Multistate Bennett Acceptance Ratio replaces histogram reweighting for vapor-liquid coexistence calculations; Multistate Bennett Acceptance Ratio to enable rapid force field parameterization; Multistate Bennett Acceptance Ratio as a substitute for histogram reweighting when optimizing non-bonded parameters

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#### **ABSTRACT**

# **Key points**

- 1. MBAR is equivalent to standard histogram reweighting approaches in the limit of zero bin width
- 2. MBAR allows for estimating VLE properties of any force field
- 3. Scaling epsilon values post-simulation is straightforward with MBAR
- 4. Basis functions are an efficient means for re-computing energy of Mie potentials

## **Outline**

## 1 Introduction

- 1. Accurate and efficient computation of vapor-liquid coexistence is an important but challenging task for molecular simulation
- 2. Reweighting simulation outputs between different states is a well-known and powerful tool, e.g., histogram reweighting of GCMC results
- 3. Force field parameterization with VLE data is an arduous and time-consuming task
- 4. Hamiltonian scaling (histogram reweighting for multiple force fields) allows for estimating VLE properties of multiple force fields from single set of simulations
- 5. Messerly et al. demonstrated how MBAR can be combined with ITIC to predict VLE properties. Several Weakness of ITIC is need large systems, which is not ideal for MBAR

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- 6. Gross demonstrated benefits of  $\varepsilon$ -scaling for "individualized", i.e., compound-specific parameter sets
- 7. In this study, we demonstrate that:
  - (a) MBAR yields indistinguishable results from histogram reweighting (HR)
  - (b) Scaling epsilon is straightforward by scaling U with MBAR
  - (c) MBAR can estimate VLE properties for multiple force fields simultaneously
  - (d) Basis functions allow for rapid computation of VLE for non-simulated Mie parameter sets

## 2 Methods

#### 2.1 Force fields

- 1. Simulations are performed for united-atom generalized Lennard-Jones (a.k.a., Mie  $\lambda$ -6) force fields
- 2. We investigate the TraPPE, Potoff-generalized, Potoff (S/L), and NERD force fields
- 3. Details of force fields

# 2.2 Simulation set-up

- 1. Simulations performed by Mick et al. are reanalyzed using MBAR
- Additional simulations are performed in GCMC ensemble using GPU optimized Monte Carlo (GOMC)
- 3. Simulation specifications, i.e., box size, number of steps, type of moves, etc.
- 4. State points (chemical potentials and temperatures) simulated are same as those utilized in Mick et al.

#### 2.3 Multistate Bennett Acceptance Ratio

- 1. Traditionally, histogram reweighting (HR) has been applied with GCMC to calculate vapor-liquid coexistence properties
- 2. Present histogram reweighting equations
- 3. Discuss how to compute phase equilibria by equating pressures
- 4. Discuss how to compute heat of vaporization
- 5. In this study, we demonstrate how to compute VLE using MBAR-GCMC
- 6. Procedure is identical to that utilized for HR but using the MBAR equations
- 7. Present MBAR equations
- 8. MBAR for  $\theta = \theta_{ref}$  is mathematically equivalent to histogram reweighting in the limit of zero bin width
- 9. MBAR-GCMC allows for prediction of multiple force fields from single simulation without modifying force fields mid-simulation (i.e., Hamiltonian scaling approach)
- 10. MBAR uncertainties are computed using bootstrap resampling

#### 2.4 Basis functions

- 1. When applying MBAR to different parameter sets,  $\theta \neq \theta_{ref}$ , it is necessary to recompute energies
- 2. Basis functions accelerate the recompute energy step by storing the repulsive and attractive contributions that can be scaled by  $\varepsilon$  and  $\sigma$
- 3. Basis functions are computed from GOMC using the recompute feature for different  $\varepsilon$  and  $\sigma$  and solving system of equations

### 3 Results

- 1. We validate that MBAR and HR are indistinguishable by re-analyzing the simulation results of Mick et al. and Barhaghi et al. utilizing MBAR
- 2. Epsilon scaling for all the compounds that Mohammad has U and N values for (branched alkanes and alkynes) and which have good experimental data
- 3. We estimate Potoff generalized and NERD VLE from TraPPE simulations, Potoff S/L from Potoff generalized, and TraPPE from Potoff generalized
- 4. For  $\lambda_{ref} = 12$  and  $\lambda_{rr} = 16$ , MBAR-GCMC predicts vapor density, vapor pressure, and heat of vaporization more accurately than liquid density
- 5. For  $\lambda_{ref} = 12$  and  $\lambda_{rr} = 12$ , i.e., computing NERD from TraPPE simulations, MBAR-GCMC predicts all four properties accurately
- 6. We present how basis functions allow for rapid computation of wide range of parameter sets:
  - (a) *n*-hexane
  - (b) 2-methylpropane
  - (c) 2,2-dimethylpropane
  - (d) cyclopentane or cyclohexane
- 7. We provide supporting information with basis functions for several branched alkanes with TraPPE and Potoff force fields

### 3.1 Figures

- 1. Percent deviation between MBAR and HR results for rholiq, rhovap, Psat, and DeltaHv
- 2. Comparison between MBAR bootstrapping and analytical uncertainties and HR uncertainties
- 3. Scaling of epsilon post-simulation for branched alkanes and alkynes
- 4. Prediction of VLE for  $\lambda_{ref} \neq \lambda_{rr}$
- 5. Prediction of VLE for  $\lambda_{ref} = \lambda_{rr} = 12$
- 6. Prediction of VLE for  $\lambda_{ref} = \lambda_{rr} = 16$
- 7. Two-D scans of scoring functions for  $\varepsilon \sigma$  of CH3 (a) and CH2 (b) for *n*-hexane

- 8. Two-D scans of scoring functions for  $\varepsilon \sigma$  of CH3 (a) and CH (b) for 2-methylpropane
- 9. Two-D scans of scoring functions for  $\varepsilon \sigma$  of CH3 (a) and C (b) for 2,2-dimethylpropane
- 10. Two-D scans of scoring functions for  $\varepsilon \sigma$  of CH2 for cyclopentane or cyclohexane (reference is TraPPE)

## 4 Discussion/Limitations/Future work

- 1. We recommend that future GCMC-VLE studies report the snapshots of N and U and/or basis functions to recompute U as this allows for future force field optimization
- 2. Improvements are possible with multiple  $\theta$  or simulating a range of  $\mu$  values

# 5 Conclusions

# 6 Acknowledgments

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# 7 Supporting Information

#### 7.1 MBAR VLE estimates

Provide tables of MBAR estimates

### 7.2 Basis functions

1. Validation that basis functions give accurate energies

#### 7.3 Raw data

1. Comparison of 2-D histograms for TraPPE and Potoff. MBAR overlap, possible? Probably not without rerunning the simulations.