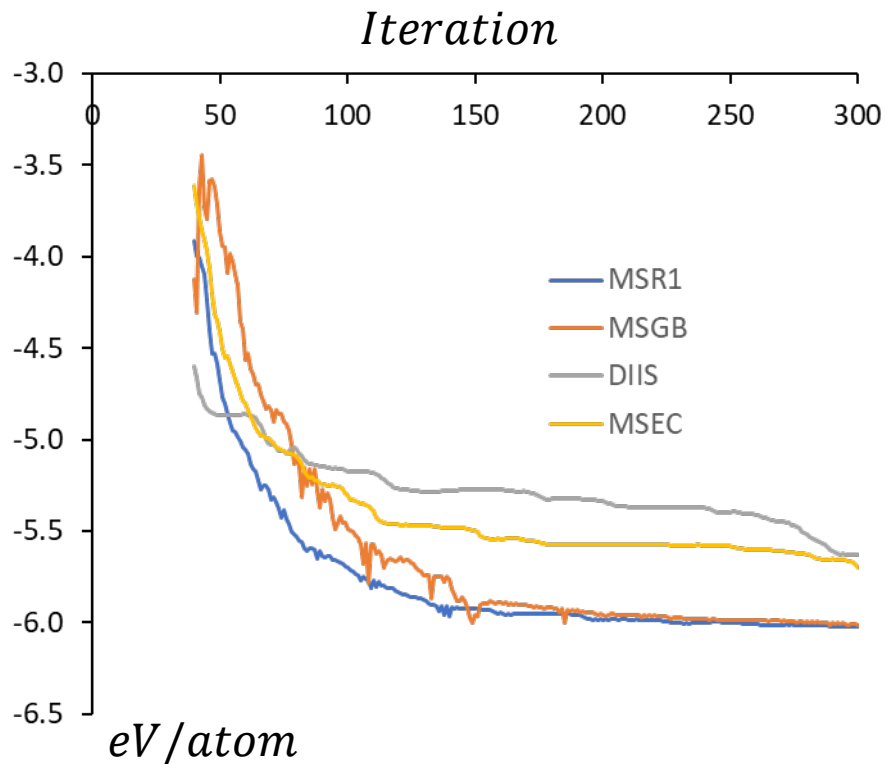
- 
- For a really bad step, recalculate along direction and discard new point (don't contaminate simplex gradients)
  - If a little bad, add to matrices & recalculate
  - SVD inversion with regularization
  - Start conservative
  - No user parameters except a couple that they think work but don't (they will fiddle)

# Does it work?



- Simple case, bulk MgO
- Greed/Trust start low, then rapidly increase

# More complex: atoms+density

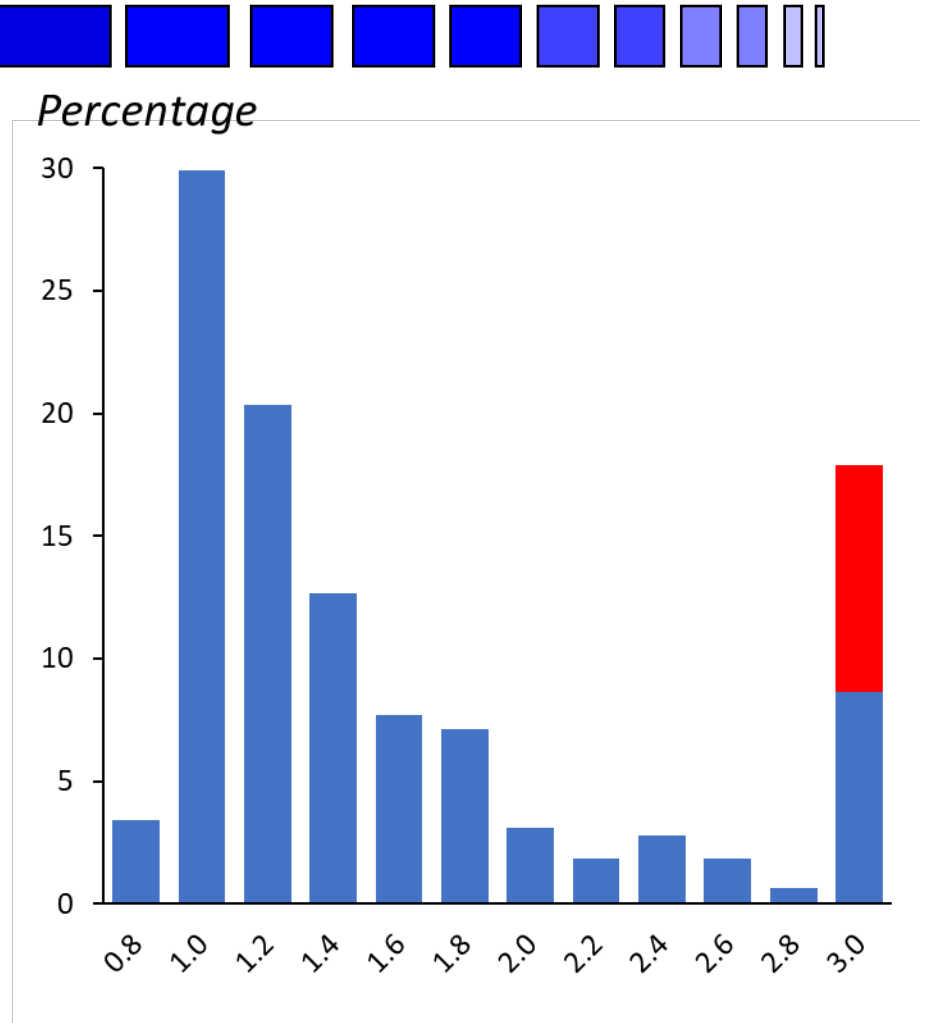


- MSR1: Predictive
- MSEC:  $\lambda=0$
- MSGB:  $\lambda=1$
- DIIS:  $\lambda=0$ ,  
different scaling

DIIS/MSEC don't converge  
MSGB is noisy (too greedy)

# Statistics

Ration of iterations to converge of predictive approach versus “best” (by search) of greed for 36 cases (blue), and 4 which never converged in red



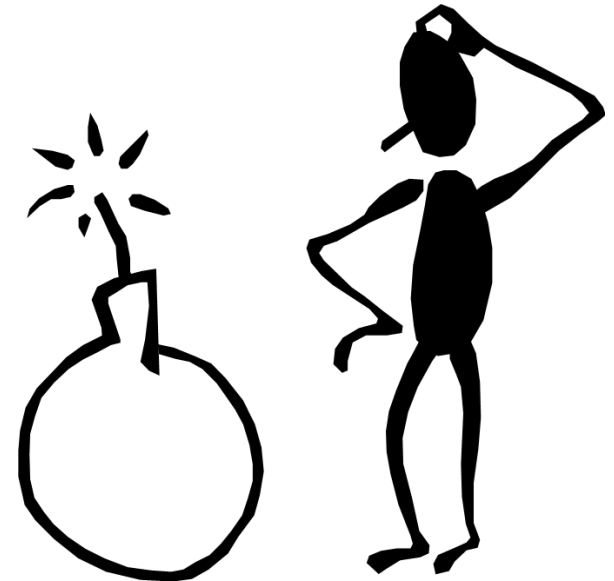
# Current status

- A detailed search can often find parameters as good – at the cost of doing 10 times or more as many calculations
- It (and earlier versions) have survived the acid test – novice users
- Follows the math
- No rubber bands



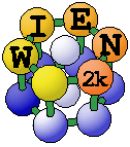
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# If something goes wrong...



NO  
THINKING

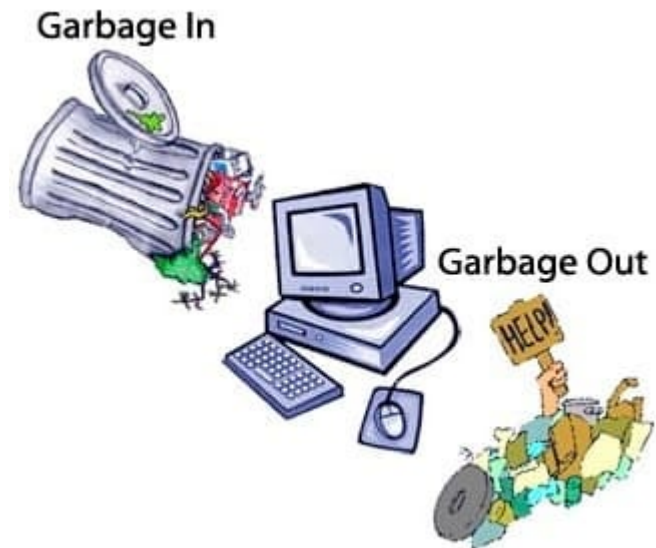


- In most cases this is because:
  - The problem is poorly posed
  - Too few k-points
  - The structure is very far from equilibrium
- So
  - Increase k-points
  - Maybe increase T (TEMPS), sometimes does not work
  - Think




# Some solutions are more equal than others

- How do we know the model is correct?
  - All models will refine to an improved fit, but this does not prove they are right
  - It is rarely plausible to test with DFT calculations all possible models



# Be careful

- 
- All software has bugs – by definition
  - Many computers are “broke”
  - Many calculations are at the limit of what can be done
  - Few computer codes will tell you “don’t do that you stupid #!?”
  - Never assume theory (or experiment) is correct
  - Think about what you are doing as computer experiments
  - Think science, not just typing at a terminal



# Questions ?



*It is through science that we prove, but  
through intuition that we discover.*

Jules H. Poincaré

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