Intelligent Water Drops with Perturbation Operators for Atomic Cluster Optimization

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Overview

The Intelligent Water Drops algorithm was modified (MIWD) and adapted to allow it to determine the most stable configurations, for the first time, of Lennard-Jones (LJ), Binary LJ (BinLJ), Morse and Janus Clusters. The algorithm, referred as MIWD+PerturbOp, is an unbiased type of algorithm where no a priori cluster geometry information and construction were used during initialization. Cluster perturbation operators were applied to clusters generated by MIWD to further generate lower energies. A limitedmemory quasi-Newton algorithm, called L-BFGS, was utilized to further relax clusters to its nearby local minimum.

Basic Properties of IWD a)

On LJ Clusters



Figure 2: Five independent LJ_{98} test runs (color lines) (10,000 iterations/run) for Chen bounding volume showing decline in cluster energy



Figure 3: Cubic Bounding volume and Grow Etch perturbation operator combination shows energy decline as tested on LJ_{38} .

perturbation Combination of operators (CombiOp) in Phase 2 (CutSplice+Knead, Cut-Splice+H1L2, CutSplice+H2L1, Knead+H1L2 and Knead+H2L1) were further done. Combinations were able to arrive at the GM except for N = 45 for $\sigma_{BB} = 1.05$ (Fig. 6).

MORSE : Tested for up to 60 atoms on 2 values of interparticle force range (a = 6, 14). MIWD+GrowEtch located the GM for most of the clusters except for N = 47, 55, 57, 58, 60 for a = 14 (Fig. 7).





Figure 7: configurations GMfrom MIWD+GrowEtch for selected Morse Clusters.

Figure 1: A path measures quality of connectivity between particles. (a) An IWD gathers soil (brown ellipse) as it flows from particle ito particle j while path(i,j) loses an amount of soil; (b) Soil gathered increases with IWD velocity; (c) An IWD travelling on a path with lesser soil, path(m,n), will gather more soil and higher velocity. (d) The algorithm progressively builds the cluster by choosing the connectivity with desirable measures.

FlowChart



Modifications to IWD

1. The probability of choosing a path depends on amount of soil and the potential energy. f(soil(i,j))n(i,j)IWD

Runs of MIWD alone shows improvement as iterations progress (Fig. 2). Final runs for MIWD+GrowEtch, utilizing spherical bounding volume for scattering of initial sites (Fig. 3), agrees with high-accuracy to (Cambridge Cluster Database) CCD results of up to 104 atoms. Compactness measures (Fig. 4) of this study versus CCD results show high-accuracy. Rotation and translation reveal that chiral clusters were generated (Fig. 5). MIWD+GrowEtch achieved relatively high-success rates for difficult clusters compared to Basin-Hopping with Occasional Jumping (BHOJ)(Table 1).



On Janus Clusters

MIWD+CombiOP was applied on Janus clusters using the LJ potential as the patchy particles model but where anisotropic attraction and repulsion is modulated by an orientational dependent term MV_{ang} (Fig. 8). Preliminary results were generated for cluster sizes N = 3 - 30 (Fig. 9). MIWD with GrowEtch and Patch Orientation Mutation produced the configurations with the lowest energies.



Figure 8: Orientation measure, MV_{ang} , for pairs of angles between 0 to 180. MV_{ang} for $\sigma = 90$ (Green) and $\sigma = 30$ (Red).



$$p_{i,j}^{IWD} = \frac{\int (e^{i(i)}(y))^{i(i,j)}}{\sum_{k \in V_{a}} IWD} f(soil(i,j))\eta(i,j)}$$

$$\eta(i,j) = \frac{1}{2+V_{t}ype(r_{i,j})}$$

$$V_{M} = e^{a(1-r_{i,j})} \left(e^{a(1-r_{i,j})} - 2\right)$$

$$V_{LJ}(r_{i,j}) = 4\varepsilon_{i,j} \left(\left(\frac{\sigma_{i,j}}{r_{i,j}}\right)^{12} - \left(\frac{\sigma_{i,j}}{r_{i,j}}\right)^{6}\right)$$

$$V_{J}(r_{i,j}) = V_{LJ}(r_{i,j})f\left(\Omega_{i}\right)f\left(\Omega_{j}\right)$$

$$f(\Omega_{i}) = -exp\left(\frac{\theta_{i,j}^{2}}{2\sigma^{2}}\right) + exp\left(\frac{(\theta_{i,j}-180)^{2}}{2\sigma^{2}}\right)$$
2. An appropriate heuristic undesirability factor,

$$HUD, \text{ is chosen to fit atomic cluster optimiza-
tion.$$

$$HUD_{i,j} = 2 + V_{type}(r_{i,j}) + \mu r_{i,j} + \beta(max(0, r_{i,j}^{2} - D^{2}))^{2}$$
3. Worst iteration agent, TIW, affects the soil
content as well.

$$soil_{i,j} = (1+\rho)soil_{i,j} + P_{i,j}$$

$$P_{i,j} = \rho(\frac{soil^{IWD}}{N-1})$$
4. L-BFGS was used as a relaxation algorithm

for IWDs.



Figure 5: Row 1 : Overlayed clusters showing unmatched positions. Row 2 : Rotated and translated clusters showing matching configurations.

On Binary LJ and Morse

BINARY LJ : Tested for up to 50 atoms on 6 instances of $\sigma_{BB} = 1.05 - 1.30$. MIWD+Knead rediscovered the global minima (GM) for most of the clusters except for N = 41,43, 45 -49 for $\sigma_{BB} = 1.05$ and N = 47 for $\sigma_{BB} = 1.10$. MIWD+CutSpliceVar rediscovered most of the GM except for N = 30-32 for σ_{BB} = 1.30, N = 35 for $\sigma_{BB} = 1.05, 1.15, N = 36, 39-50$ for σ_{BB} = 1.05 and N = 47, 49-50 for $\sigma_{BB} = 1.10$.

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