

Tuning HMC parameters with gradients

James C. Osborn
Argonne National Laboratory

Lattice 2023, FNAL

Tuning HMC Parameters

- Hybrid Monte Carlo (HMC)
 - Choose random gaussian momentum, \mathbf{p}
 - Integrate Hamiltonian, $H(\mathbf{p}, \mathbf{U}) = \mathbf{p}^2/2 + S(\mathbf{U})$, along equations of motion
 - $\mathbf{p}', \mathbf{U}' = F(\mathbf{p}, \mathbf{U}, \alpha)$
 - α = integration time, τ , other MD parameters
 - Conditionally accept new configuration
 - $P_{acc}(\mathbf{p}', \mathbf{U}', \mathbf{p}, \mathbf{U}) = \min[1, \exp(-\Delta H)]$
 - $\Delta H = H(\mathbf{p}', \mathbf{U}') - H(\mathbf{p}, \mathbf{U})$
- Tuning parameters effects efficiency of update
 - Integration time, step size, higher order schemes, multiple time scales
 - Also action parameters (fermion action w/ Hasenbusch ratios)
- Can be tuned by hand, using measurements of error terms, forces
- Experimenting with autotuning using gradient information

Integrator parameters

- For example: 2 step integrator
 - ABABA pattern
 - A = gauge field update
 - B = momentum update with force term
- $e^{\varepsilon(A+B) + E} = e^{\varepsilon\lambda A} e^{\varepsilon/2 B} e^{\varepsilon(1-2\lambda) A} e^{\varepsilon/2 B} e^{\varepsilon\lambda A}$
 - $E = \varepsilon^3 \{ (1-6\lambda+6\lambda^2)/12 [A,[B,A]] + (1-6\lambda)/24 [B,[B,A]] \} + \varepsilon^5 \dots$
- Minimum norm solution (I. P. Omelyan, I. M. Mryglod, R. Folk 2001)
 - Assume similar magnitude of operators $[A,[B,A]]$ and $[B,[B,A]]$
 - Minimize norm of error term ($\lambda \approx 0.193\dots$)
 - In LQCD, terms typically not equal, need to tune
- Can also consider higher order integrators (force gradient) with more parameters

Cost function

- MD integration time: τ
- After N HMC updates (momentum refresh, accept/reject)
 - effective MD integration time $\approx \tau \text{sqrt}(\langle P_{acc} \rangle N)$
- Number of updates needed to go T effective MD time units
 - $N_T = T^2 / (\langle P_{acc} \rangle \tau^2)$
- Cost = $N_T \times$ (cost per update)
- For simplicity using
 - (cost per update) = # force evaluations (N_{force})
 - $T = 1$

Loss function optimization

- Want to minimize
 - Cost = $N_{force} / (\langle P_{acc} \rangle \tau^2)$
- Involves average in denominator, inconvenient
- Instead minimize
 - Loss = $-\langle P_{acc} \rangle \tau^2$
- Using methods from ML
 - Calculate gradient of loss, use to update parameters
 - Using Adam optimizer, accumulates gradient during update
 - Single update stream (no batch)
 - Currently updating parameters after every step

Calculating gradients

- Need gradient of P_{acc} , function of $H(\mathbf{p}', \mathbf{U}') \equiv H'$ $\varepsilon = \tau / n$
 - $(\mathbf{p}', \mathbf{U}') = [e^{\varepsilon\lambda A} e^{\varepsilon/2 B} e^{\varepsilon(1-2\lambda) A} e^{\varepsilon/2 B} e^{\varepsilon\lambda A}]^n (\mathbf{p}, \mathbf{U})$
- Using back propagation
- Generate gradients from chain rule
- $\partial H' / \partial \alpha = \sum_k (\partial H' / \partial \mathbf{q}_k) (\partial \mathbf{q}_k / \partial \alpha$ [fixed \mathbf{q}_{k-1}])
- $\mathbf{q}_k = (\mathbf{p}_k, \mathbf{U}_k)$, state after k^{th} update step (A or B)
- $\partial H' / \partial \mathbf{q}_k = (\partial H' / \partial \mathbf{q}_{k+1}) (\partial \mathbf{q}_{k+1} / \partial \mathbf{q}_k)$
- Need gradients of all update steps (gauge update, forces, ...)
- Need to save all intermediate states

Gradient of gauge force

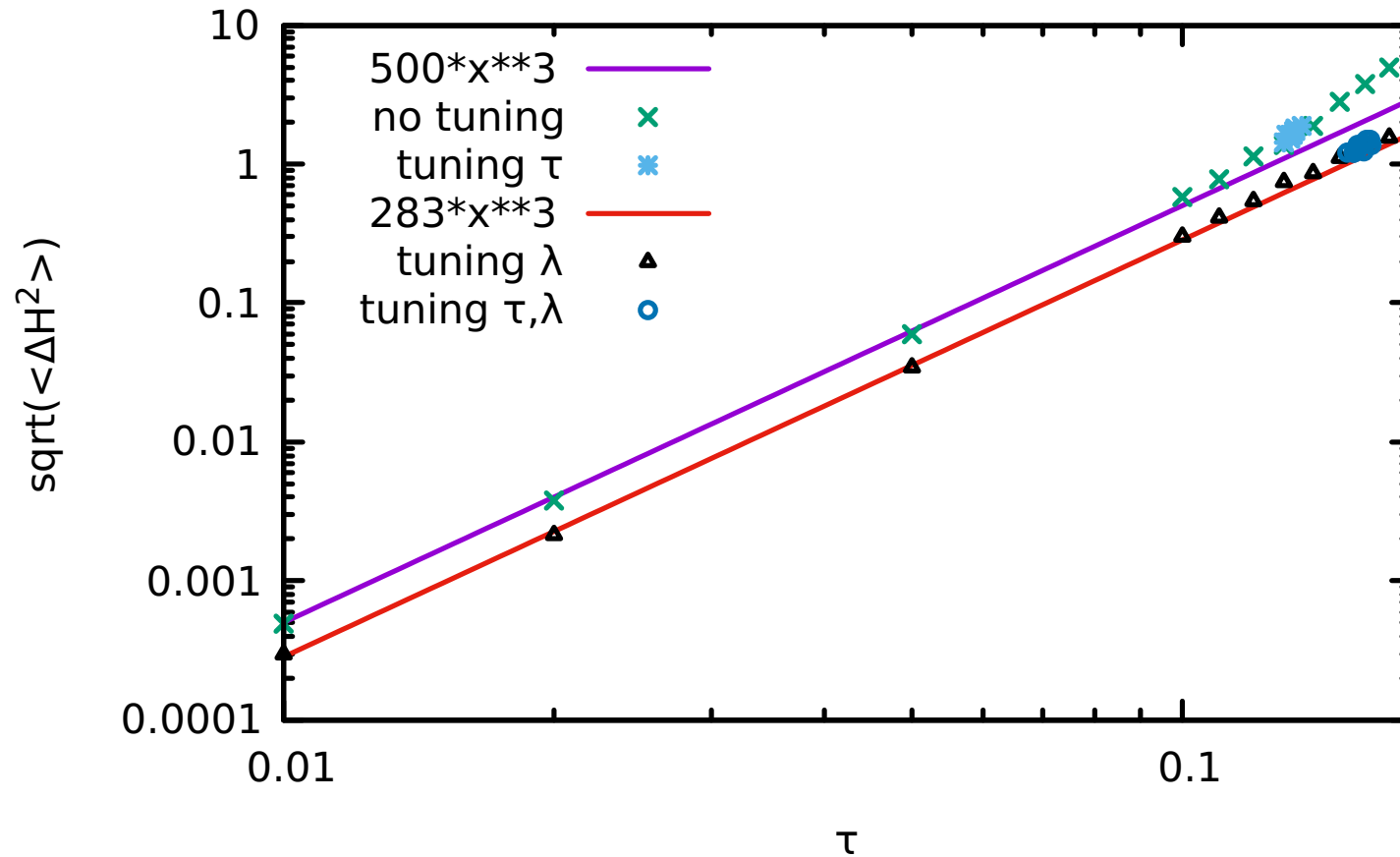
- Gauge force
 - Main piece is calculating staples
 - $S = U_\nu(x) U_\mu(x+\nu) U_\nu^\dagger(x+\mu) + \text{backward staple}$
- Chain rule
 - $dH'/dU = (dH'/dS) (dS/dU) = C dS/dU$
 - $dH'/dU = C_\nu(x) U_\mu(x+\nu) U_\nu^\dagger(x+\mu)$
+ $U_\nu(x) C_\mu(x+\nu) U_\nu^\dagger(x+\mu)$
+ $U_\nu(x) U_\mu(x+\nu) C_\nu^\dagger(x+\mu)$
+ **backward**
- Also need adding/scaling/multiplying fields, traceless anti-Hermitian projection
 - Relatively easy to work out gradients

Tests on pure gauge

- Implemented in QEX
 - LFT framework in Nim language
 - <https://github.com/jcosborn/qex>
 - Using simple “tape” implementation, save list of operations
 - Run forwards to do update, run backwards for gradient
- $12^3 \times 24$ lattice (running on single desktop)
- Plaquette action at $\beta = 5.6$
- Starting from thermalized config
- 200 tuning updates, 400 measurement updates

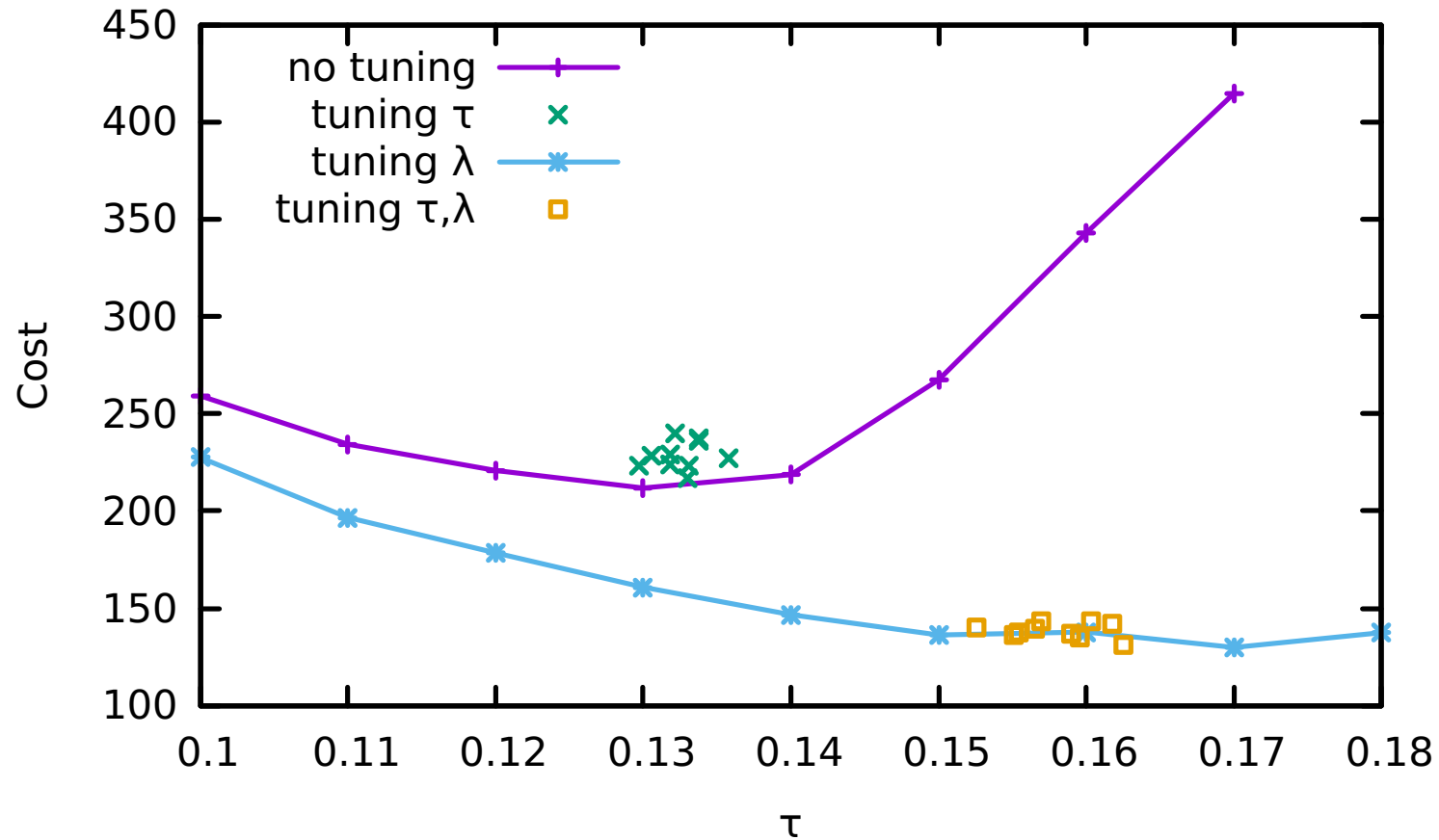
Gauge ABABA: Error vs. trajectory length

Single integrator copy ($n=1$, $\varepsilon = \tau$)



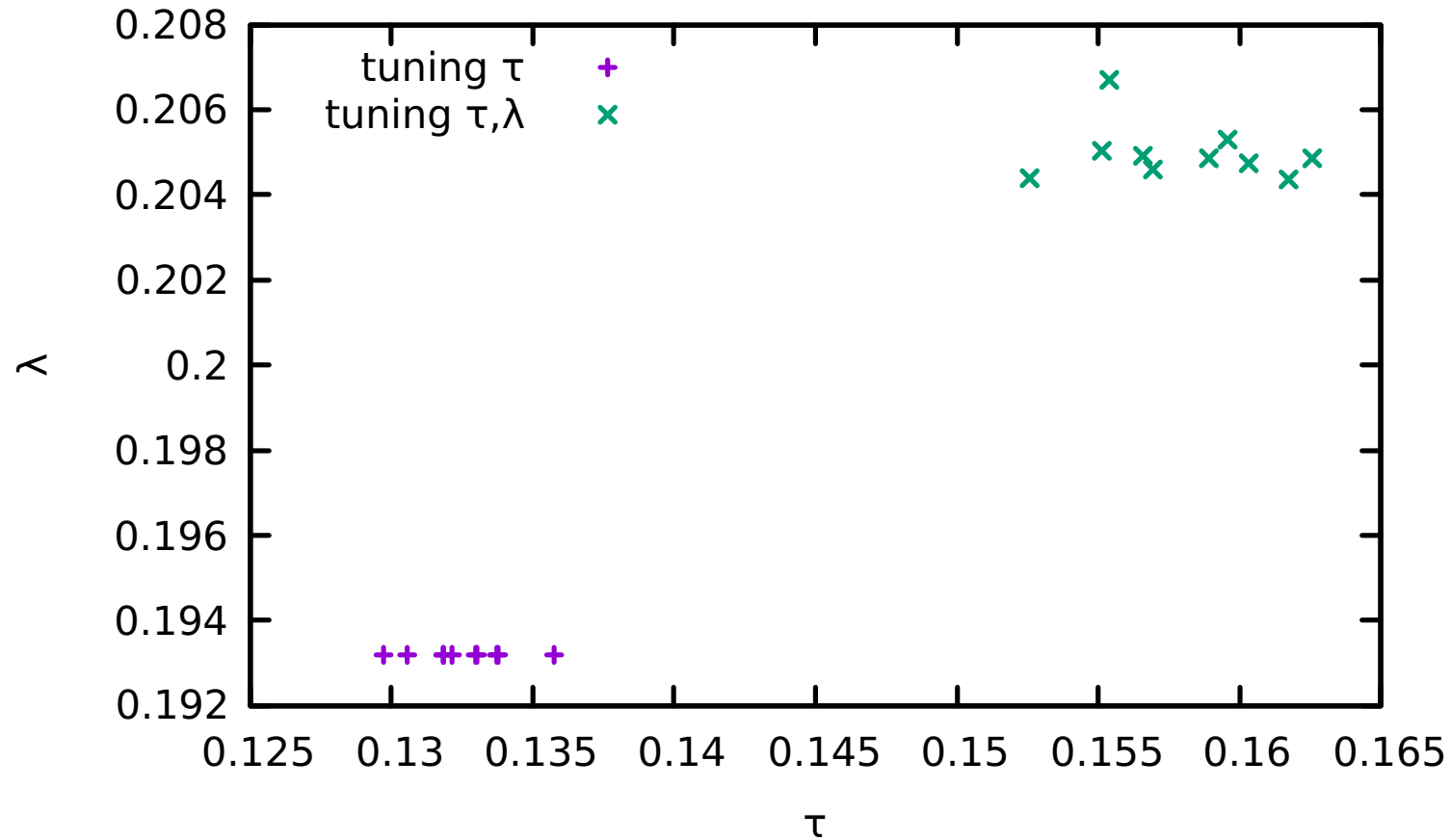
Gauge ABABA: Cost vs. trajectory length

Single integrator copy ($n=1$, $\varepsilon = \tau$)



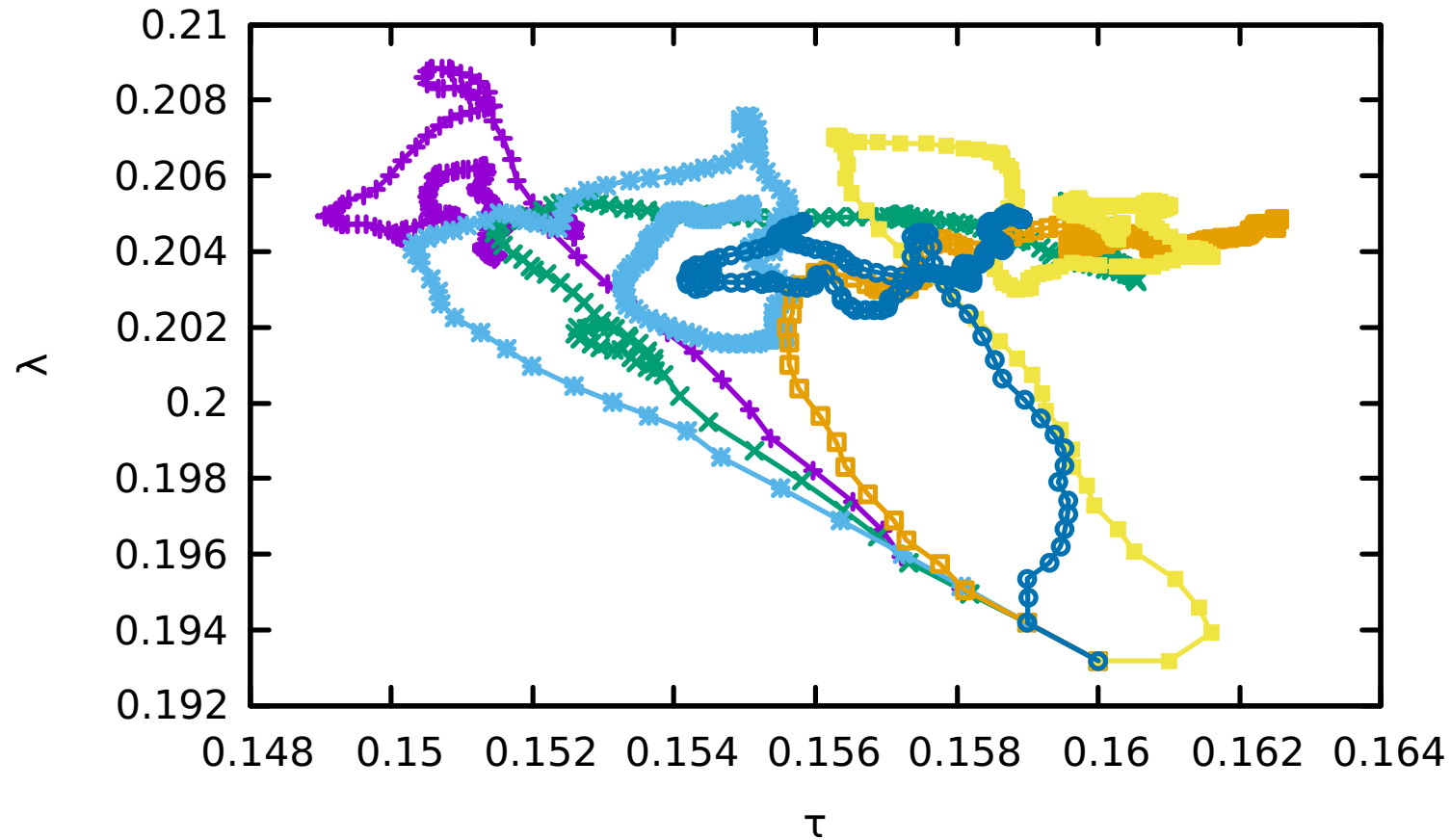
Gauge ABABA: Tuned parameters

Single integrator copy ($n=1$, $\varepsilon = \tau$)

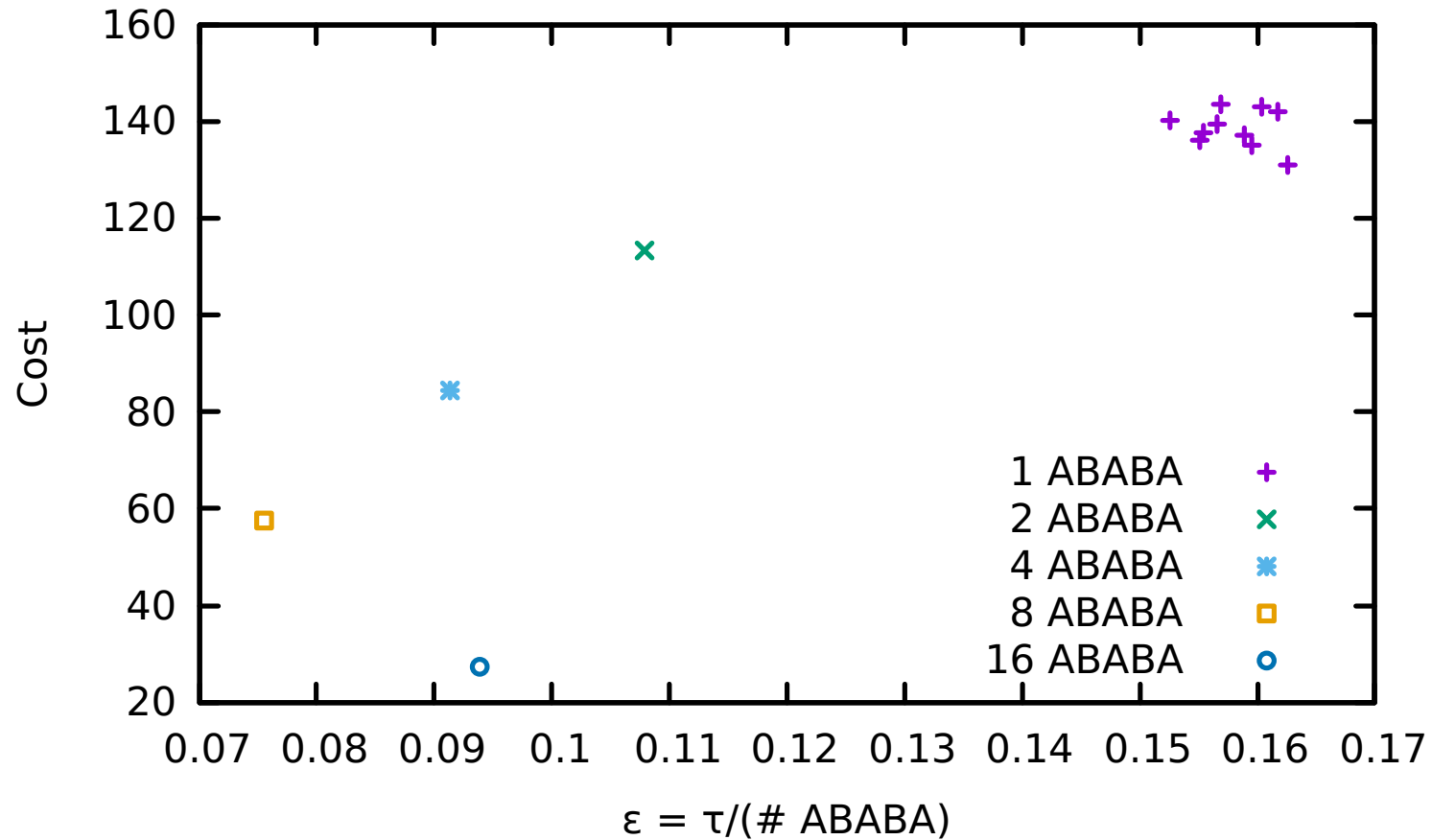


Gauge ABABA: Parameter paths during tuning

Single integrator copy ($n=1, \epsilon = \tau$)



Gauge ABABA: Multiple copies

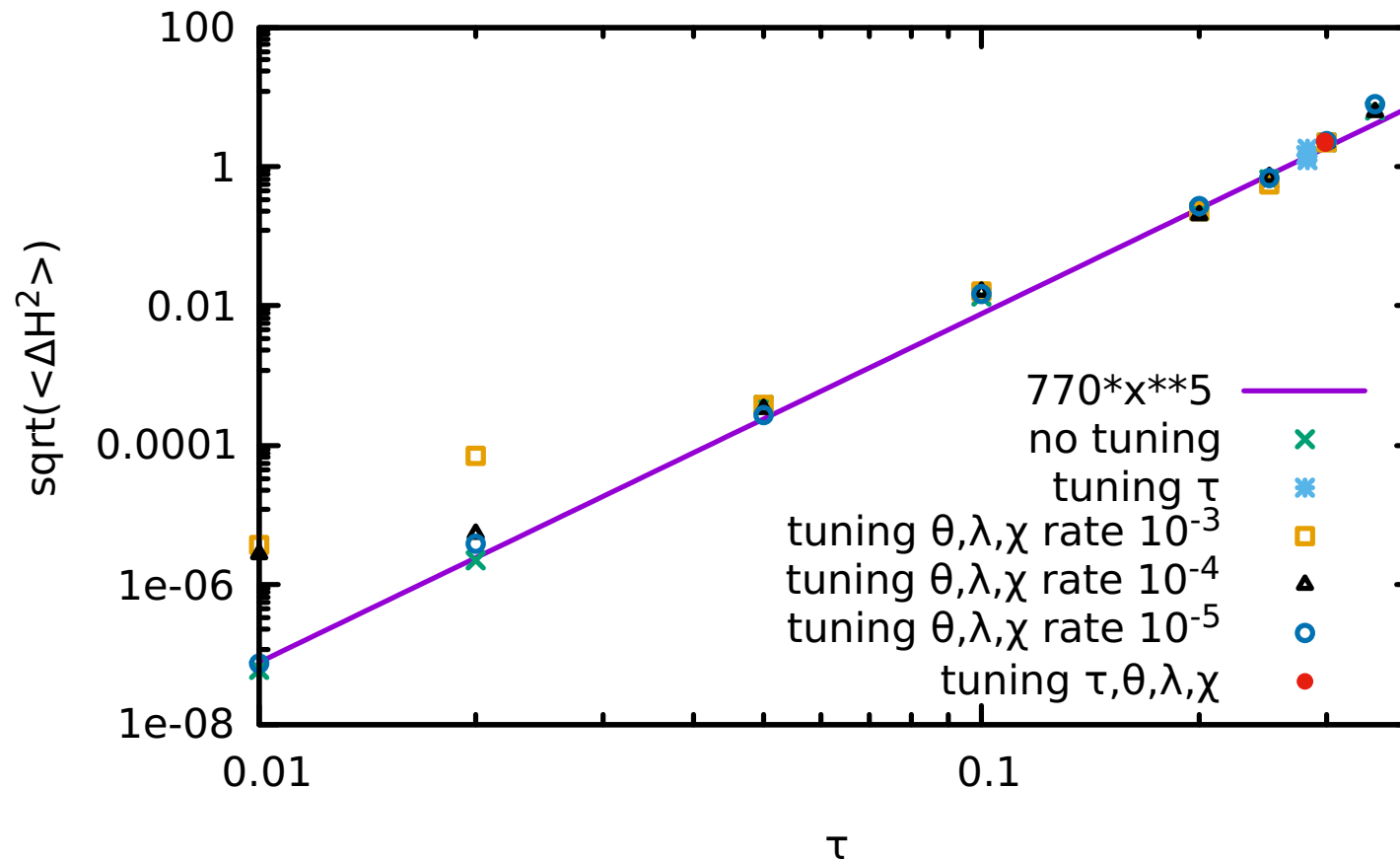


4th order integrator

- Using Omelyan, et al.'s recommended 4th order integrator
(I. P. Omelyan, I. M. Mryglod, R. Folk 2002)
- Uses force gradient term
 - Approximated using 2 force evaluations (H. Yin, R. D. Mawhinney 2011)
- ABACABA pattern
 - C = force gradient
 - $e^{\varepsilon\vartheta} A e^{\varepsilon\lambda} B e^{\varepsilon/2(1-2\vartheta)} A e^{\varepsilon(1-2\lambda)} B + (\varepsilon^3)\chi C e^{\varepsilon/2(1-2\vartheta)} A e^{\varepsilon\lambda} B e^{\varepsilon\vartheta} A$
 - 3 parameters: ϑ, λ, χ
 - Minimum norm solution cancels 3rd order error, minimizes 5th order

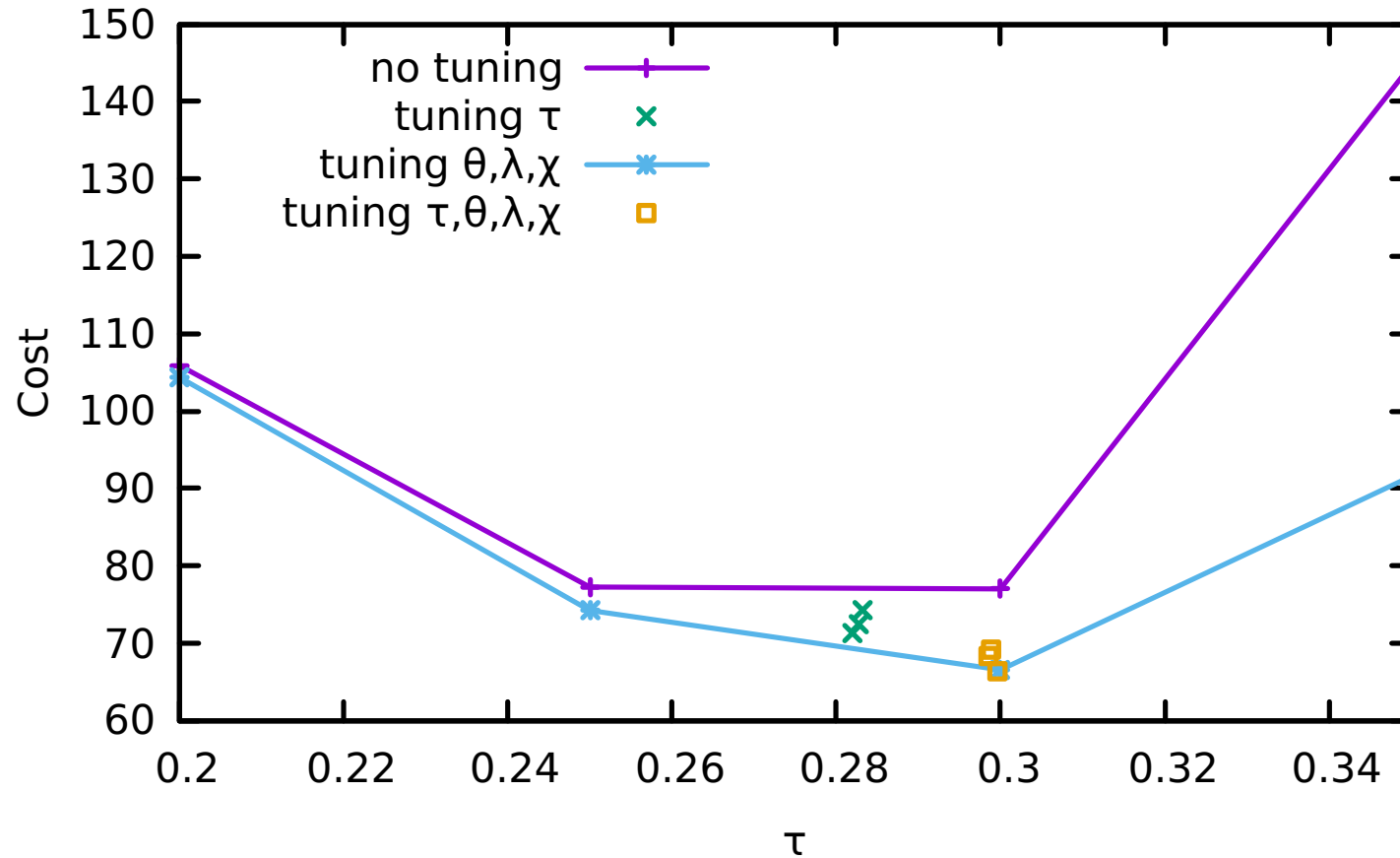
Gauge ABACABA: Error vs. trajectory length

Single integrator copy ($n=1$, $\varepsilon = \tau$)



Gauge ABACABA: Cost vs. trajectory length

Single integrator copy ($n=1$, $\varepsilon = \tau$)



Gradient of fermion force

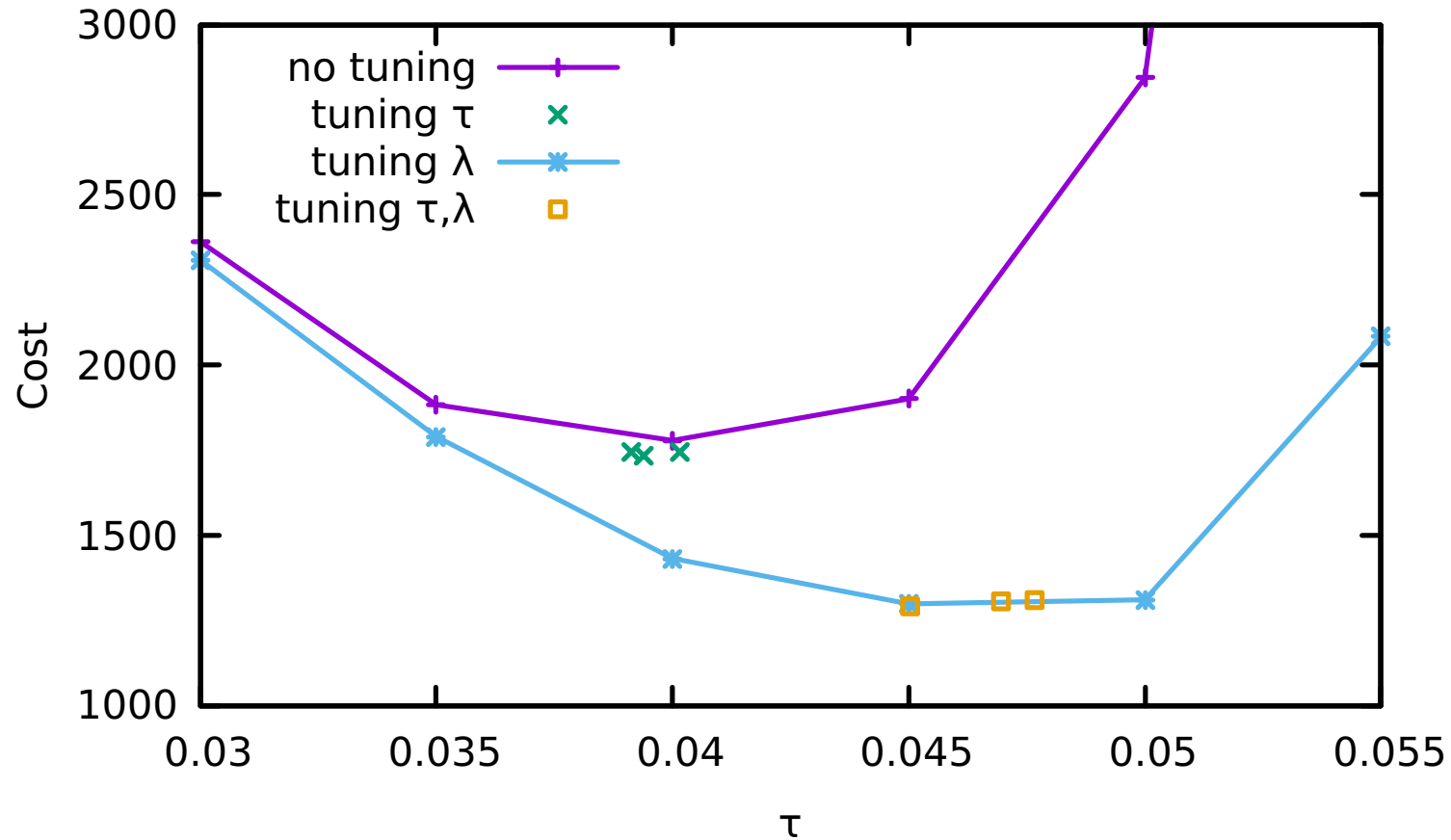
- Action term
 - $S = \phi^\dagger (D^\dagger D)^{-1} \phi$
- Force
 - $F_x = - \partial S / \partial U_x = \phi^\dagger (D^\dagger D)^{-1} (\partial D^\dagger D / \partial U_x) (D^\dagger D)^{-1} \phi + h.c.$
- Gradient of momentum update
 - $\partial H' / \partial U_Y = (\partial H' / \partial F_x) (\partial F_x / \partial U_Y) = C_x \partial F_x / \partial U_Y$
- Break into 2 pieces
 - $\psi = (D^\dagger D)^{-1} \phi \Rightarrow C_w \partial \psi_w / \partial U_Y = \phi^\dagger (D^\dagger D)^{-1} (\partial D^\dagger D / \partial U_Y) (D^\dagger D)^{-1} C_w$
 - $C_x \partial / \partial U_Y \psi^\dagger (\partial D^\dagger D / \partial U_x) \psi = \partial / \partial U_Y \psi^\dagger (D^\dagger D_c + D_c^\dagger D) \psi$
 - $D_c = C_x \partial D / \partial U_x$

Fermion tests

- Plain staggered, no rooting, 4 continuum flavors
- $12^3 \times 24$ lattice
- $\beta = 5.6$
- $m = 0.04$
 - $a M_\pi \sim 0.5$
- Cost function only counts fermion force evaluations (not gauge)

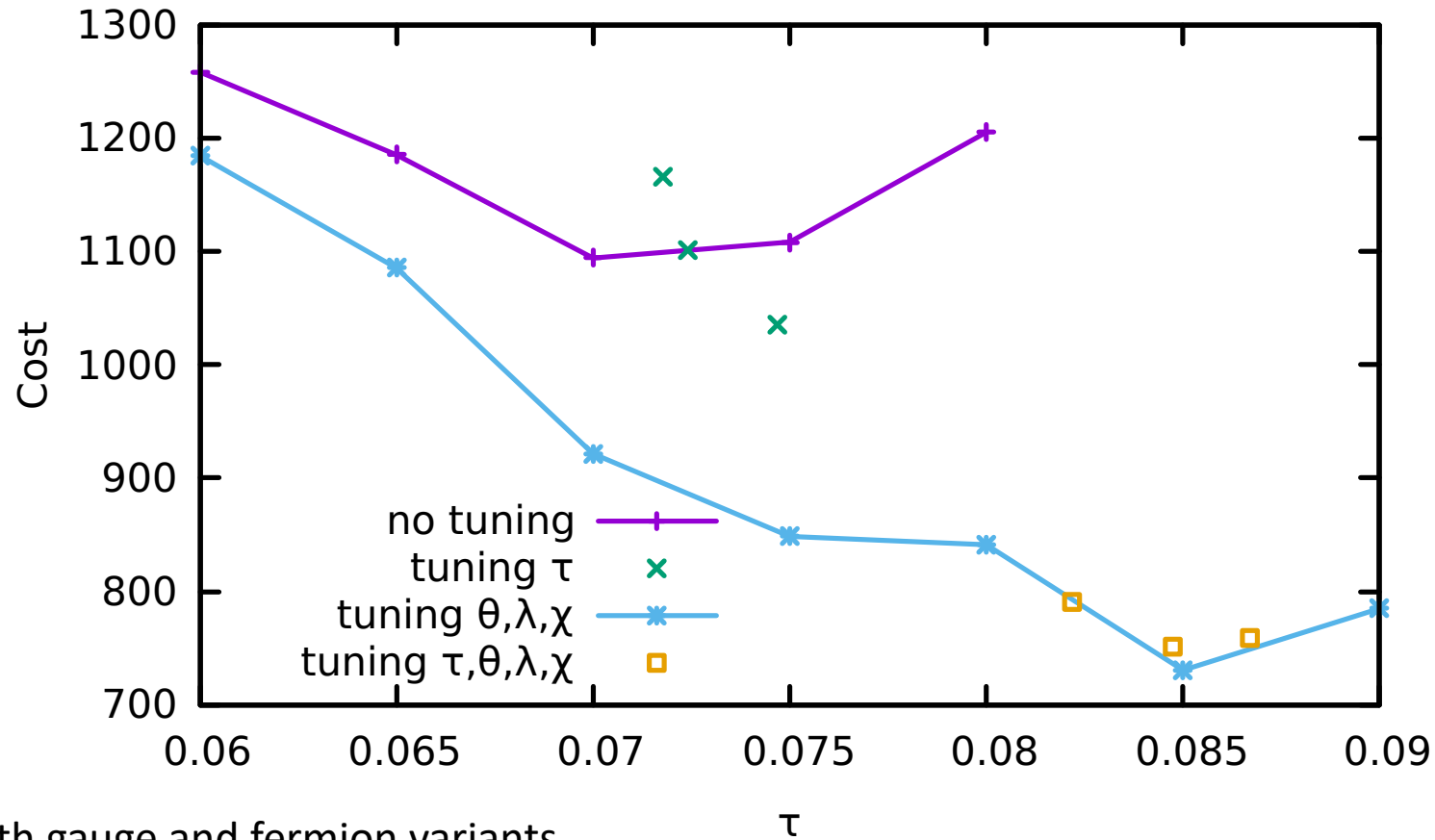
Staggered ABABA: Cost vs. trajectory length

Single integrator copy ($n=1, \varepsilon = \tau$)



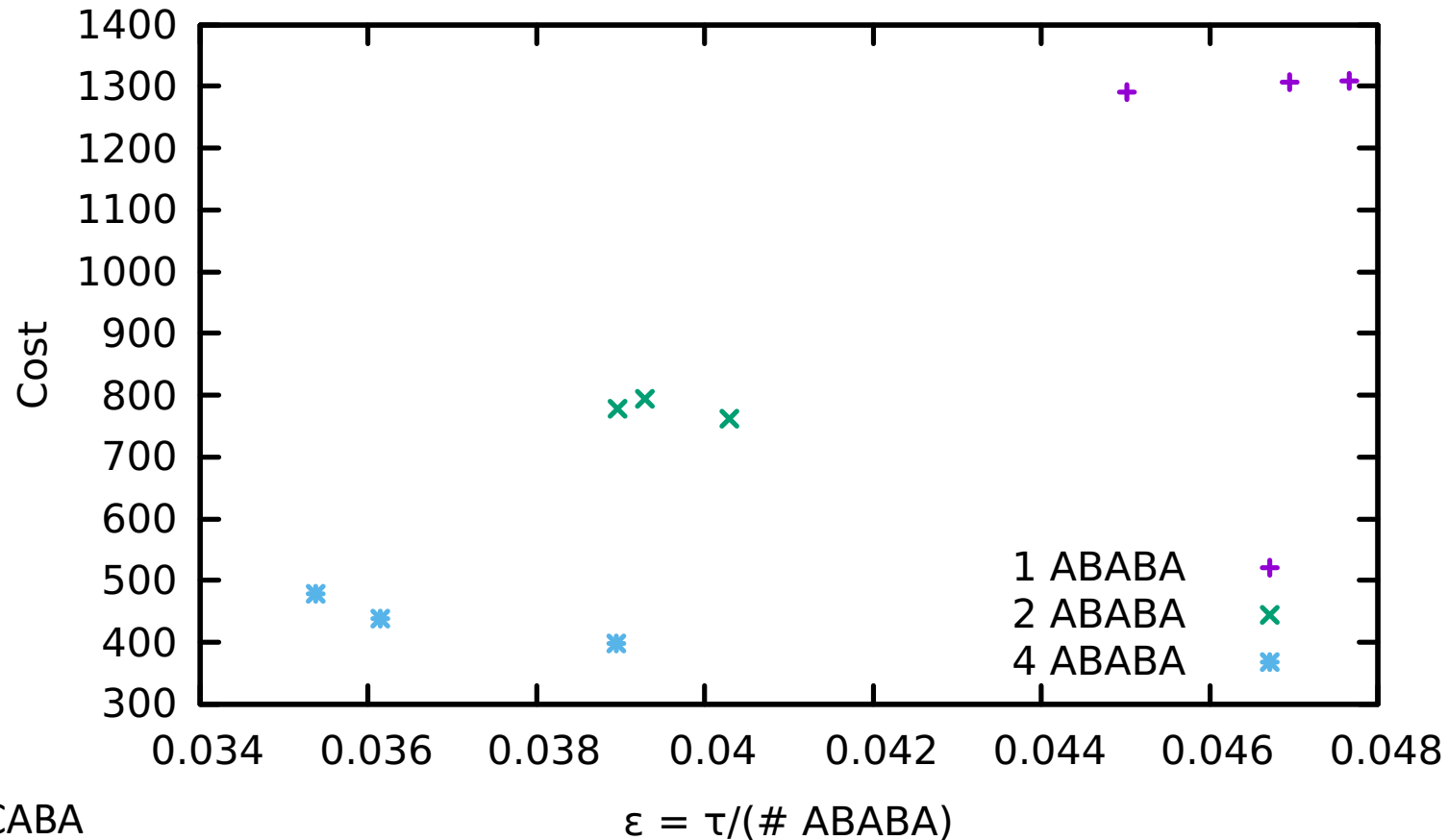
Staggered ABACABA: Cost vs trajectory length

Single integrator copy ($n=1, \varepsilon = \tau$)



Note: λ and χ have both gauge and fermion variants

Staggered ABABA: Multiple copies



Note 2 ABABA \approx 1 ABACABA
on small volume,
but only after tuning

Summary

- Tuning HMC parameters using gradient information seems to work well in simple cases
- Very convenient way to tune HMC, after initial investment in implementation
- Force gradient integrators may work much better when tuned, could be competitive even on small volumes
- Need to try with mass parameters in Hasenbusch ratios
- Plan to implement other actions (improved gauge, smearing)