## Active Learning of Atomistic Force Fields

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Ab initio molecular dynamics is a powerful tool for accurately probing the dynamics of molecules and solids, but it is limited to system sizes on the order of 1000 atoms and time scales on the order of 10 ps. We present a scheme for rapidly training a machine learning model of the interatomic force field that approaches the accuracy of ab initio force calculations but can be applied to larger systems over longer time scales. Gaussian Process models are trained on-the-fly, with density-functional theory calculations of the atomic forces performed whenever the model encounters chemical configurations outside of the training set. We demonstrate the flexibility of our approach by testing it on vacancy diffusion in bulk metals.

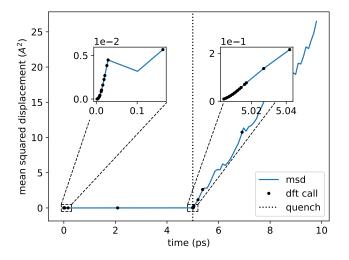


FIG. 1. Mean square displacement of aluminum quench.

	Solid	Liquid	Slab	Vacancy
OTF	test			
EAM				

TABLE I. On-the-fly force field error compared to a recent EAM potential.

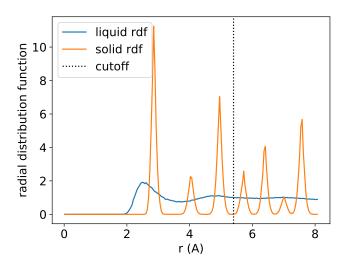


FIG. 2. RDF of Al quench.

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Algorithm 1 Active Learning of Atomistic Force Fields
Require: initial structure (positions, velocities, periodic cell)
Require: initial GP model (kernel and hyperparameters)
Require: \Delta t: molecular dynamics time step
Require: T: total simulation time
Require: \mathcal{U}: initial uncertainty threshold
 1: Initialize time: t = 0
   while t < T do
       predict forces and uncertainties with GP model
 3:
       if uncertainty above threshold then
 4:
          compute forces with DFT
 5:
          add highest uncertainty atom to training set
 6:
 7:
          update GP hyperparameters
 8:
          update structure with DFT forces
 9:
10:
          update structure with GP forces
       end if
11:
12:
       update time: t = t + \Delta t
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13: end while

Energy Kernel	$k_{ m inv}$	$\sigma^2 \sum_{c,p} k f_{\text{cut}}(\vec{d}_c) f_{\text{cut}}(\vec{d}_p)$
-	k	$\exp\left(-\frac{  \vec{d}_c - \vec{d}_p  ^2}{2\ell^2}\right)$
-	$\vec{d}^{(2)}$	$(r_{i_1})$
-	$d^{(3)}$	$(r_{i_1}, r_{i_2}, r_{i_1, i_2})$
Force Kernel	$\frac{\partial^2 k_{\text{inv}}}{\partial \xi_i \partial \chi_j}$	$\sigma^2 \sum_{c,p} (k_0 + k_1 + k_2 + k_3)$
-	$k_0$	$k \frac{\partial f_{\mathrm{cut}}(\vec{d}_c)}{\partial \xi_i} \frac{\partial f_{\mathrm{cut}}(\vec{d}_p)}{\partial \chi_j}$
-	$k_1$	$rac{\partial k}{\partial \xi_i} f_{ m cut}(ec{d}_c) rac{\partial f_{ m cut}(ec{d}_p)}{\partial \chi_j}$
-	$k_2$	$rac{\partial k}{\partial \chi_j} rac{\partial f_{ m cut}(ec{d}_c)}{\partial \xi_i} f_{ m cut}(ec{d}_p)$
-	$k_3$	$rac{\partial^2 k}{\partial \xi_i \partial \chi_j} f_{ m cut}(ec{d}_c) f_{ m cut}(ec{d}_p)$
-	$\frac{\partial k}{\partial \xi_i}$	$\frac{kB_1}{\ell^2}$
	$B_1$	$\sum_{q=1}^{\frac{ND_1}{\ell^2}} \sum_{q=1}^{N-1} \frac{(r_{i_q} - r_{j_q})\xi_{i_q}}{r_{i_q}} - \frac{kB_2}{\ell^2}$
-	$\frac{\partial k}{\partial \chi_j}$	$-rac{kB_2}{\ell^2}$
-	$B_2$	$\sum_{q=1}^{N-1} \frac{(r_{iq} - r_{jq})\chi_{jq}}{r_{jq}}$
-	$\frac{\partial^2 k}{\partial \xi_i \chi_j}$	$\frac{k}{\ell^4} \left( A\ell^2 - B_1 B_2 \right)$
-	A	$\sum \frac{\xi_{iq}\chi_{jq}}{r_{iq}r_{jq}}$
$\ell$ Derivative	$\frac{\partial^3 k_{\text{inv}}}{\partial \ell \partial \xi_i \partial \chi_j}$	$\sigma^{2} \sum_{c,p} \left( \frac{\partial k_{0}}{\partial \ell} + \frac{\partial k_{1}}{\partial \ell} + \frac{\partial k_{2}}{\partial \ell} + \frac{\partial k_{3}}{\partial \ell} \right)$
-	$\frac{\partial k}{\partial \ell}$	$rac{k  ec{d}_{ec{c}}-ec{d}_{ec{p}}  ^2}{I^3}$
-	$\frac{\partial^2 k}{\partial \ell \partial \xi_i}$	$B_1\left(\frac{1}{\ell^2}\frac{\partial k}{\partial \ell} - \frac{2k}{\ell^3}\right)$
-	$\frac{\partial^2 k}{\partial \ell \partial \chi_j}$	$-B_2\left(\frac{1}{\ell^2}\frac{\partial k}{\partial \ell} - \frac{2k}{\ell^3}\right)$
-	$\frac{\partial^3 k}{\partial \ell \partial \xi_i \partial \chi_j}$	$\left(A\ell^2 - B_1 B_2\right) \left(\frac{\partial k}{\partial \ell} \frac{1}{\ell^4} - \frac{4k}{\ell^5}\right) + \frac{2kA}{\ell^3}$
$\sigma$ Derivative	$\frac{\partial^3 k_{\text{inv}}}{\partial \sigma \partial \xi_i \partial \chi_j}$	$2\sigma \sum_{c,p} (k_0 + k_1 + k_2 + k_3)$

TABLE II. Quantities used to calculate the smoothed  $N\text{-}\mathrm{body}$  force kernel and its derivatives.