

# TorchMD-NET Integration Report

## Machine Learning Forcefield Benchmarks

Date: 2025-11-17

### System Information:

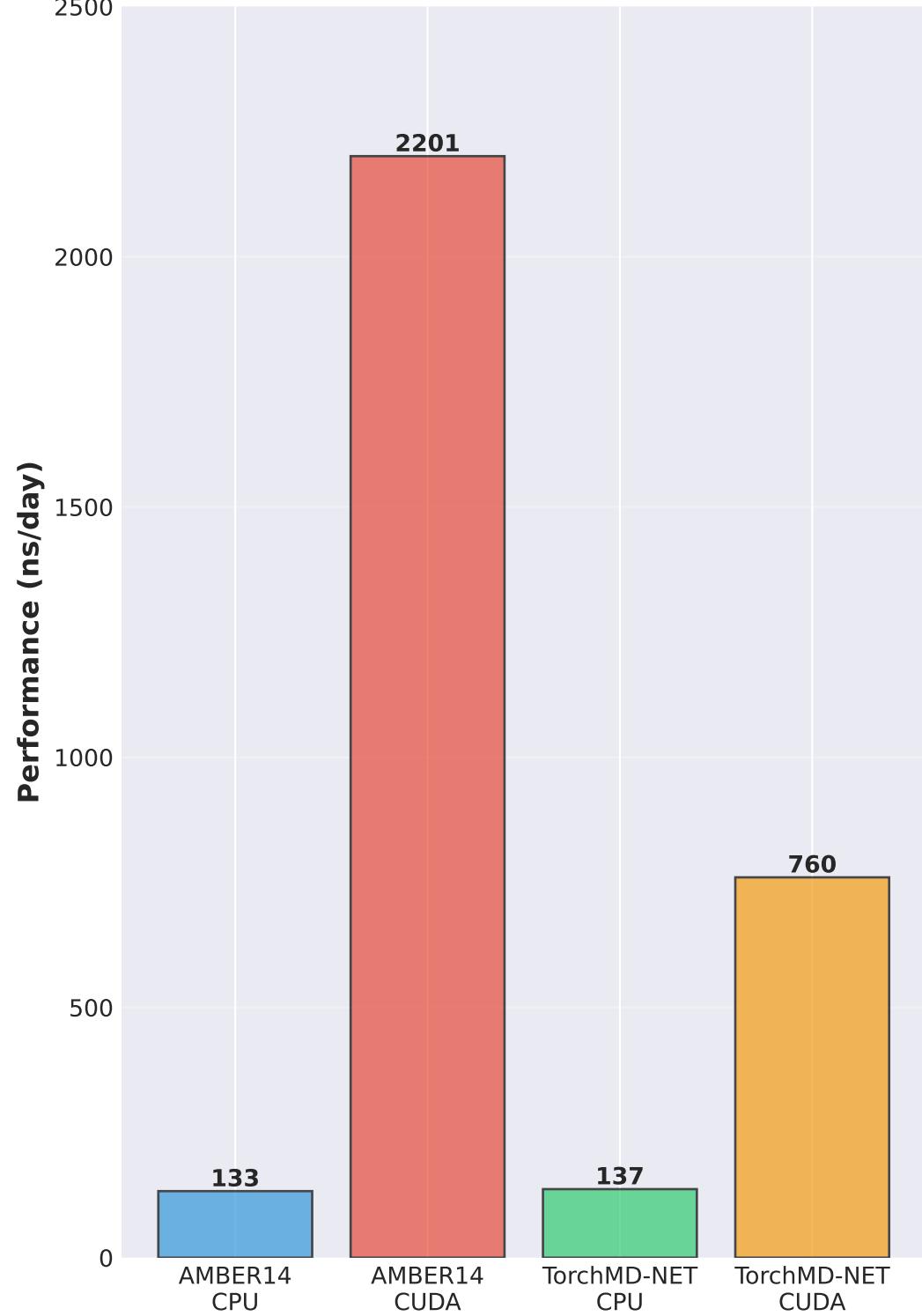
- GPU: Tesla T4 (CUDA 12.6)
  - PyTorch: 2.7.1
  - TorchMD-NET: 2.4.12
- OpenMM: Latest with OpenMM-Torch

### Test System:

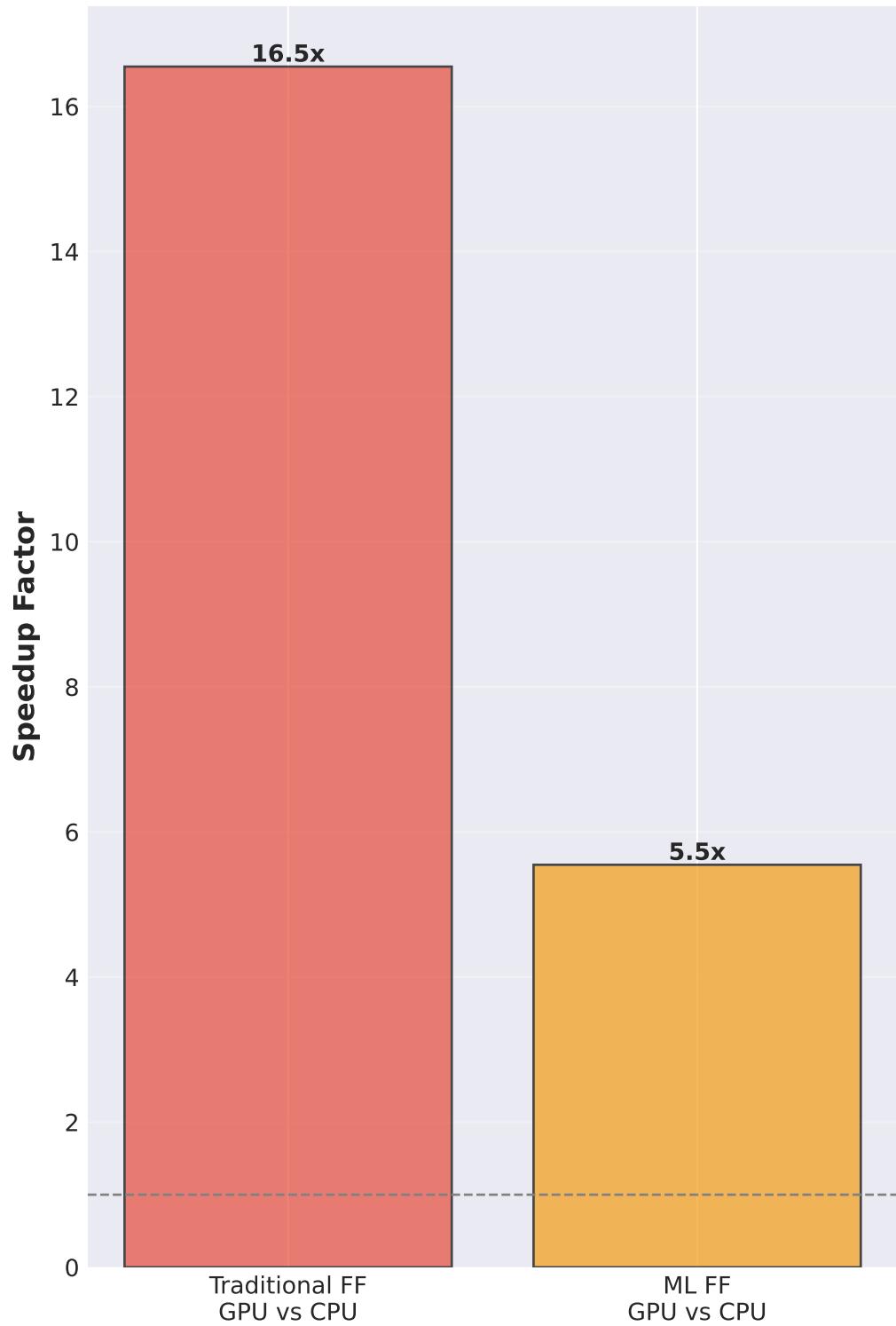
- RNA Hairpin Structure
- 700 atoms, 43 residues
- Benchmark: 1000 MD steps

*Complete Installation & Integration Test*

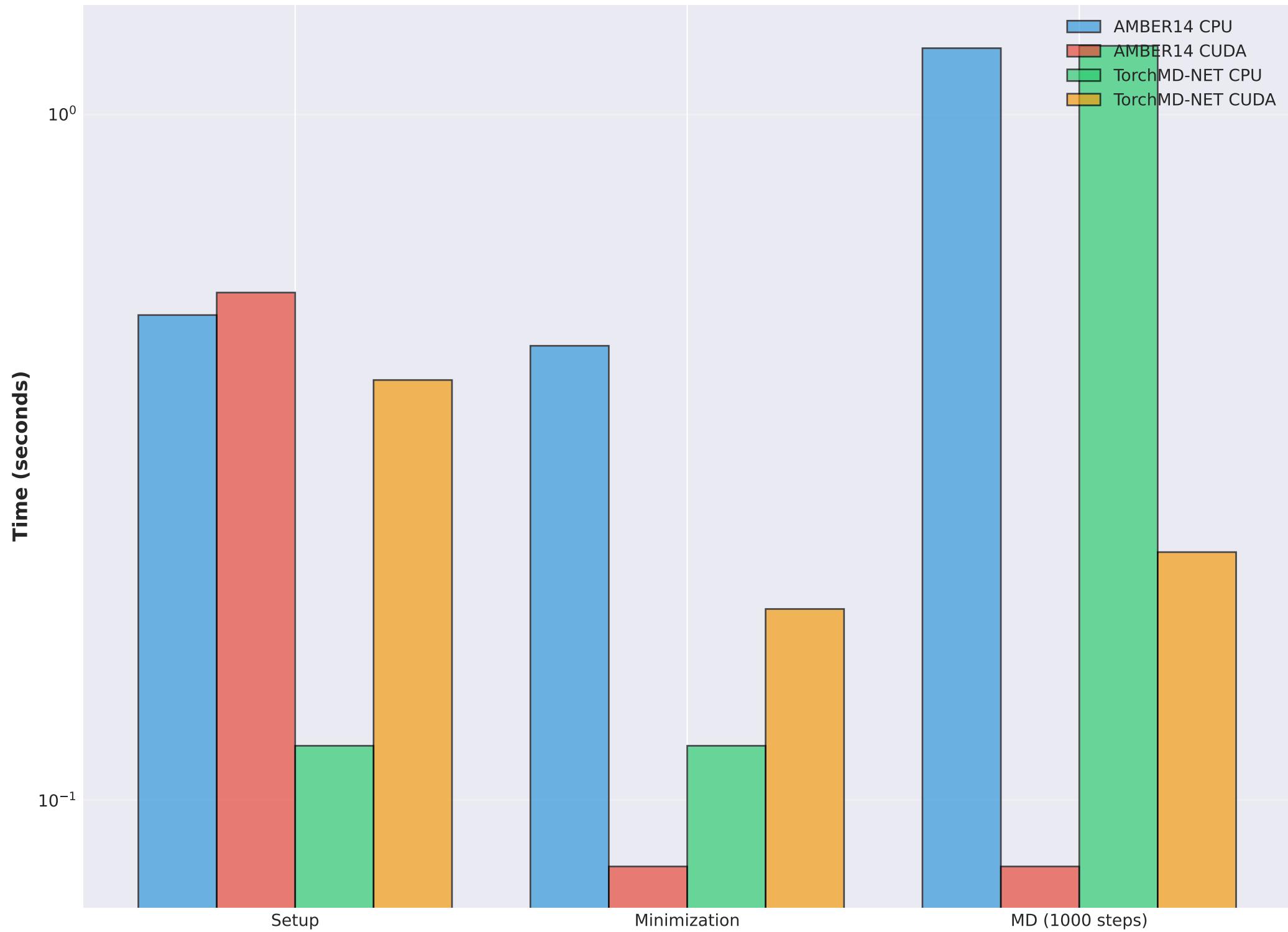
**MD Performance Comparison**  
**(1000 steps, 700 atoms)**



**GPU Acceleration**  
**(CUDA vs CPU)**



## Timing Breakdown by Phase (700 atoms)



# TorchMD-NET Model Configuration

## Model Architecture: TensorNet

### Parameters:

- Total Parameters: 756,865
- Embedding Dimension: 128
  - Number of Layers: 2
- Radial Basis Functions: 32
- RBF Type: expnorm (Exponential Normal)
  - Activation: SiLU (Swish)

### Cutoff & Neighbors:

- Cutoff Distance: 5.0 Å
  - Max Neighbors: 64
- Equivariance Group: O(3)

### Architecture Features:

- Equivariant message passing
- Tensor field network layers
  - Scalar output for energy
  - Forces computed via autograd

### Integration Details:

- Framework: OpenMM with TorchForce
- Unit Conversion: nm  $\leftrightarrow$  Å, kJ/mol  $\leftrightarrow$  eV
- TorchScript: JIT compiled for performance
  - Platform Support: CPU and CUDA

Note: Current benchmarks use UNTRAINED model for framework testing. Production use requires trained models on appropriate QM datasets (e.g., SPICE, ANI).

# Installation & Integration Summary

✓ Installation Completed Successfully

## Components Installed:

- PyTorch 2.7.1 with CUDA 12.6
- TorchMD-NET 2.4.12 (compiled from source)
- OpenMM with OpenMM-Torch plugin
- All Python dependencies

## Tests Passed:

- ✓ Model creation (CPU & GPU)
  - ✓ Energy computation
- ✓ Force evaluation via autograd
  - ✓ OpenMM integration
- ✓ MD simulation propagation
  - ✓ TorchScript compilation

## Key Technical Solutions:

- Import Order: PyTorch before OpenMM (library conflict)
- Generator Fix: List comprehension for OpenMM's sum()
  - Unit Conversion: Automatic nm↔Å, kJ/mol↔eV
- TorchScript: JIT compilation for deployment

## Available Scripts:

- test\_torchmdnet.py - Validation tests
- torchmdnet\_openmm\_integration.py - Integration demo
- comprehensive\_ff\_benchmark.py - Benchmark suite

## Documentation:

- TORCHMDNET\_SETUP\_COMPLETE.md - Full setup guide
- TORCHMDNET\_BENCHMARK\_ANALYSIS.md - Detailed analysis
  - This PDF - Visual summary

# Next Steps & Recommendations

## Production Deployment Roadmap

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### Phase 1: Model Training (Required for Production)

- Train TorchMD-NET on relevant QM dataset
  - SPICE: Diverse organic molecules & peptides
    - ANI: Small molecules, proteins
    - Custom: Your specific system type
      - Validation on test set
  - Checkpoint saving for deployment

### Phase 2: Performance Optimization

- Benchmark trained model on target systems
- Tune hyperparameters (cutoff, neighbors, layers)
  - Profile GPU memory usage
  - Optimize batch sizes for throughput

### Phase 3: Production Integration

- Load trained checkpoint in production scripts
  - Set up simulation protocols
  - Implement analysis pipelines
    - Run convergence tests

### Phase 4: Scaling & Deployment

- Test on production-scale systems
  - Multi-GPU support (if needed)
    - Long-timescale simulations
  - Integration with analysis tools

## Quick Start Commands:

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```
# Test installation
python3 test_torchmdnet.py

# Run demo simulation
python3 torchmdnet_openmm_integration.py hairpin.pdb \
--platform CUDA --steps 100

# Benchmark comparison
python3 comprehensive_ff_benchmark.py your_system.pdb
```

## Resources:

- TorchMD-NET repo: /home/ubuntu/MD/torchmd-net
- Example configs: /home/ubuntu/MD/torchmd-net/examples/
  - Documentation: TORCHMDNET\_SETUP\_COMPLETE.md