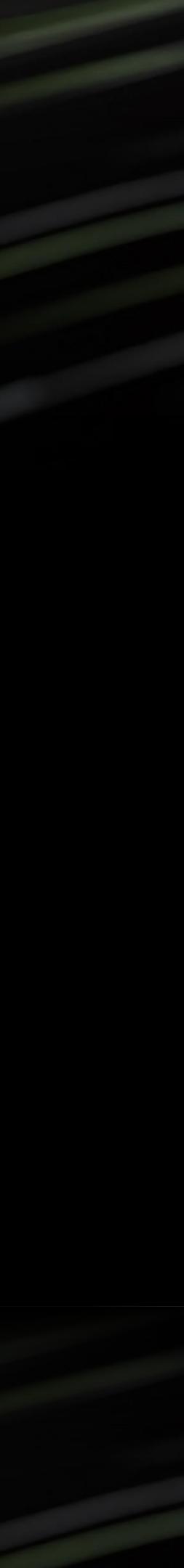


# HISQy Business

Evan Weinberg, Senior Developer Technology Compute Engineer, NVIDIA Lattice2023, July 31, 2023 (In collaboration with Venkitesh Ayyar, Richard Brower, Kate Clark)



# There's going to be a GPU raffle! The drawing will be during the Lattice 2024/2025 announcement on Friday.





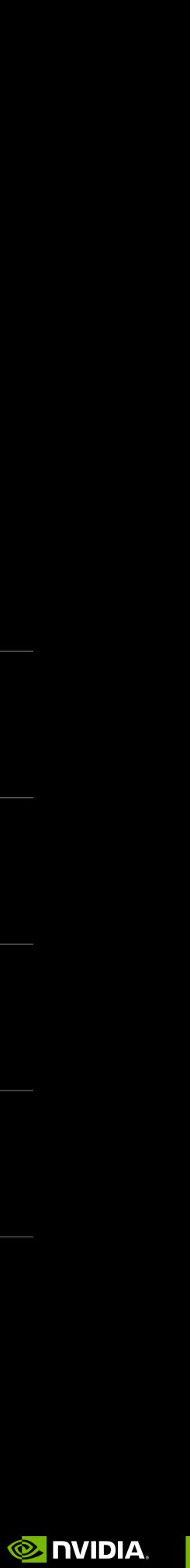
## Agenda

Takeaways & Challenges

- HISQ Crash Course
- **HISQ** Force
- **Future Work**



#### **HISQ Domain-Decomposed Preconditioning**

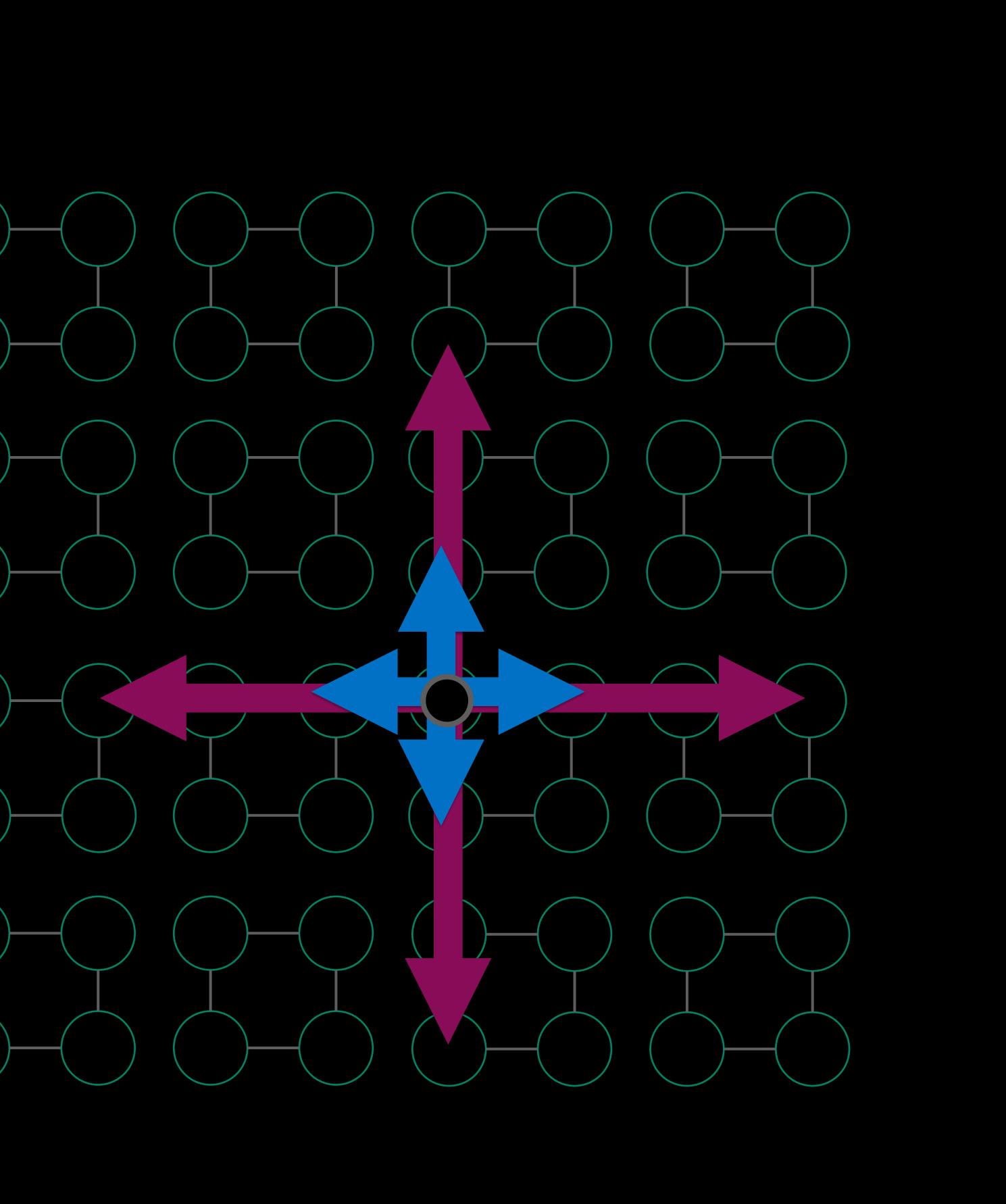


# Takeaways & Challenges



- HISQ: highly improved staggered quarks Smeared links: lots of locality to exploit
- New: hugely fused HISQ force implementation in QUDA
  - Merged: <u>https://github.com/lattice/quda/pull/1367</u>

### Takeaways Speeding up HISQ workflows



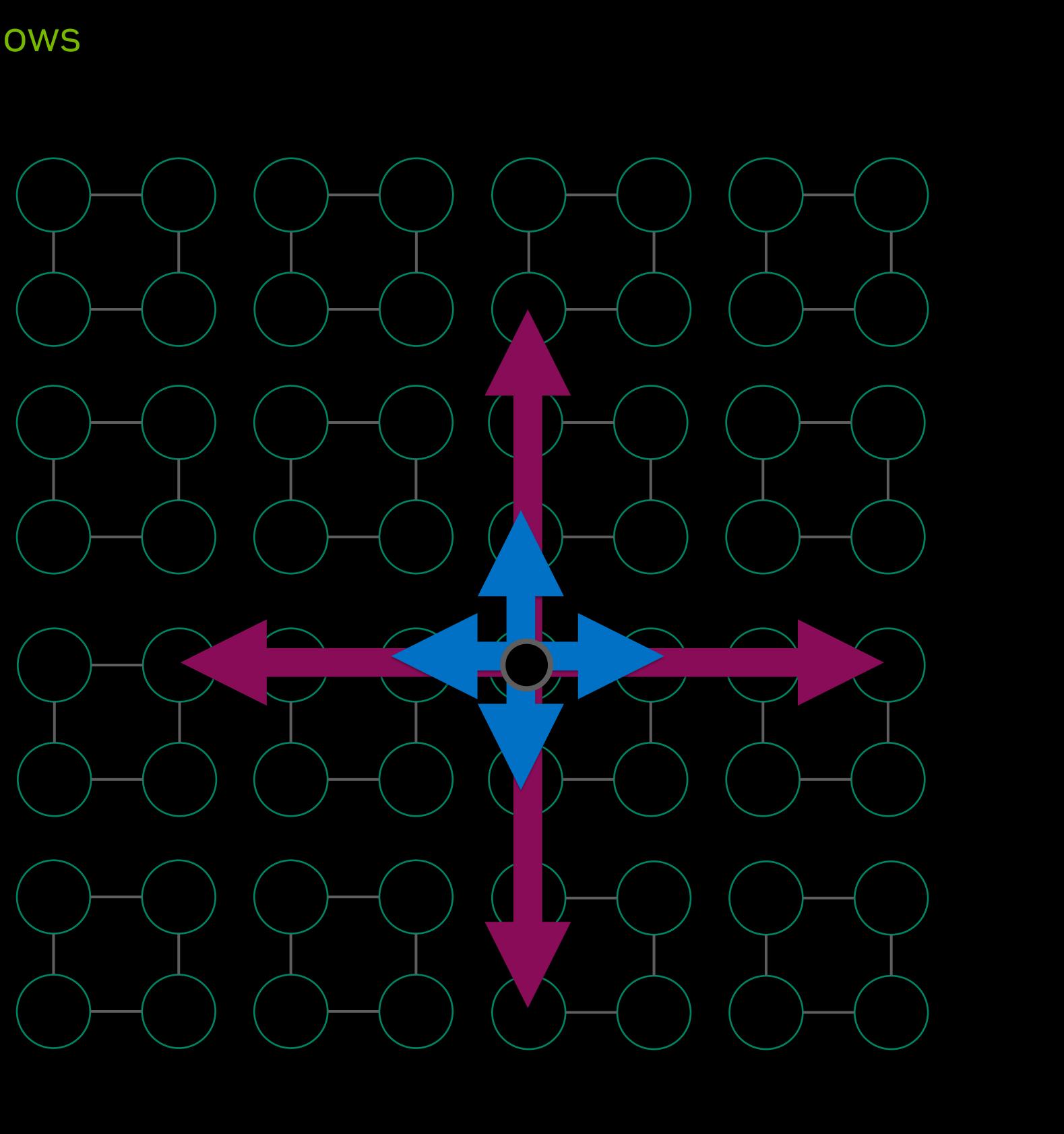


#### HISQ: highly improved staggered quarks Smeared links: lots of locality to exploit

- New: hugely fused HISQ force implementation in QUDA Merged: https://github.com/lattice/quda/pull/1367
- Modern machines have varying degrees of network performance Domain-decomposition algorithms become increasingly important HISQ's distance one and three terms introduce conceptual challenges

- New: (mostly-)optimized implementation of a local HISQ preconditioner in QUDA
  - successes and failures understood
  - We have demonstrated *numerical stability* And, in some cases, faster propagator solves---with performance
  - WIP branch, constantly in flux: https://github.com/lattice/quda/tree/feature/stag-invert-cleanup

### Takeaways Speeding up HISQ workflows



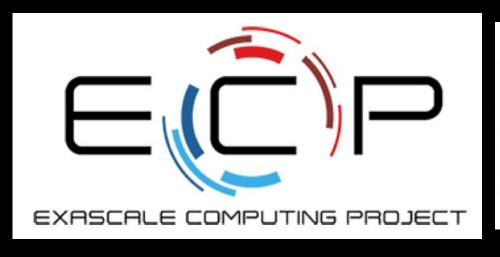


### **\*\*ECP benchmark applications**

- Not just CUDA anymore
- Chroma\*\*, CPS\*\*, MILC\*\*, TIFR, etc.
- Maximize performance
  - Mixed-precision methods

  - Multigrid solvers for optimal convergence
  - NVSHMEM for improving strong scaling







## QUDA

"QCD on CUDA" – <u>http://lattice.github.com/quda</u> (open source, BSD license)

• Effort started at Boston University in 2008, now in wide use as the GPU backend for BQCD,

Provides solvers for major fermionic discretizations, pure gauge algorithms, etc.

Autotuning for high performance on all architectures

Portable: HIP (merged), SYCL (in review) and OpenMP (in development)

 A research tool for how to reach the exascale (and beyond) Optimally mapping the problem to hierarchical processors and node topologies







<mark> NVIDIA</mark>.

LQCD simulations are particularly sensitive to network bandwidth

### Challenges Or: the state of the network



Not all HPC facilities prioritize network bandwidth

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• Regardless, it's not always possible (or practical) to control process placement • You can't always take advantage of all hierarchies of bandwidths/latencies



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- LQCD simulations are particularly sensitive to network bandwidth Not all HPC facilities prioritize network bandwidth
- Regardless, it's not always possible (or practical) to control process placement • You can't always take advantage of all hierarchies of bandwidths/latencies
- Communication reducing or avoiding algorithms are increasingly important for mitigating these challenges
- Our community has been and continues to be fully aware of this: Communication-reducing solvers
- - Domain-decomposed preconditioners
  - Domain-decomposed HMC

### Challenges Or: the state of the network



# HISQ Crash Course



#### Staggered fermions

- Spin-diagonalize the discrete Dirac matrix
- Phases  $\eta_{\mu}(x)$  preserve the Dirac structure

### Why Staggered Fermions? Aka Kogut-Susskind Fermions

$$\sum_{x,y}^{stag} \approx \sum_{\mu=0}^{3} \eta_{\mu}(x) \left[ U_{\mu}(x) \delta_{x,y-1} - U_{\mu}^{\dagger}(x-\hat{\mu}) \delta_{x,y+1} \right]$$

Lose shift-by-one translational invariance, but preserve a shift-by-two

 $+2m\delta_{x,y}$ 



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  - There's an exact chiral symmetry in contrast to Wilson/clover/twisted/etc There's no extra dimension in contrast to domain wall/Mobius/etc

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- In contrast to other discretizations...
  - There's an exact chiral symmetry in contrast to Wilson/clover/twisted/etc
  - There's no extra dimension in contrast to domain wall/Mobius/etc
- Like the continuum operator, it's just a symmetric first derivative: anti-Hermitian and normal

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#### $+2m\delta_{x,v}$



- Anti-Hermitian + normal:  $D_{eo} = -D_{oe}^{\dagger}$

### Why Staggered Fermions? Continued

Huge secondary benefit: the even/odd preconditioned operator is Hermitian Positive-Definite

$$\begin{bmatrix} 2m & D_{eo} \\ D_{oe} & 2m \end{bmatrix} \begin{bmatrix} x_e \\ x_o \end{bmatrix} = \begin{bmatrix} b_e \\ b_o \end{bmatrix}$$

 $(4m^2 - D_{eo}D_{oe})x_e = 2mb_e - D_{eo}b_o$ 





- Anti-Hermitian + normal:  $D_{eo} = -D_{oe}^{\dagger}$

- Obviously, there's no free lunch

### Why Staggered Fermions? Continued

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$$\begin{bmatrix} 2m & D_{eo} \\ D_{oe} & 2m \end{bmatrix} \begin{bmatrix} x_e \\ x_o \end{bmatrix} = \begin{bmatrix} b_e \\ b_o \end{bmatrix}$$

## $(4m^2 - D_{eo}D_{oe})x_e = 2mb_e - D_{eo}b_o$

• There is a residual doubling:  $2^{d/2}$  doublers (as opposed to  $2^d$ ) • "Taste-breaking" effects: only one of the "pions" feels the exact lattice chiral symmetry





#### HISQ takes staggered fermions and addresses the issues:

- Smooths the fields
- Suppresses taste-breaking effects
- Additionally performs Symanzik improvement

### Enter HISQ "Highly Improved Staggered Quarks"



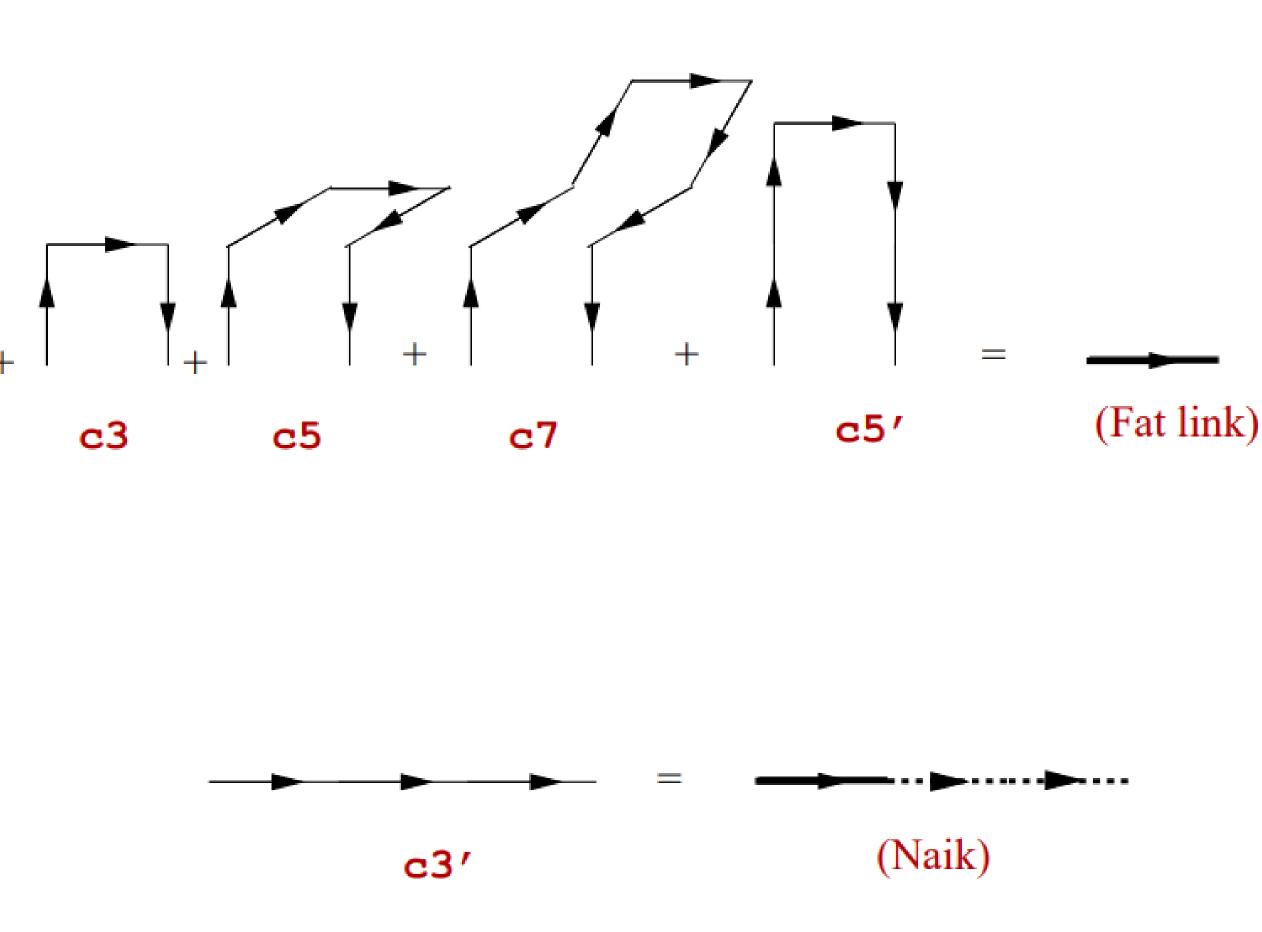
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  - "fat7" + Lepage term to suppress taste-breaking

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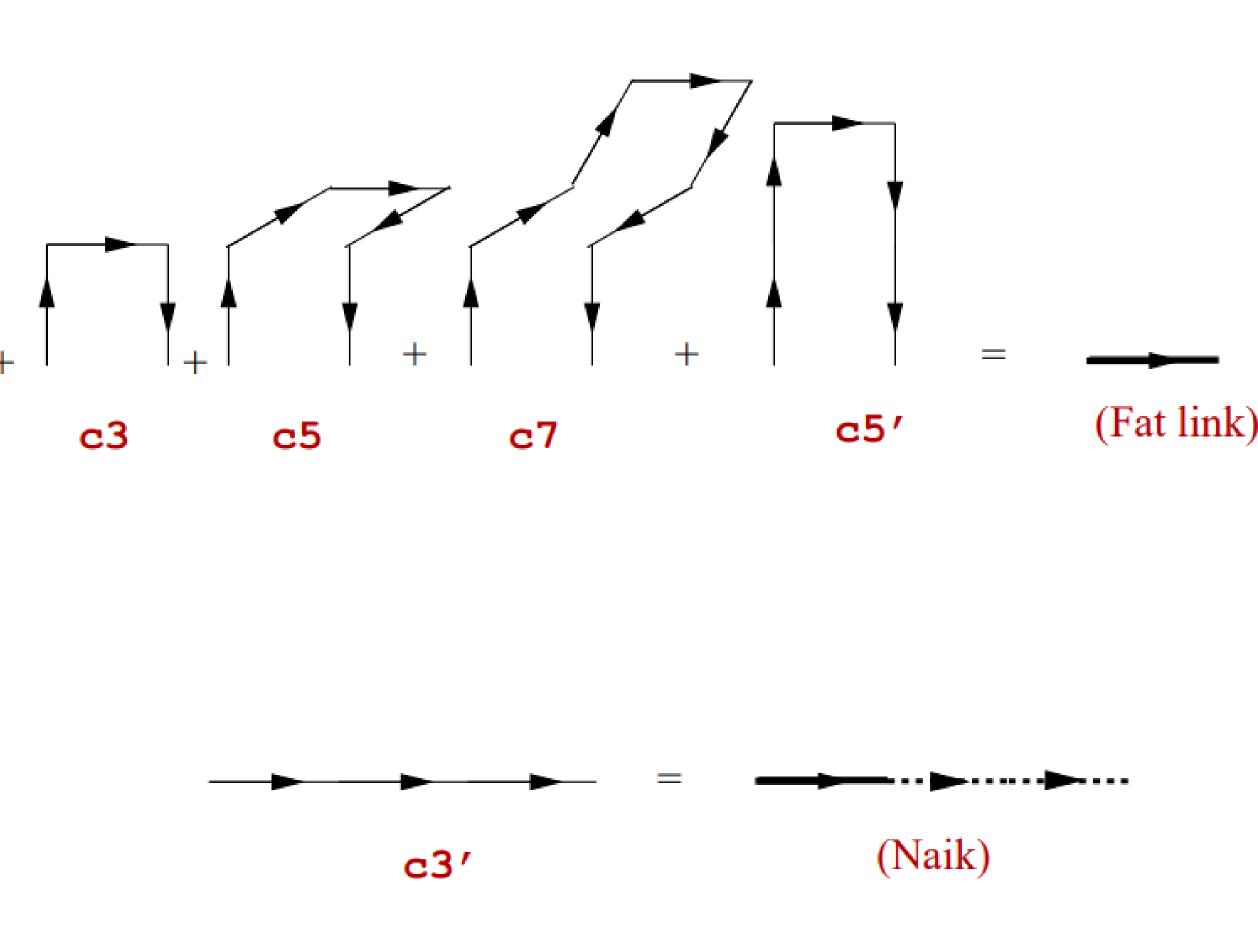
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Equations can be re-written to remove Lepage term from first step









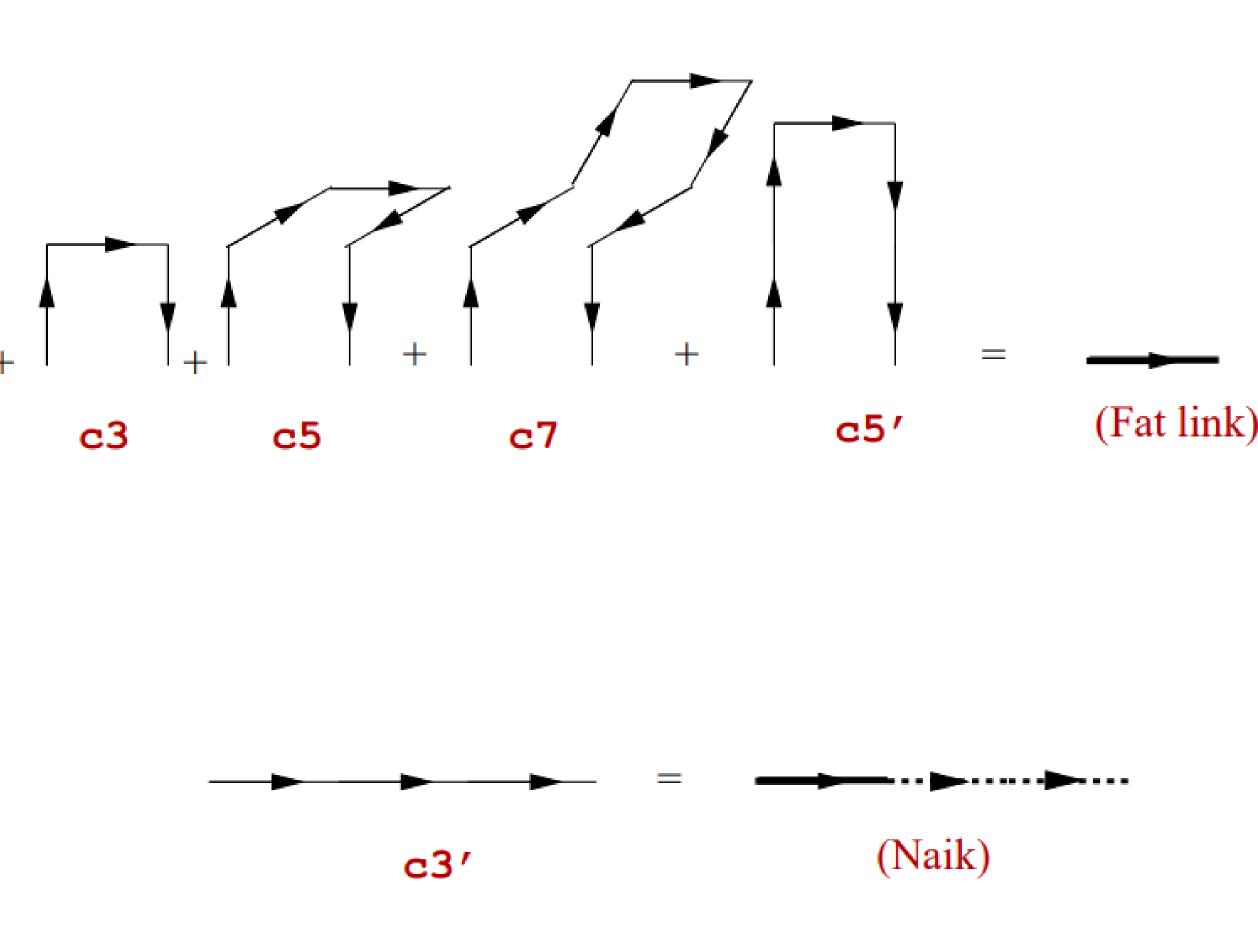
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- Full workflow: ASQTAD + re-unitarization + ASQTAD
- Addition of "long links" for Symanzik improvement

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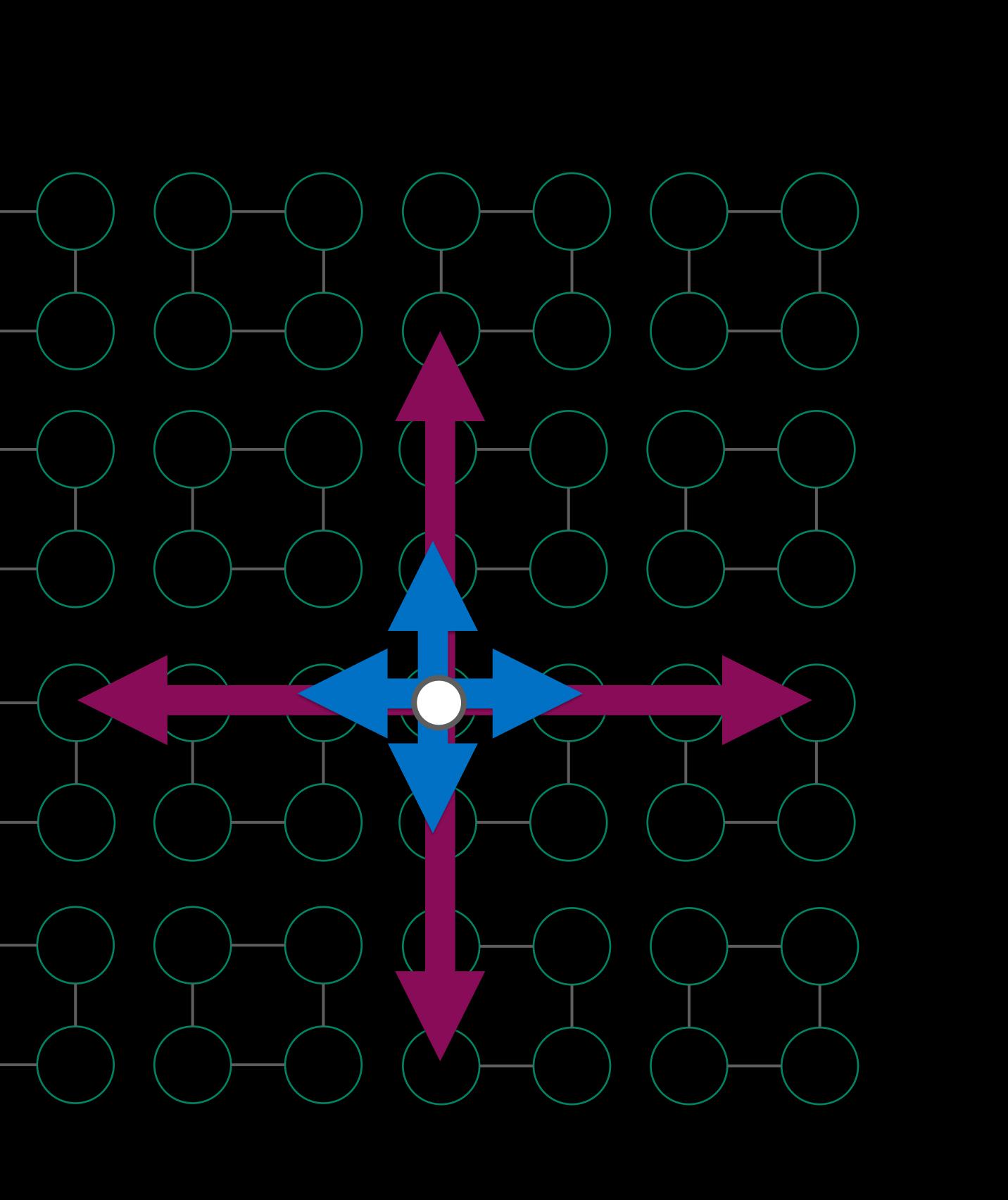






- The final (massive) HISQ stencil is a 17-point stencil
- One local mass term
- Eight distance-1 "fat link" terms: "general" Nc x Nc matrices
- Eight distance-3 "long link" terms: U(Nc) matrices

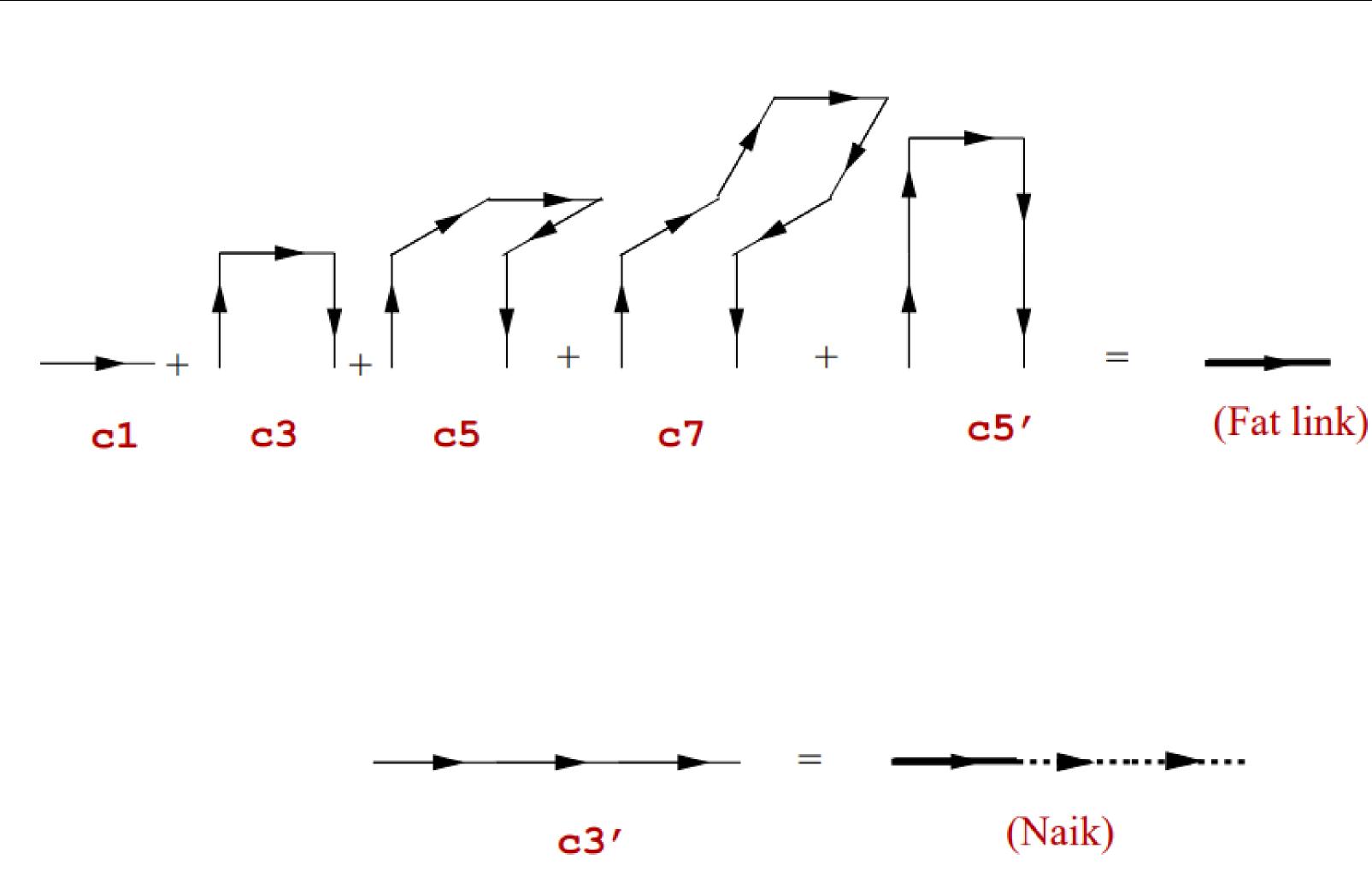
### The HISQ Stencil 17 points for Lattice Gryffindor





- Constructing the fat links is inherently recursive
- 3-link terms are built from 1-link terms
- 5-link terms can be built from 3-link terms
  - As can the Lepage (c5') staple
- 7-link terms can be built from 5-link terms

## **Recursive Link Fattening**







In one kernel: accumulate c1 + c3, store c3 term

# Data Reuse



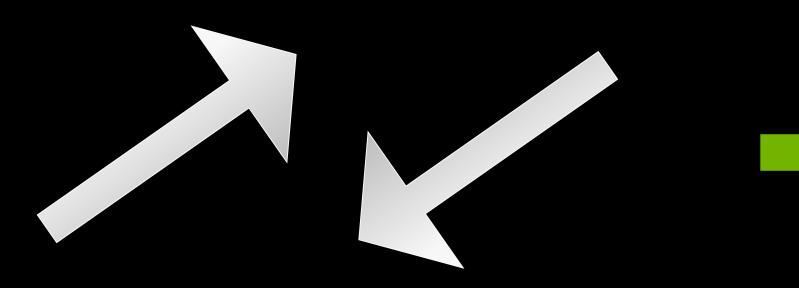
Save each length three staple



**NVIDIA** 

In one kernel: accumulate c1 + c3, store c3 term

• In the next kernel: load gauge links, load two staples, construct five-link terms, accumulate c5s into force, save fivelink terms

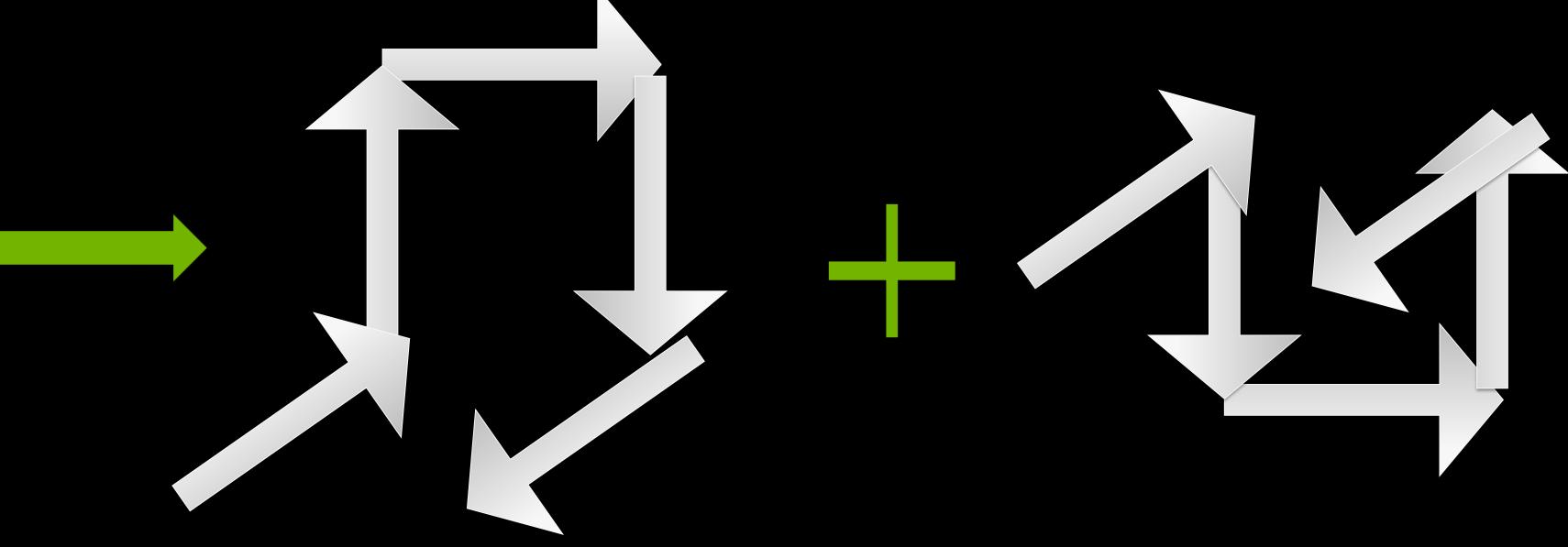


Load these two links

# Data Reuse



Save each length three staple



Save each five-link staple separately

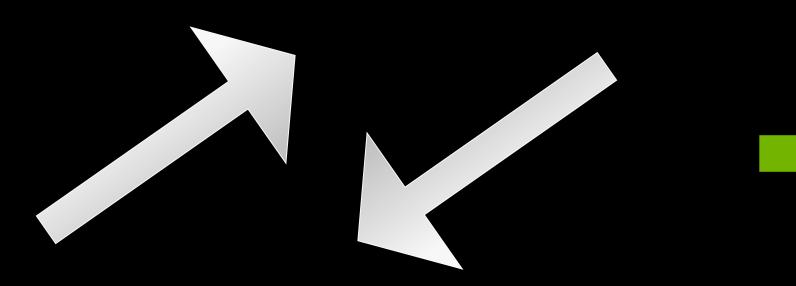
Increment five-link staples into accumulator



<mark> NVIDIA</mark>.

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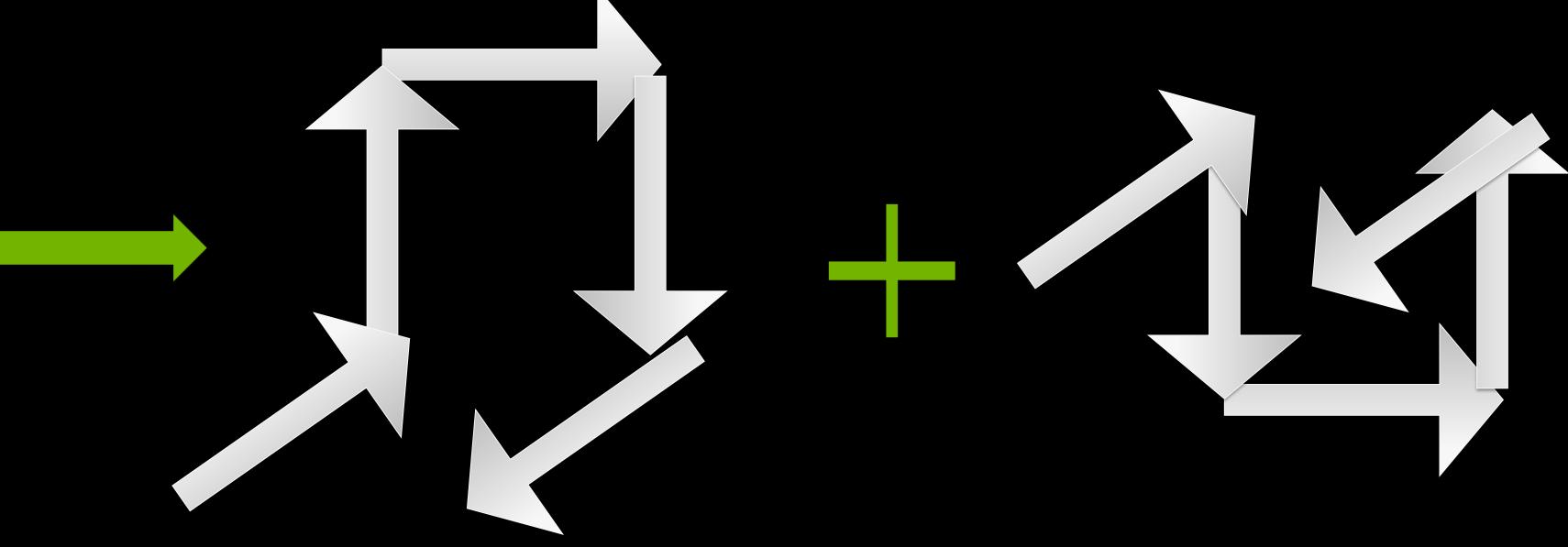
Load these two links

• ...etc

# Data Reuse



Save each length three staple



Save each five-link staple separately

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<mark> NVIDIA</mark>.

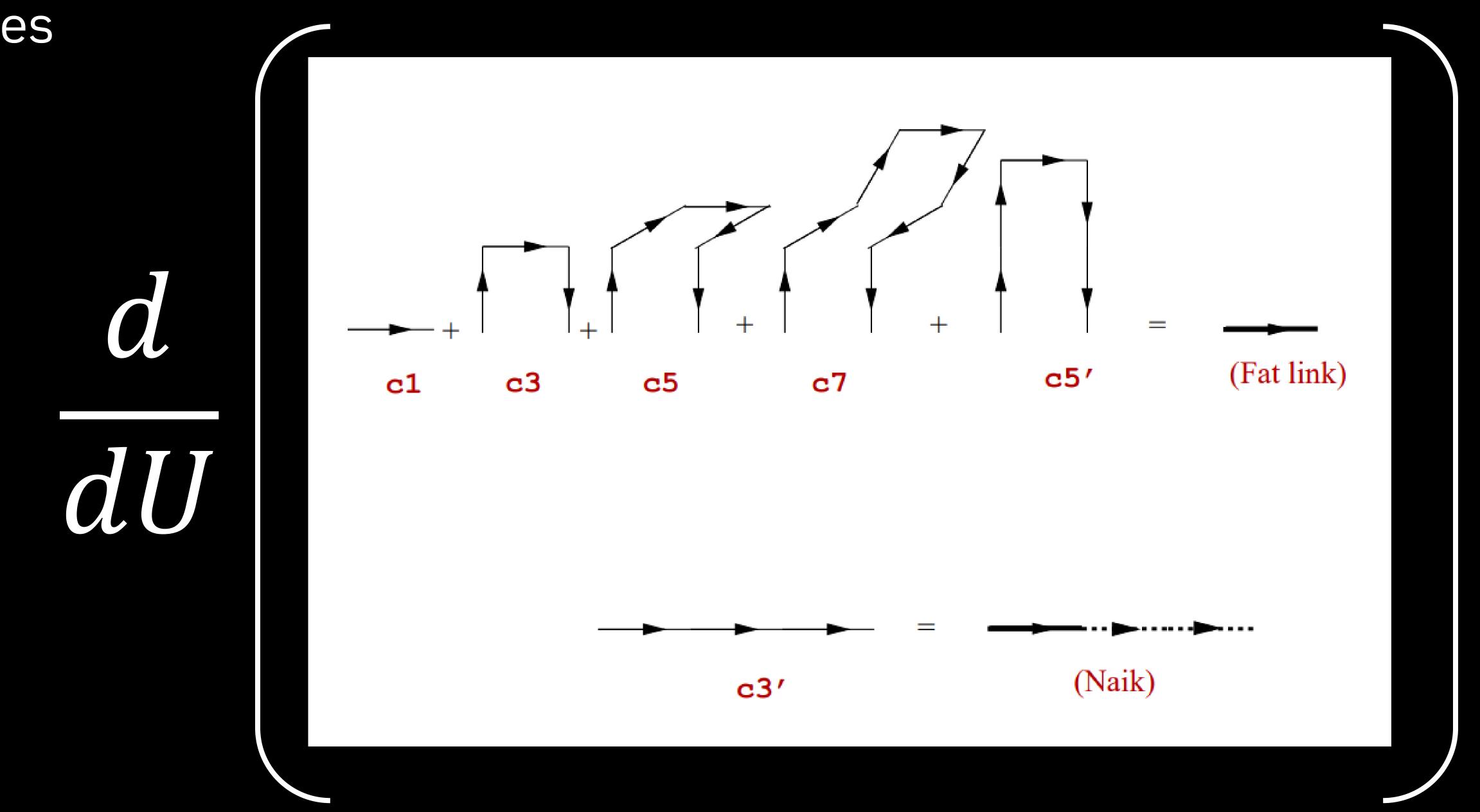
# HISQ Force



- - Reuse of intermediates
  - Kernel fusion
  - Cache Reuse

### HISQ Force "Highly Improved Staggered Quarks"

## The HISQ force is a beast: three-stage chain rule Similar to the fat link construction, there are a lot of opportunities for...

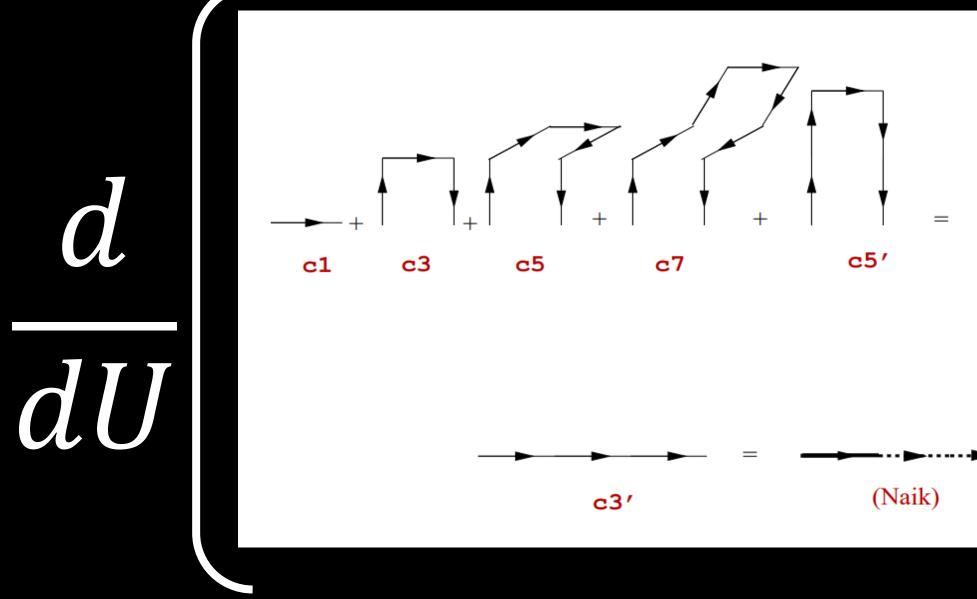






 Original implementation: Loop over sig =  $\{x, y, z, t\}$ ; forward/backward

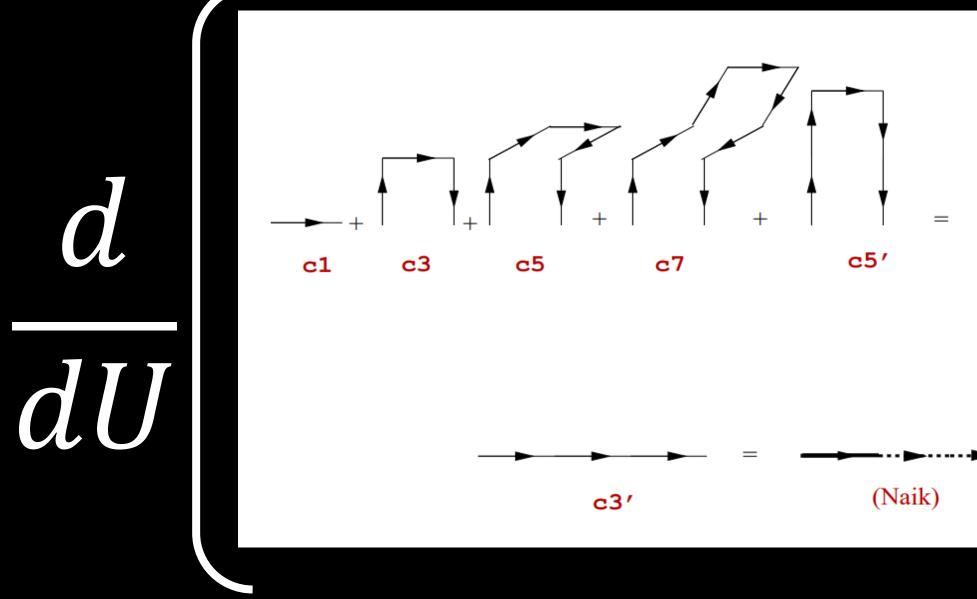
End loop (sig)





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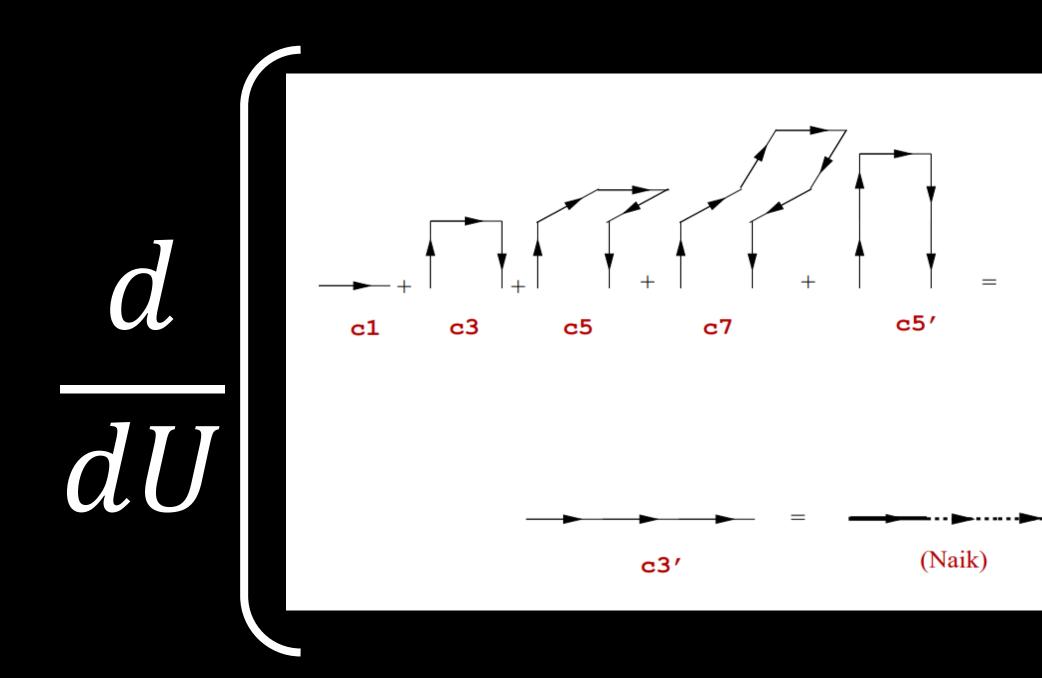
End loop (mu) End loop (sig)

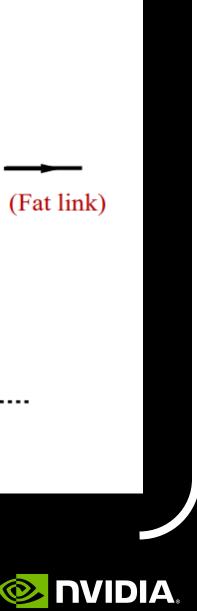




 Original implementation: Loop over sig =  $\{x, y, z, t\}$ ; forward/backward Loop over mu != |sig|; forward/backward Compute sig, mu 3-link middle force: Accumulate and store intermediates

End loop (mu) End loop (sig)





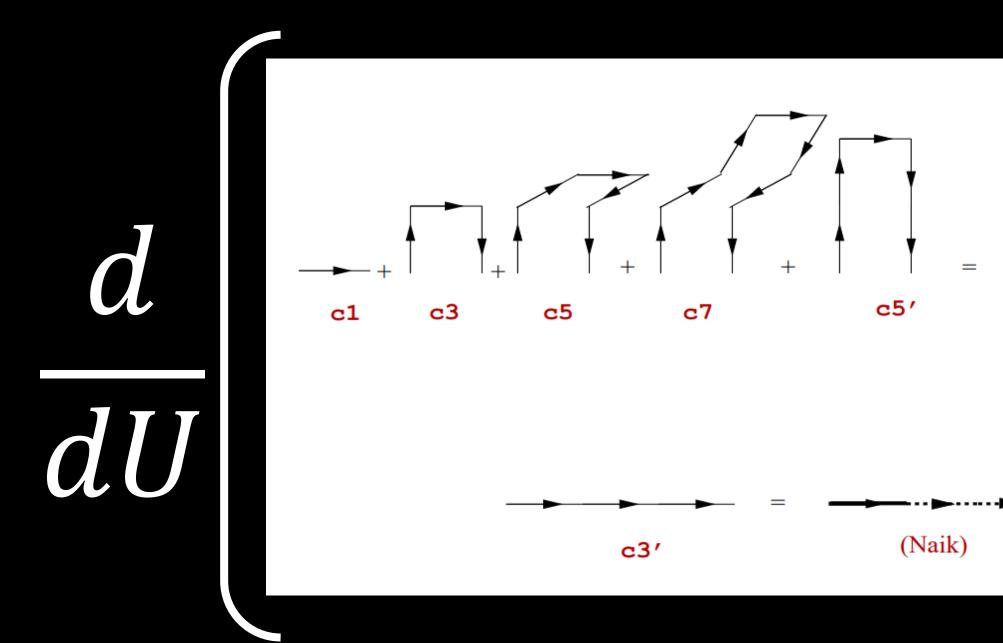
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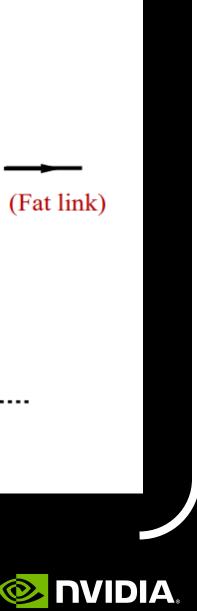
End loop (nu)

End loop (mu) End loop (sig)

### **HISQ Force** Sorry about the pseudocode

Compute sig, mu, nu 5-link middle force: reuse intermediates from before



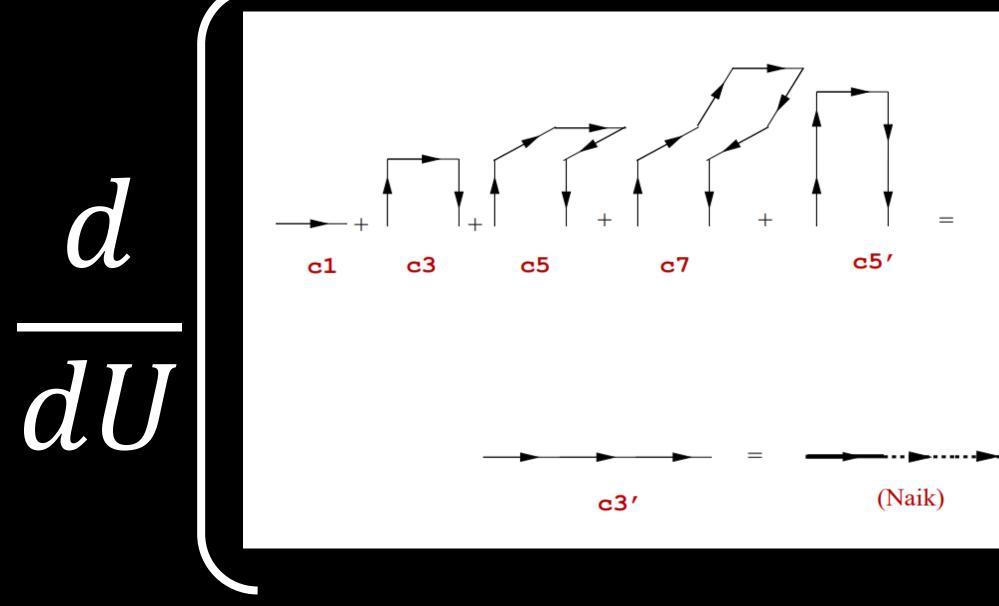


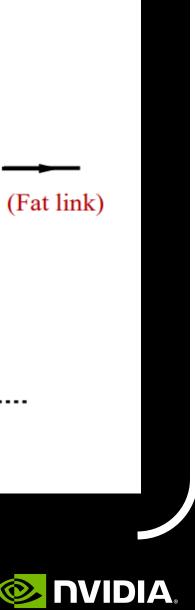
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- Loop over rho != |sig|, |mu|, |nu|, forward/backward
  - Compute sig, mu, nu, rho 7-link middle force, side force

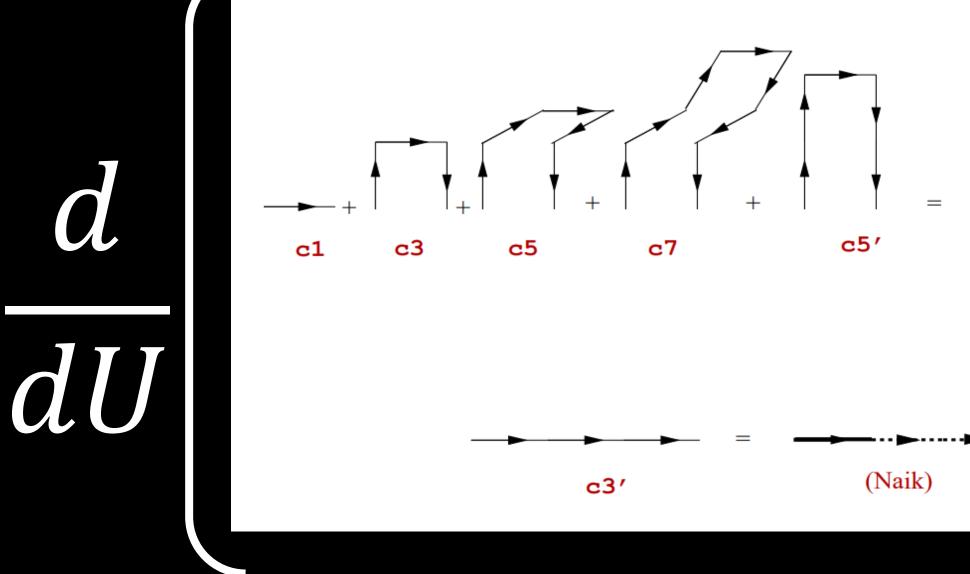


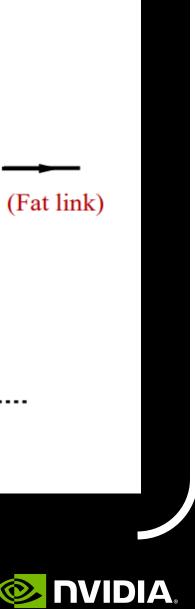


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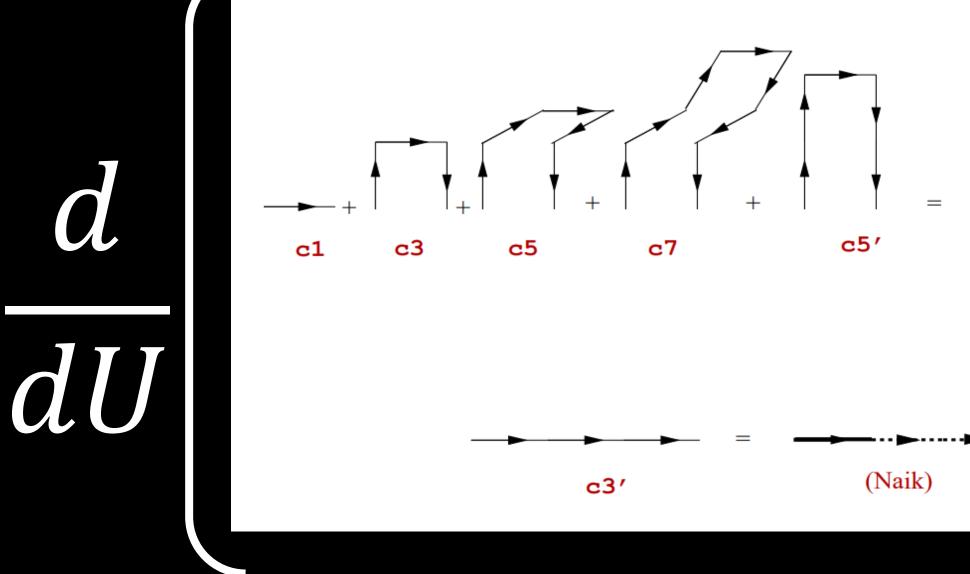


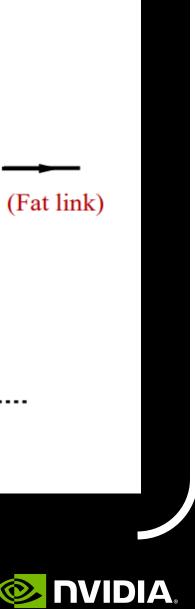


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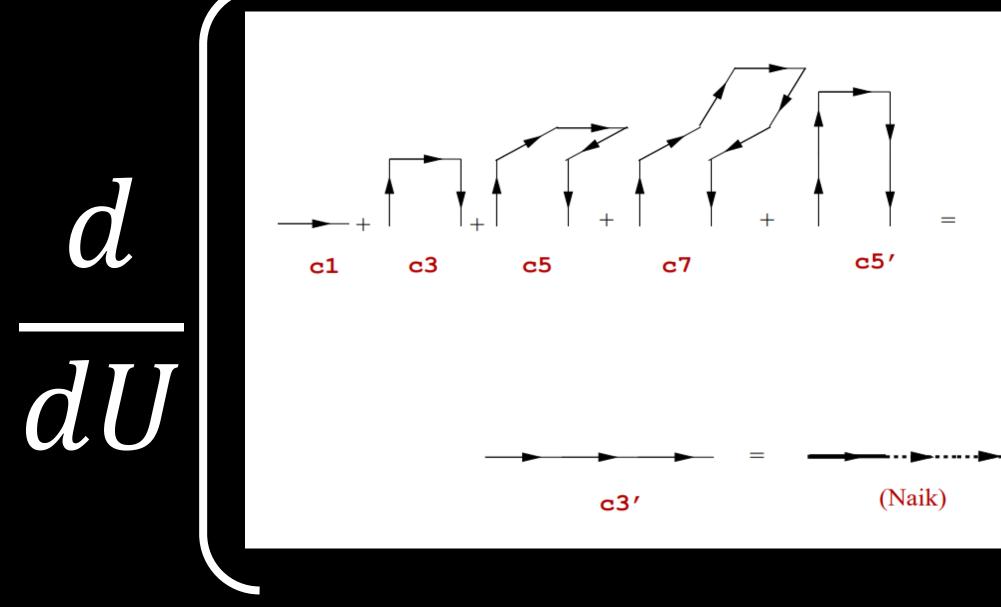


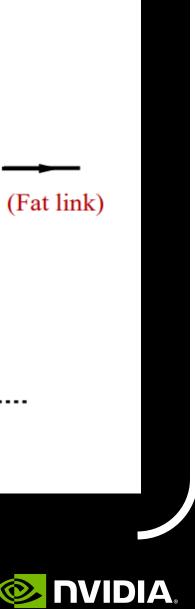


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- Loop over rho != |sig|, |mu|, |nu|, forward/backward Compute sig, mu, nu, rho 7-link middle force, side force
- Compute sig, mu, nu 5-link side force... + next middle force

 $\mathcal{O}$ dU**c**3'

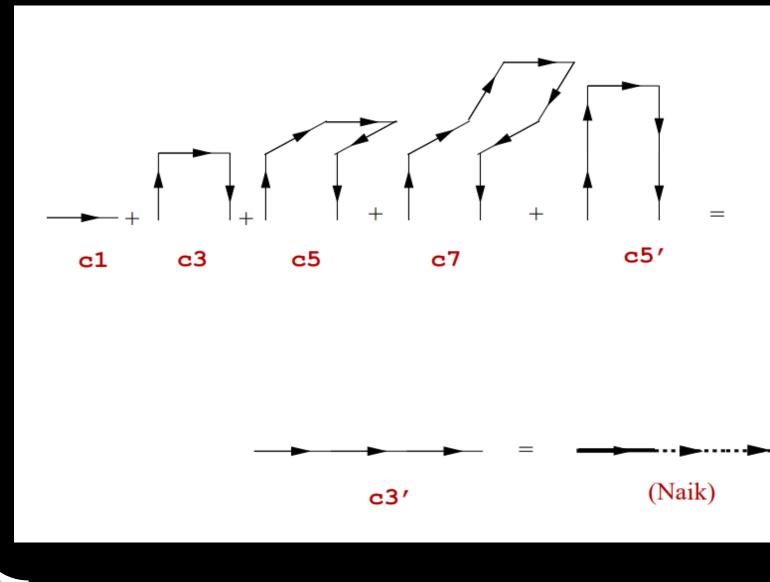
(Naik)

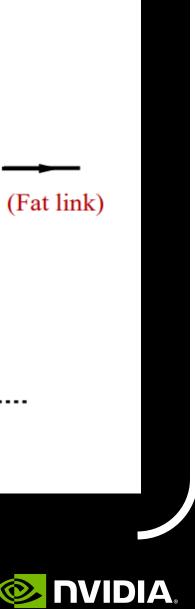


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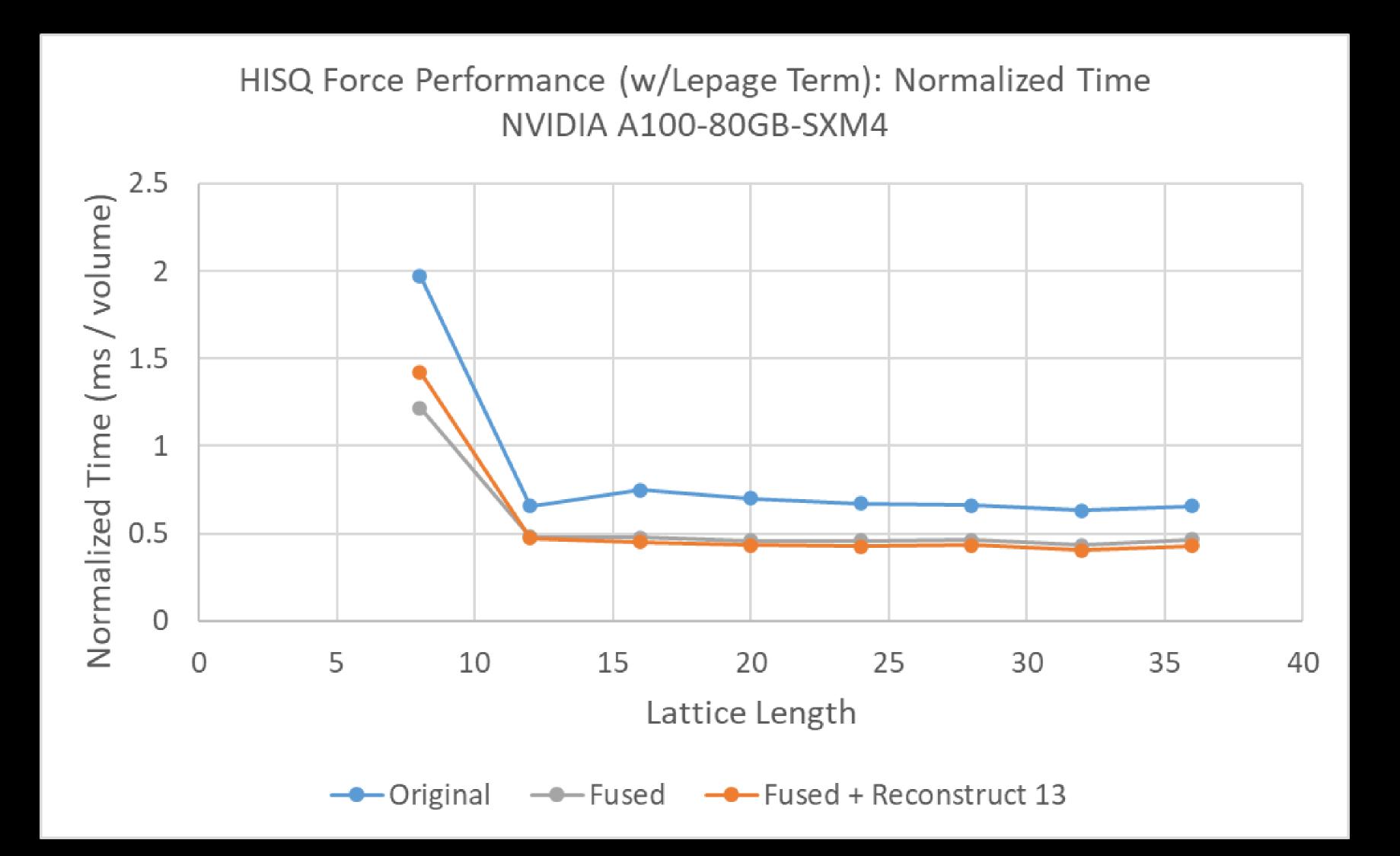


### Use Your Symmetries

While the staples are general matrices, all base gauge links are U(3)

• Take advantage of this symmetry to reduce memory traffic: store as 13 reals, recompute as needed



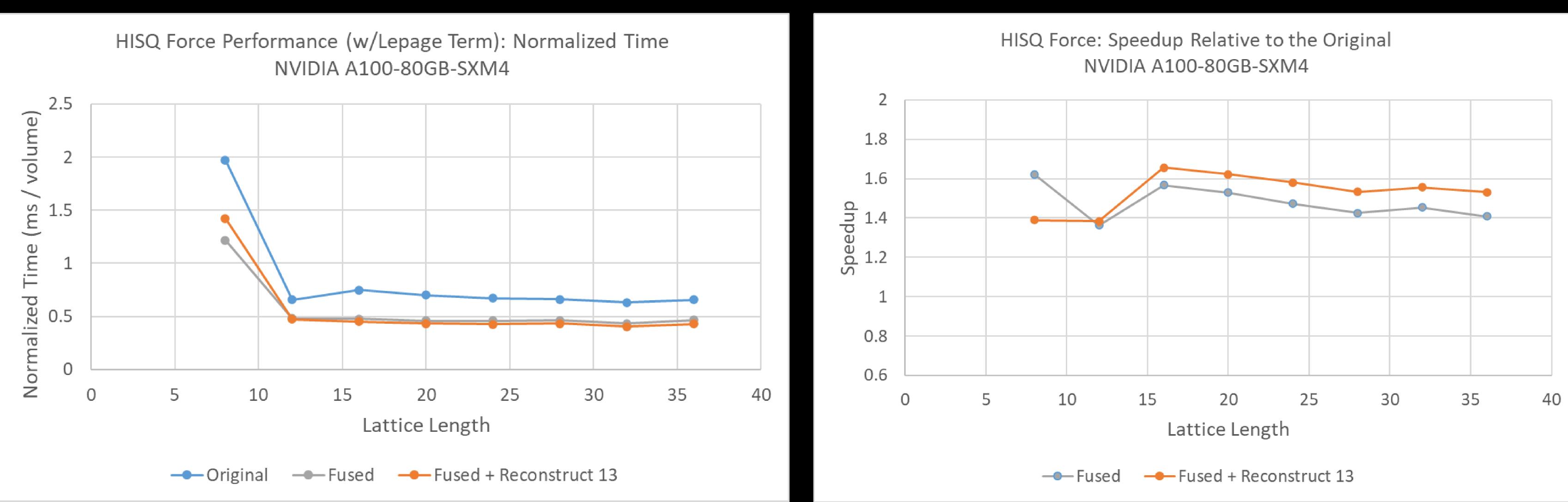


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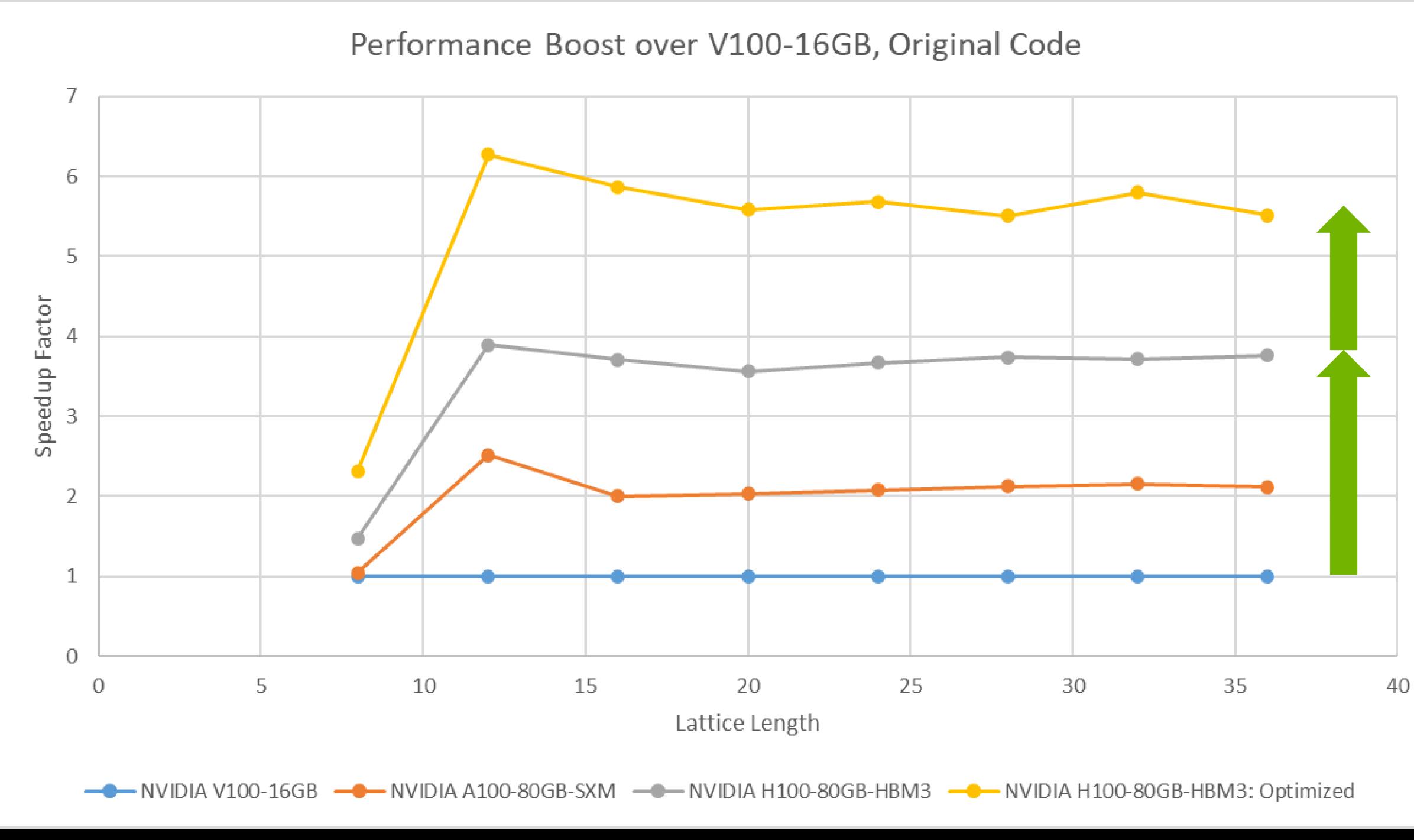


### **Use Your Symmetries**

### • While the staples are general matrices, all base gauge links are U(3) • Take advantage of this symmetry to reduce memory traffic: store as 13 reals, recompute as needed



### Improvements are algorithmic and architectural



### Architecture and algorithm boosts multiply: ~5.6x

# Algorithm: 1.5x

### Architecture: 3.75x



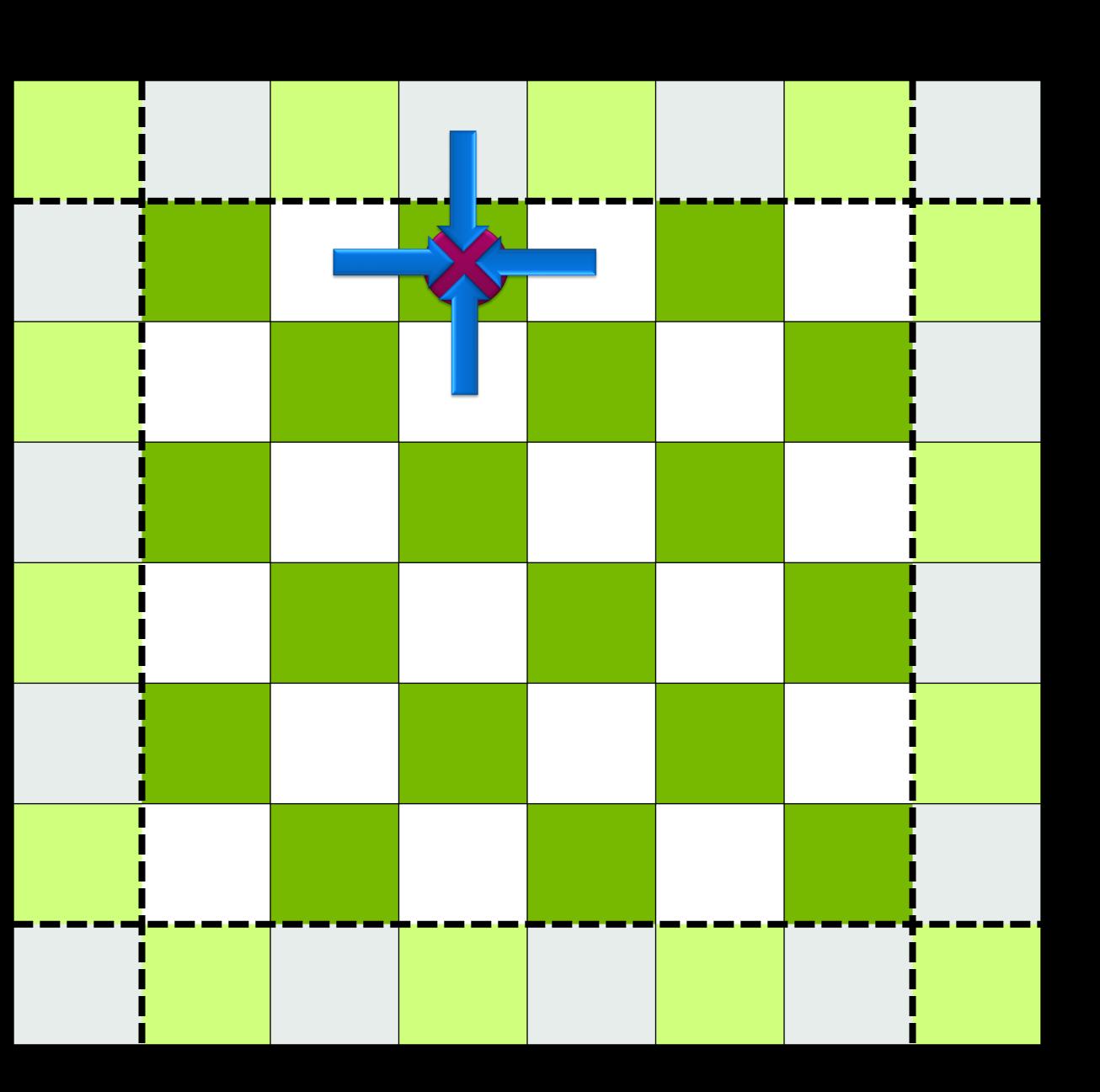
## HISQ Domain-Decomposed Preconditioning



- Simple idea: expand the idea of site-local preconditioning...

  - Example B: 4-d preconditioning of Mobius fermions

Preconditioning (twisted-)clover with the (twisted-)clover inverse

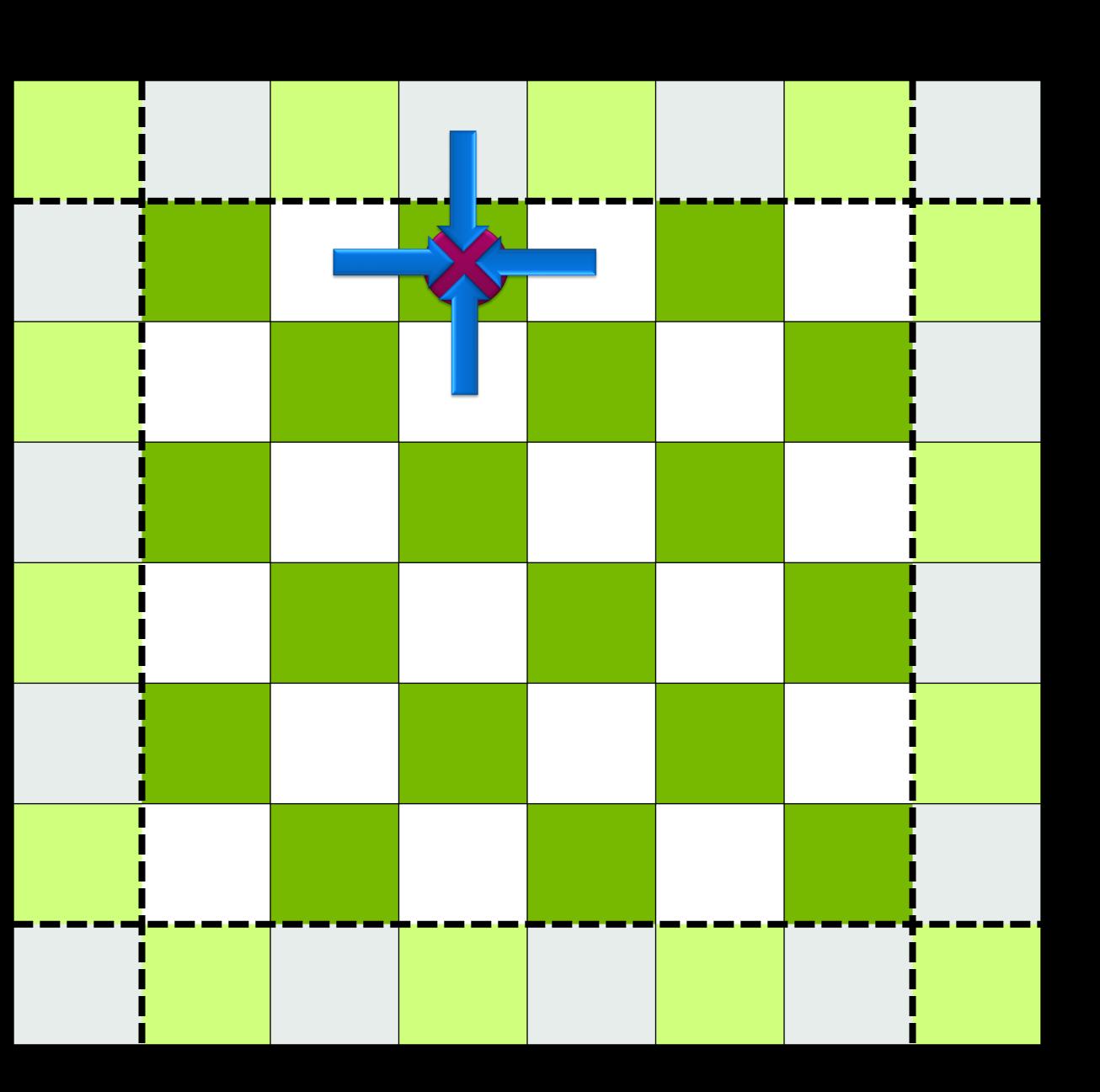




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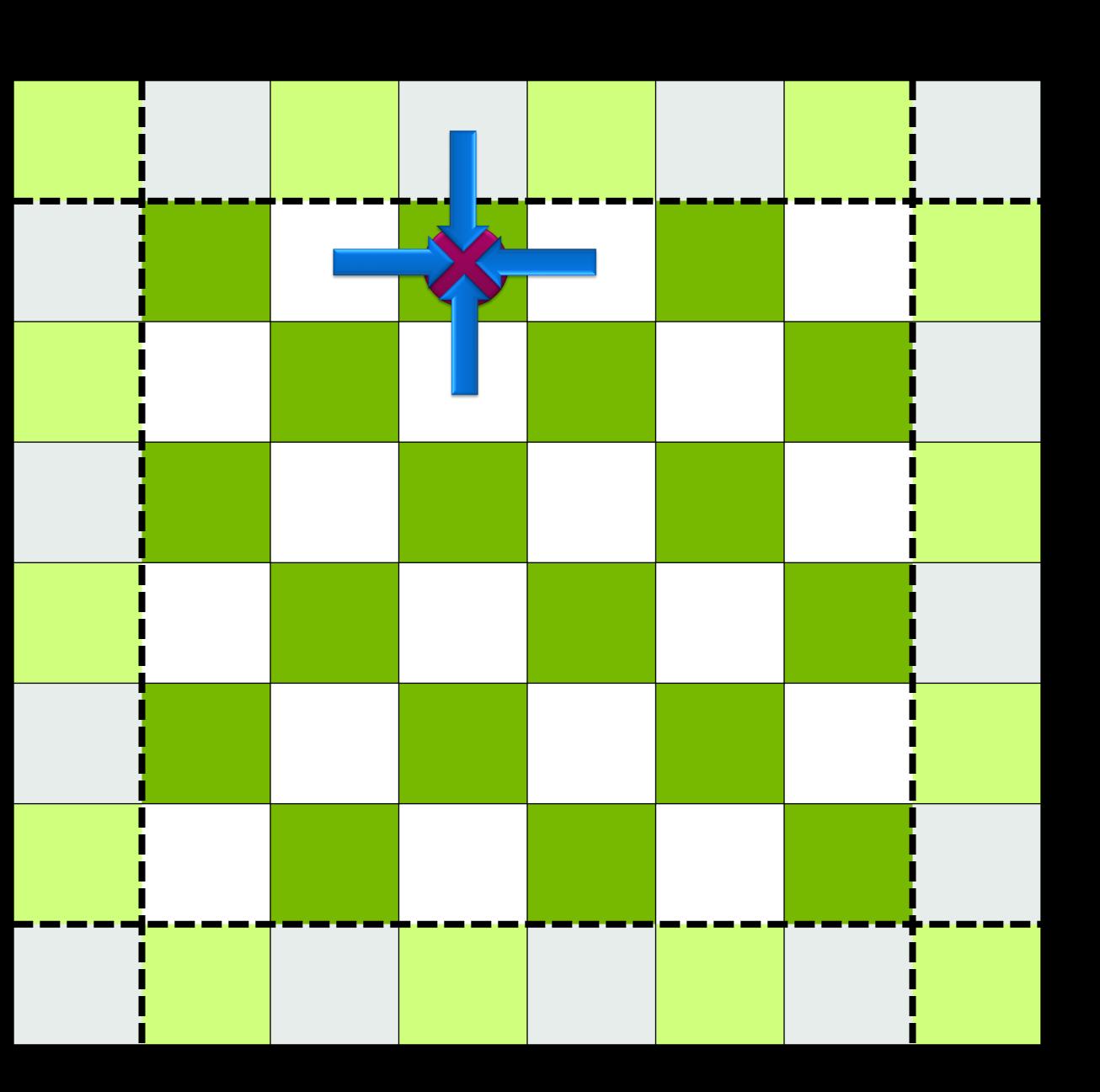
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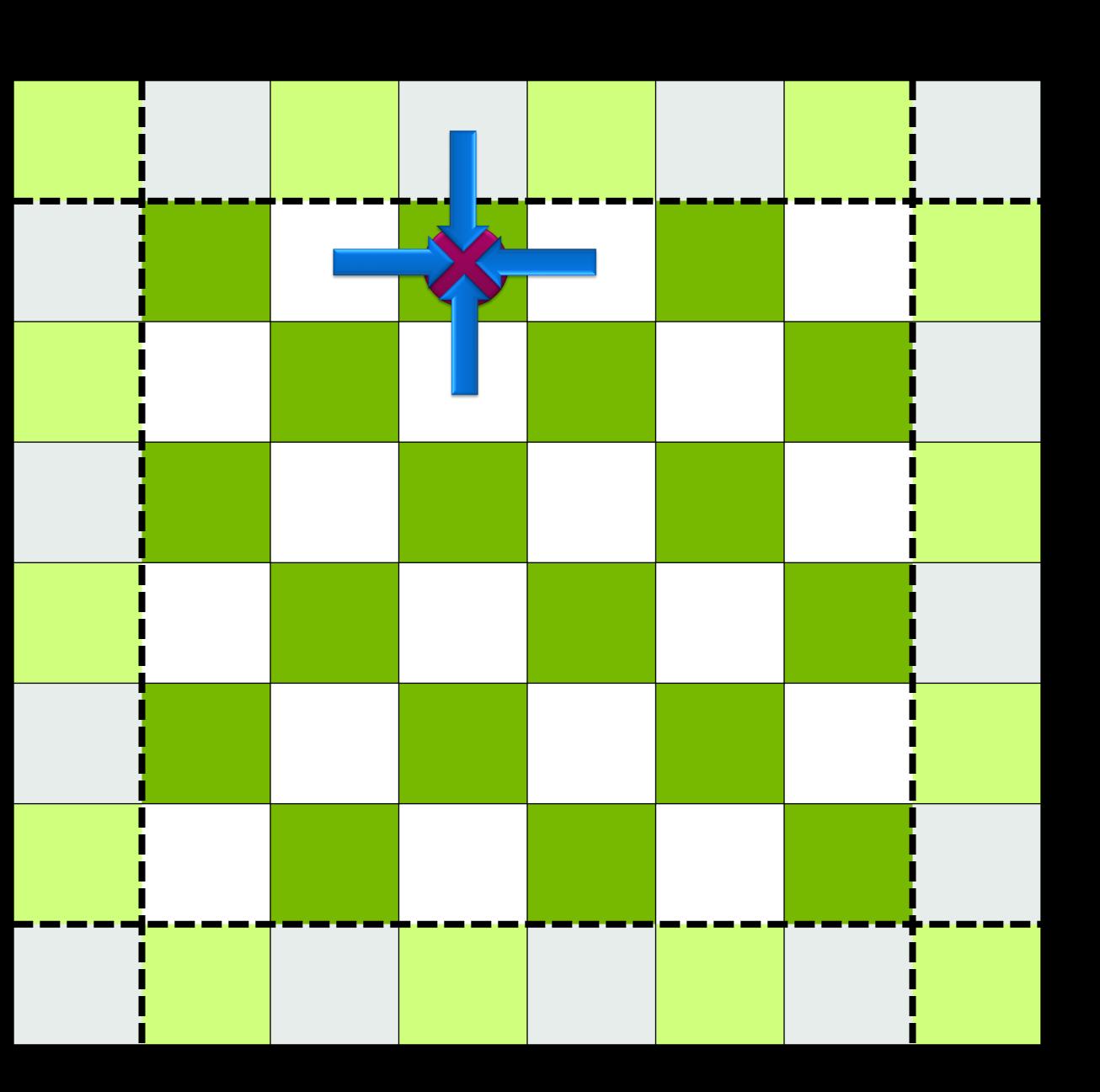
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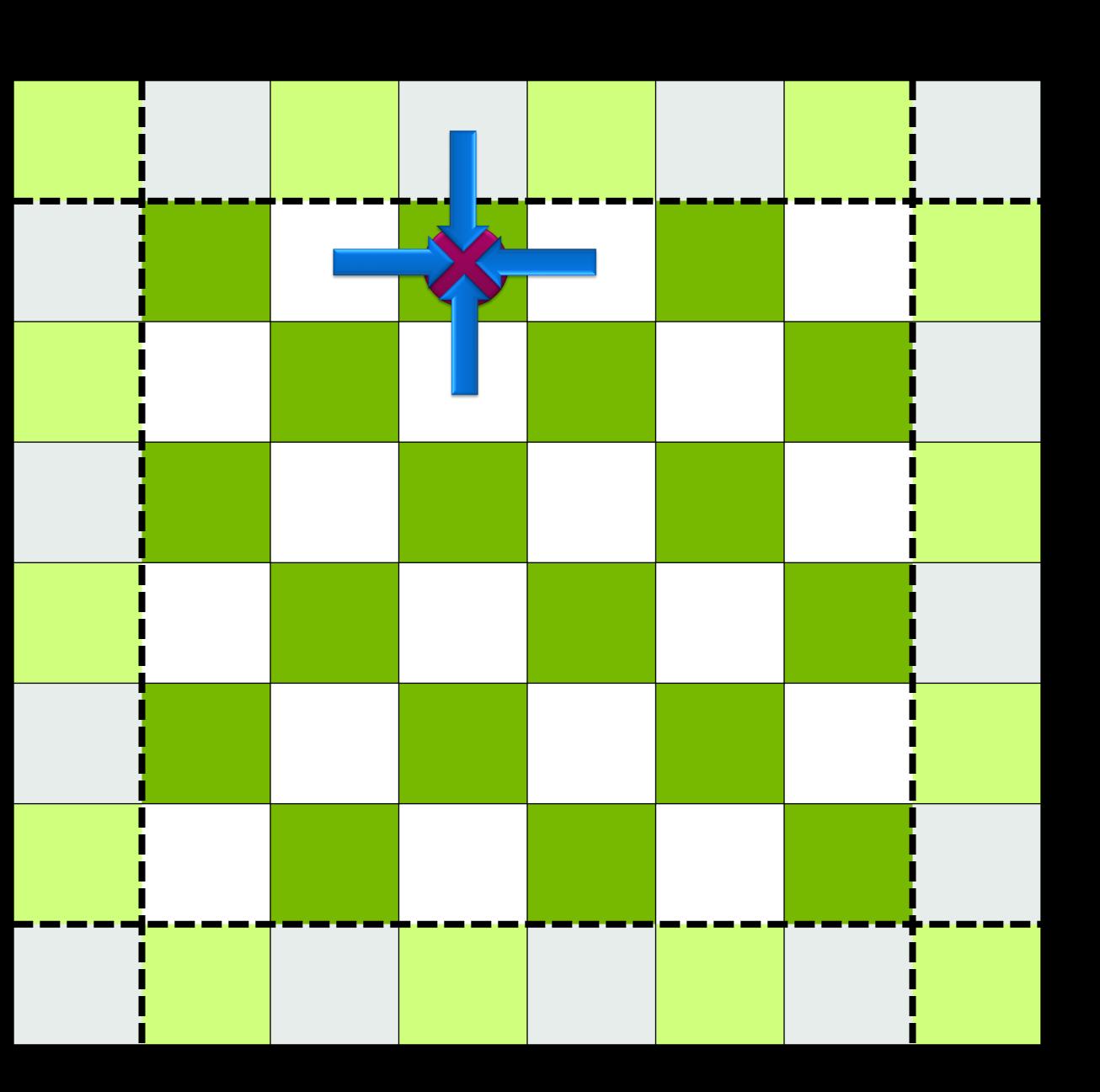
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  - Example B: 4-d preconditioning of Mobius fermions
- ...to larger domains: Schwarz preconditioning
- Additive Schwarz is analogous to Jacobi Iterations, but for domains
- For this talk: domains are non-overlapping
- Here: one domain per MPI rank (== one GPU)
  - This is a person-hour coding and debugging constraint
  - There's no inherent algorithmic or machine constraint





- The theory and use of Schwarz preconditioners is long-lived and exhaustive----the idea isn't anything new-fangled
- The challenge is constructing the algorithm and the implementation





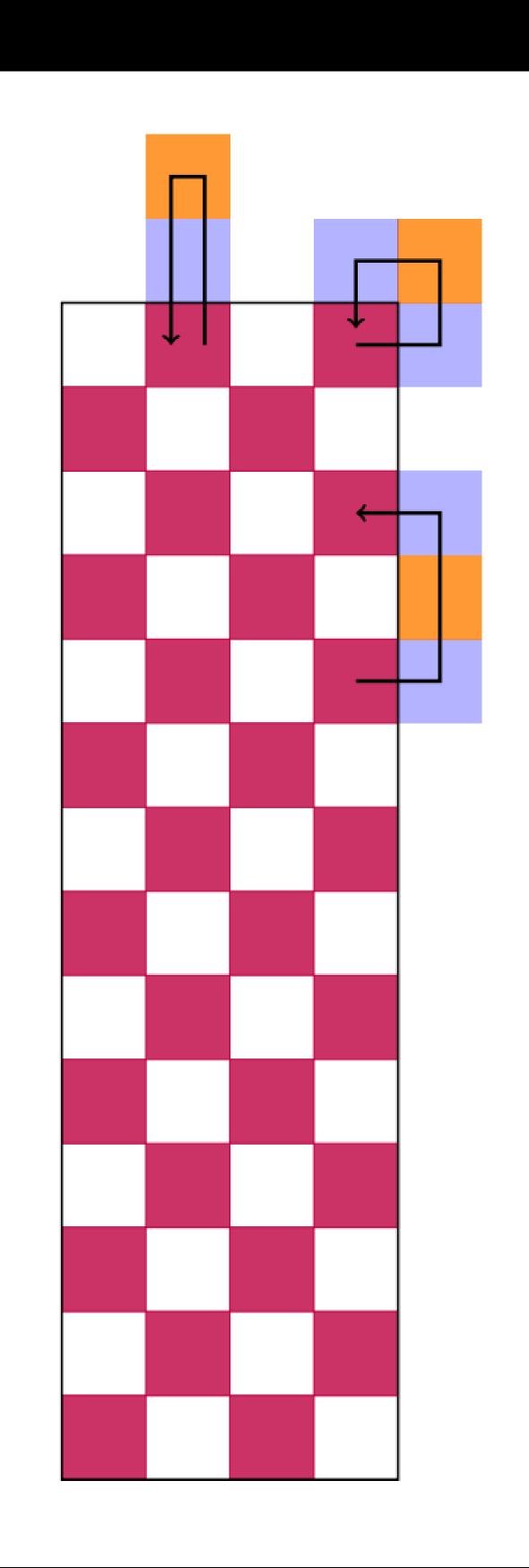
- The theory and use of Schwarz preconditioners is long-lived and exhaustive----the idea isn't anything new-fangled
- The challenge is constructing the algorithm and the implementation
- A recent example in LQCD is Multi-Splitting Preconditioned Conjugate Gradient (MSPCG)
  - [arxiv:2104.05615]
- For Mobius fermions, the relevant HPC operator is the normal 4-d preconditioned operator

$$(1 - D_{eo}D_{oe})$$

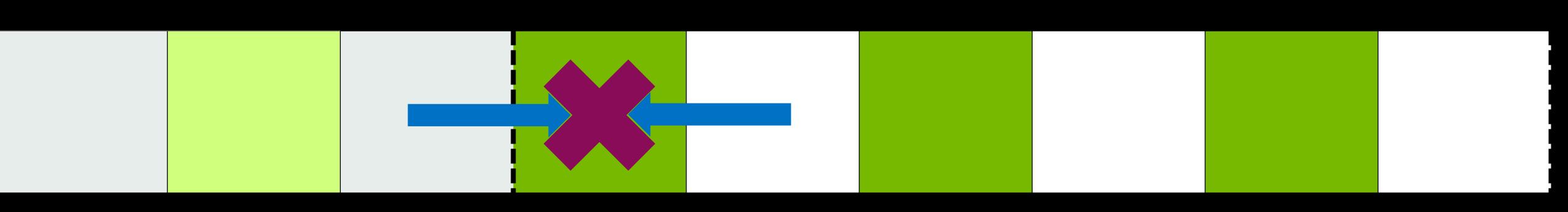
The product of four Ds generates so-called snake terms

### **Existing Work** Mobius Fermions

## )<sup>†</sup>(1 – $D_{eo}D_{oe}$ )







Exterior domain

### **Zero Boundaries**

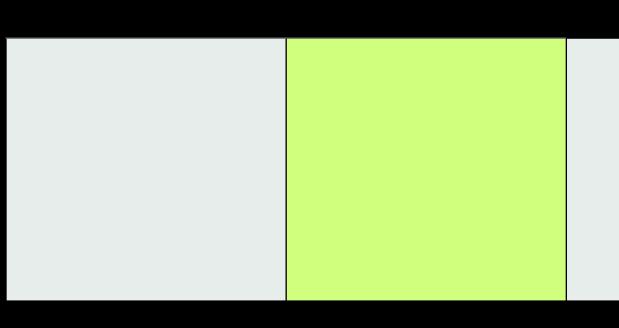
"Boundary clovers"

 Let's consider the massless staggered operator... in one dimension, for extreme simplicity  $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger}(x-\hat{\mu}) \delta_{x,y+1} \right]$ 

The stencil gathers from two sites: one on the left, and one on the right

Interior domain





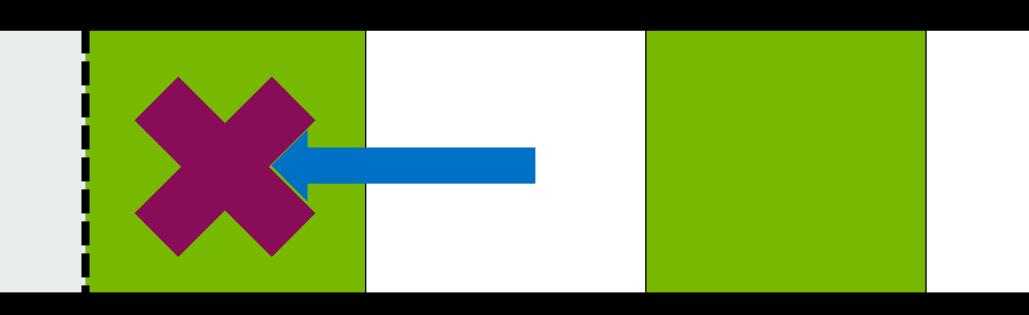
- Exterior domain
- For non-overlapping blocks, there's no contribution from outside the domain
- Above: contribution from the left is zero
- For this simple stencil, this is equivalent to zeroing out the hopping term itself... ...that thinking is trouble

### Zero Boundaries

"Boundary clovers"

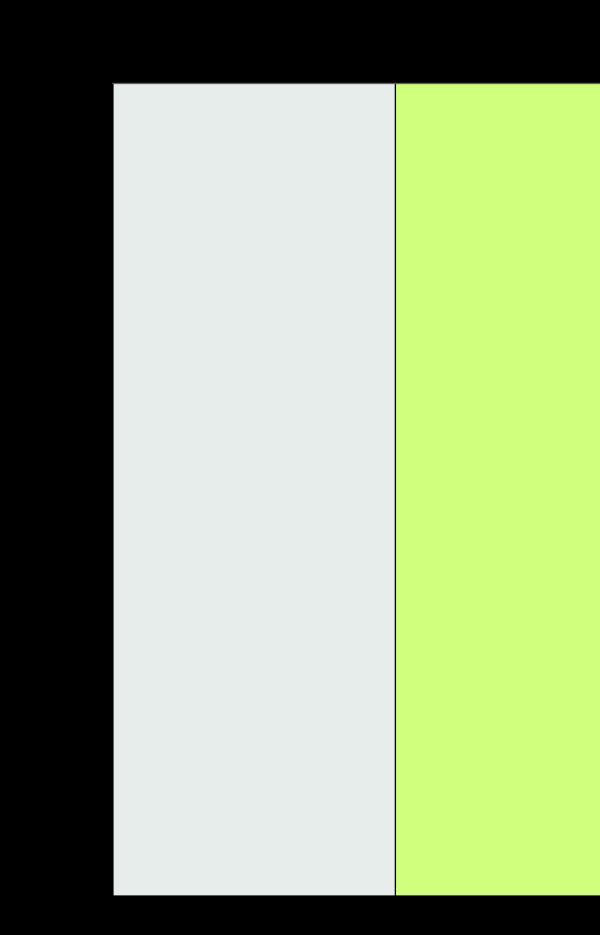
 Let's consider the massless staggered operator... in one dimension, for extreme simplicity  $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger}(x-\hat{\mu}) \delta_{x,y+1} \right]$ 

### • The stencil gathers from two sites: one on the left, and one on the right



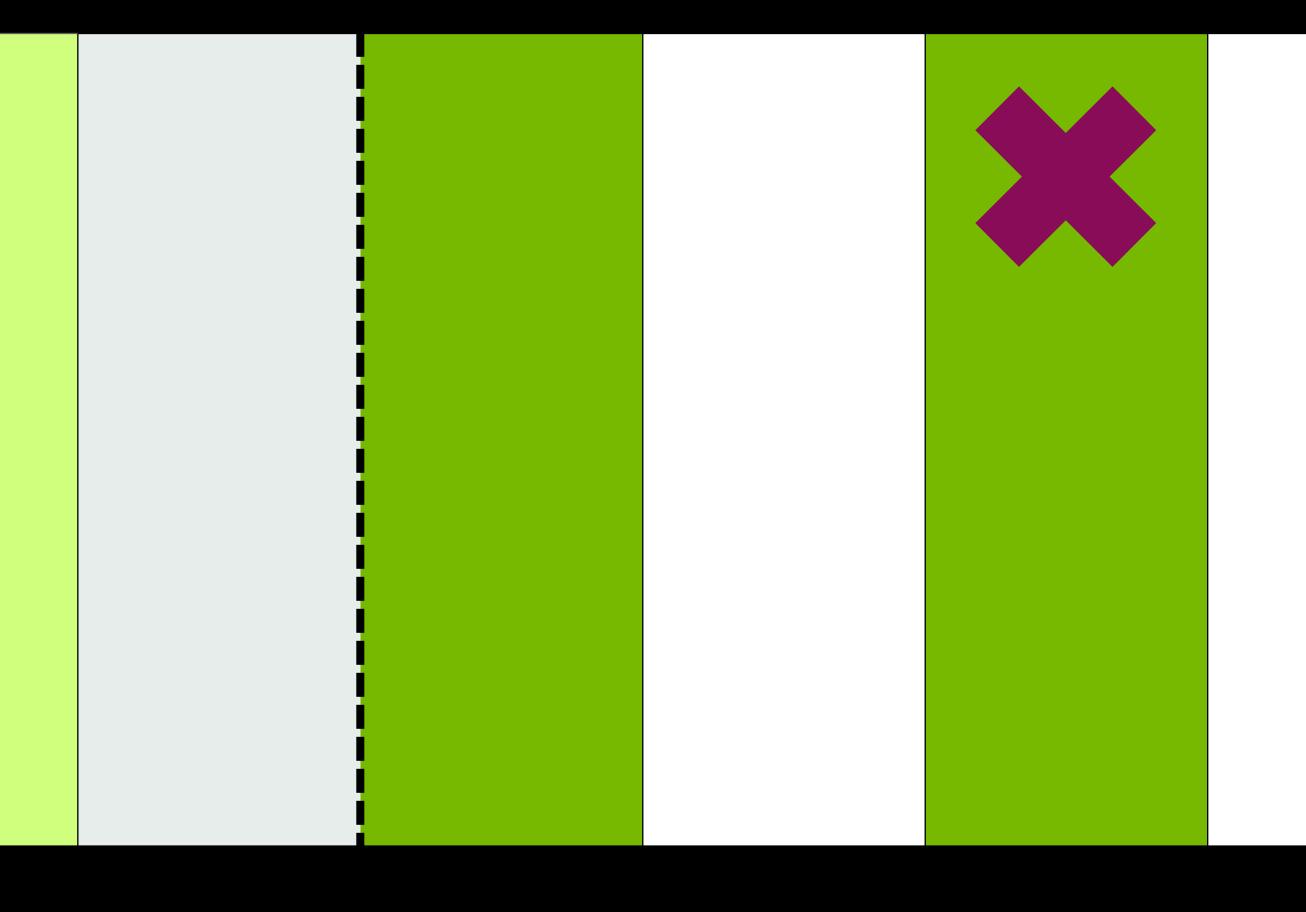
### Interior domain





### Squared operator

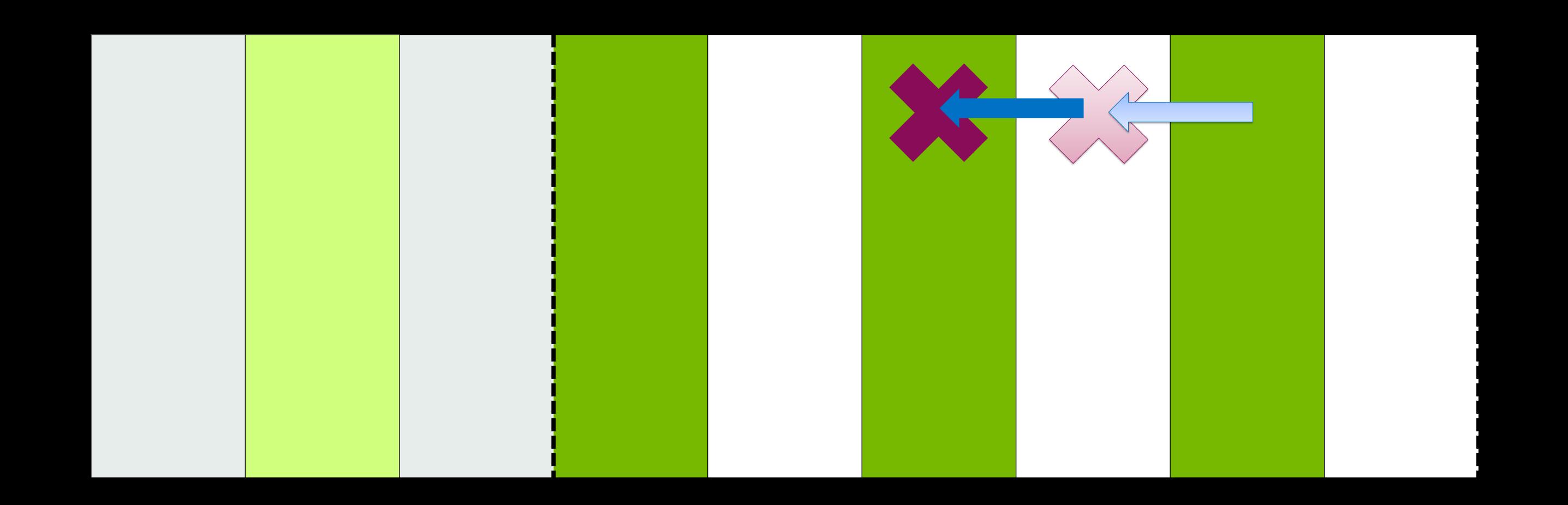
• Let's consider the massless operator **squared**... in one dimension, to keep bookkeeping easy  $D_{x,y}^{stag} \approx \left[M_{\mu}(x)\delta_{x,y-1} - M_{\mu}^{\dagger}(x-\hat{\mu})\delta_{x,y+1}\right]$ 







### $\approx M_{\mu}(x)M_{\mu}(x+\hat{\mu})\delta_{x,y-2}$ From the right

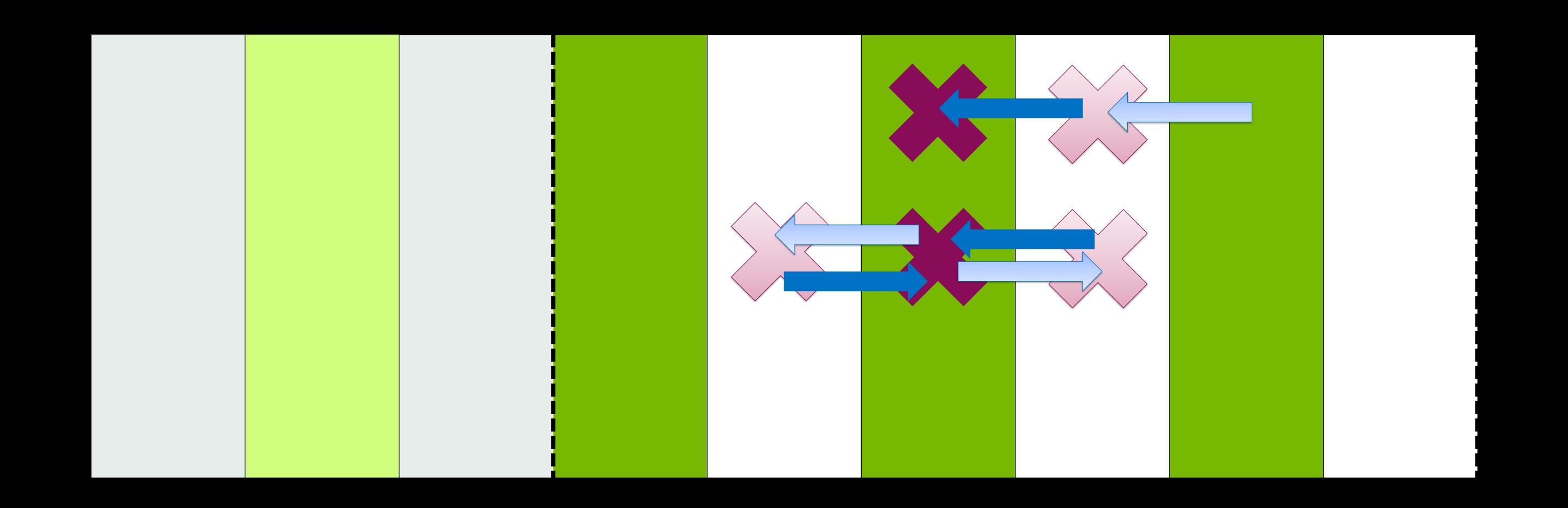


### Squared operator

 Let's consider the massless operator squared... in one dimension, to keep bookkeeping easy  $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger}(x - \hat{\mu}) \delta_{x,y+1} \right]$ 







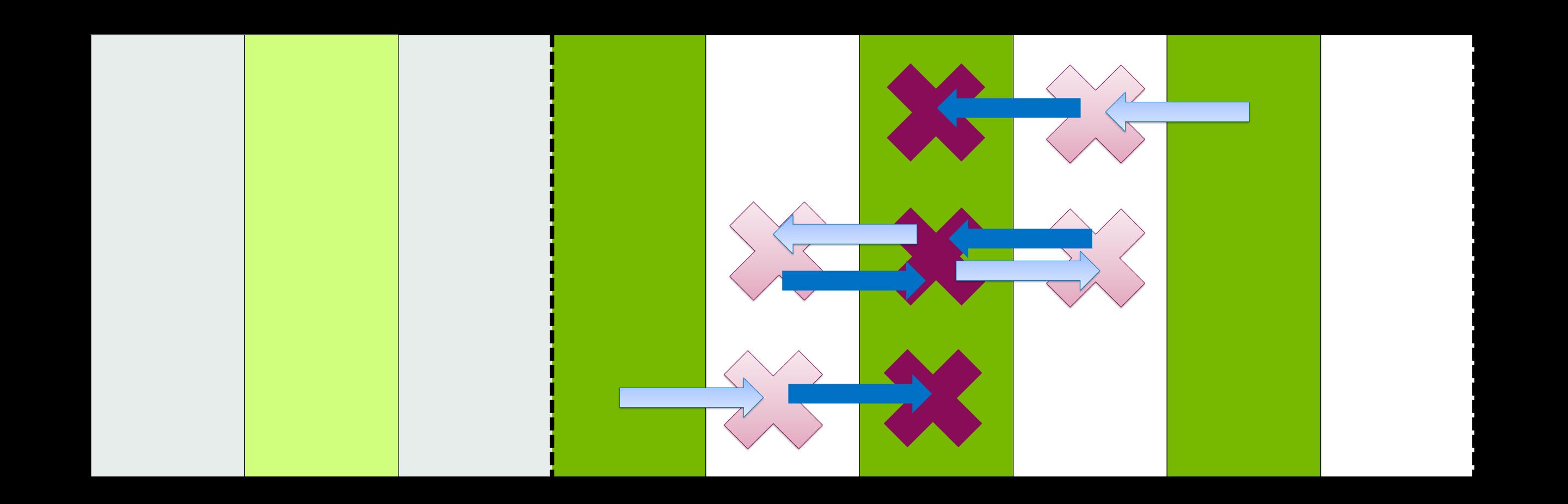
### Squared operator

 Let's consider the massless operator squared... in one dimension, to keep bookkeeping easy  $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger}(x-\hat{\mu}) \delta_{x,y+1} \right]$ 

 $\approx \underbrace{M_{\mu}(x)M_{\mu}(x+\hat{\mu})\delta_{x,y-2}}_{,y-2} - \underbrace{\left[M_{\mu}(x)M_{\mu}^{\dagger}(x) + M_{\mu}(x-\hat{\mu})M_{\mu}^{\dagger}(x-\hat{\mu})\right]\delta_{y,z}}_{,y-2}$ 

From self





### Squared operator

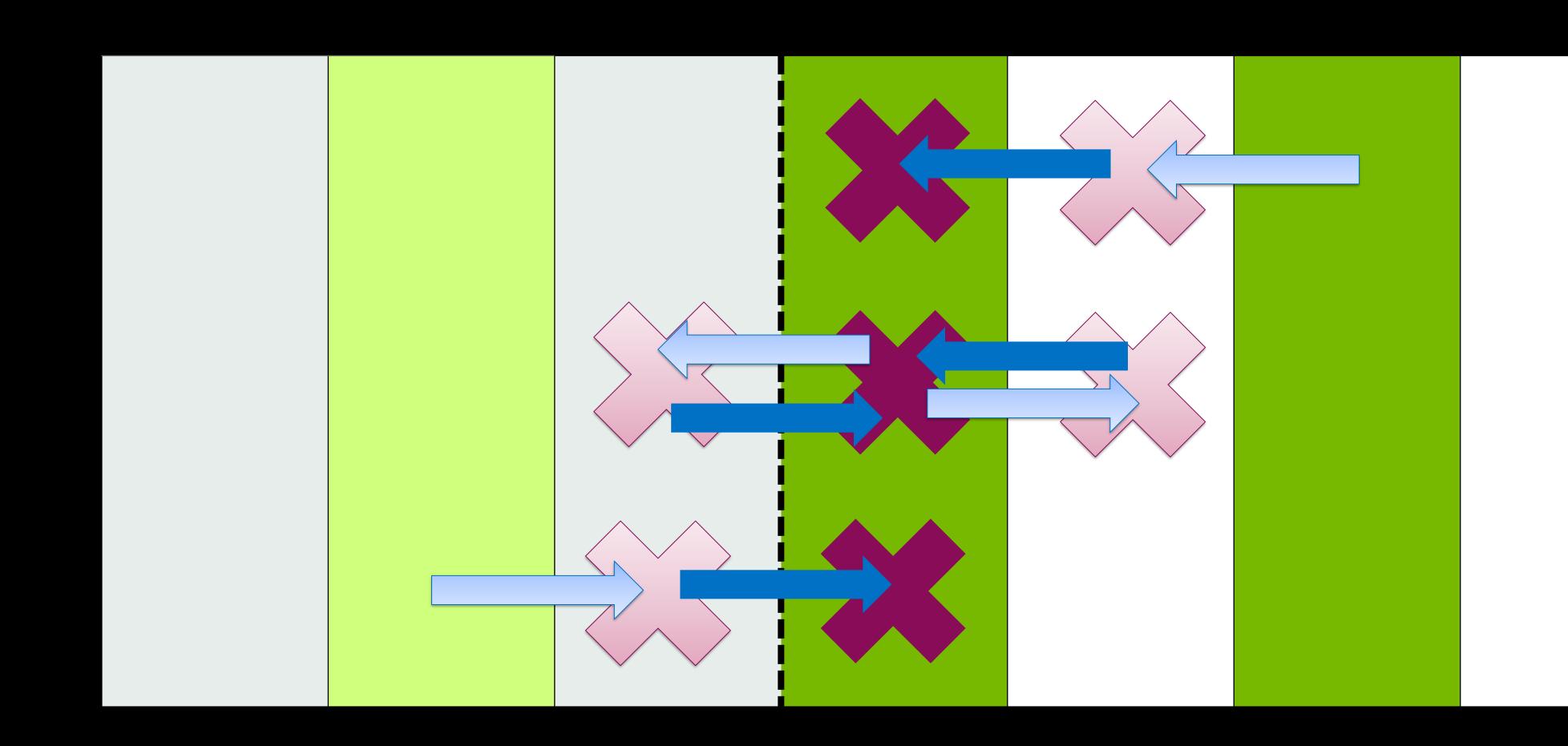
 Let's consider the massless operator squared... in one dimension, to keep bookkeeping easy  $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger} (x - \hat{\mu}) \delta_{x,y+1} \right]$ 

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From self

### From the left





### Squared operator on the Boundary There's always a catch

 Let's consider the massless operator squared... in one dimension, to keep bookkeeping easy  $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger}(x-\hat{\mu}) \delta_{x,y+1} \right]$ 

 $\approx \underbrace{M_{\mu}(x)M_{\mu}(x+\hat{\mu})\delta_{x,y-2}}_{\mu} - \left[M_{\mu}(x)M_{\mu}^{\dagger}(x) + M_{\mu}(x-\hat{\mu})M_{\mu}^{\dagger}(x-\hat{\mu})\right]\delta_{y,z} + M_{\mu}^{\dagger}(x-\hat{\mu})M_{\mu}^{\dagger}(x-2\hat{\mu})\delta_{x,y+2}$ 

From self

From the left





### Sidebar: MSPCG Work **Mobius Fermions**

 $D_{oe}^{\dagger} D_{eo}^{\dagger} D_{eo} D_{oe}$ 

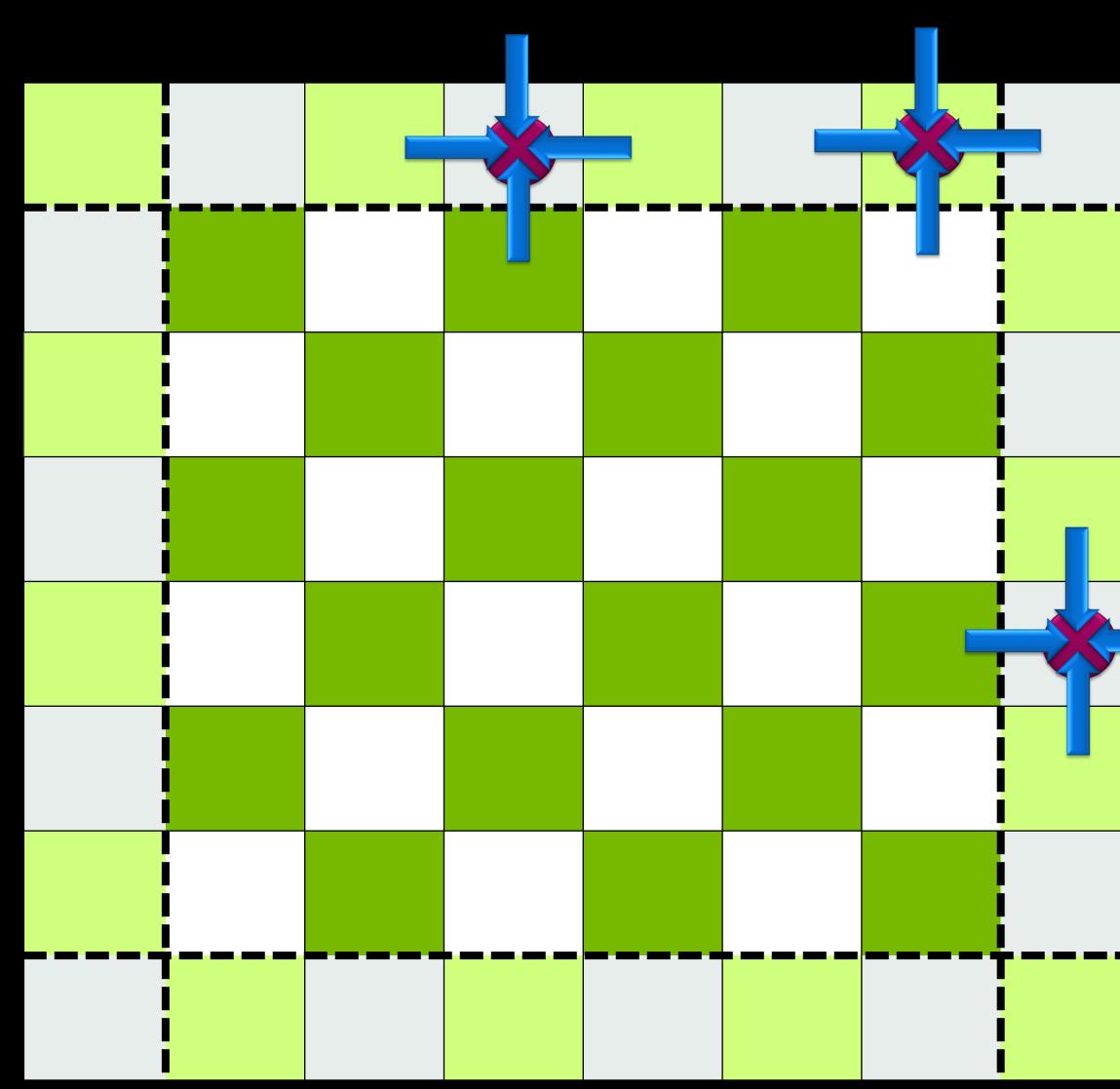


**NVIDIA**.

### Four steps, one for each operator application **1.** $D_{oe}$ on $(L + 2)^4$ volume

### **Existing Work Mobius Fermions**

 $D_{oe}^{\dagger} D_{eo}^{\dagger} D_{eo} D_{oe}^{\dagger}$ 





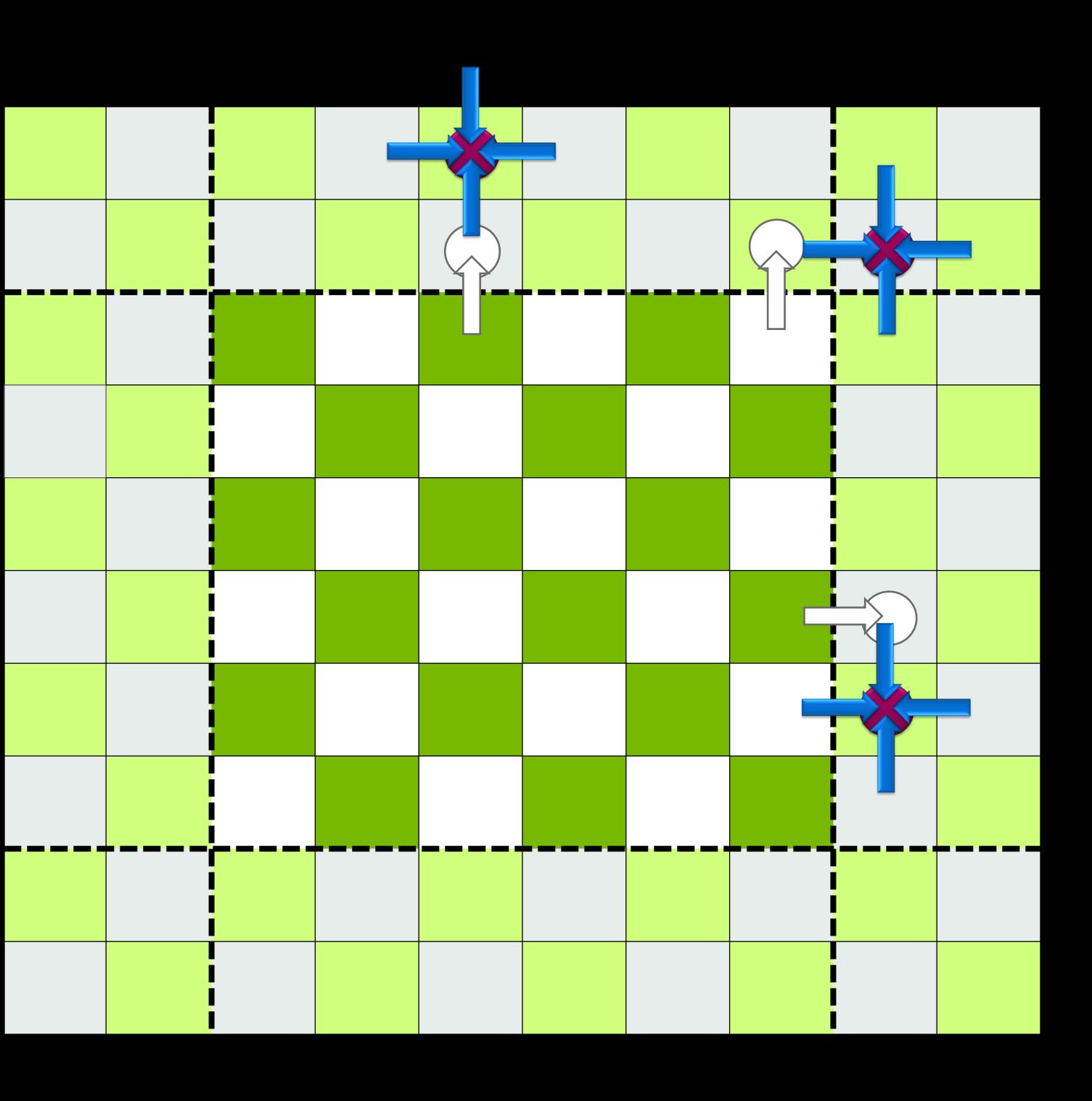
**NVIDIA**.

### Four steps, one for each operator application

- **1.**  $D_{oe}$  on  $(L + 2)^4$  volume
- 2.  $D_{eo}$  on  $(L + 4)^4$  volume

### **Existing Work Mobius Fermions**

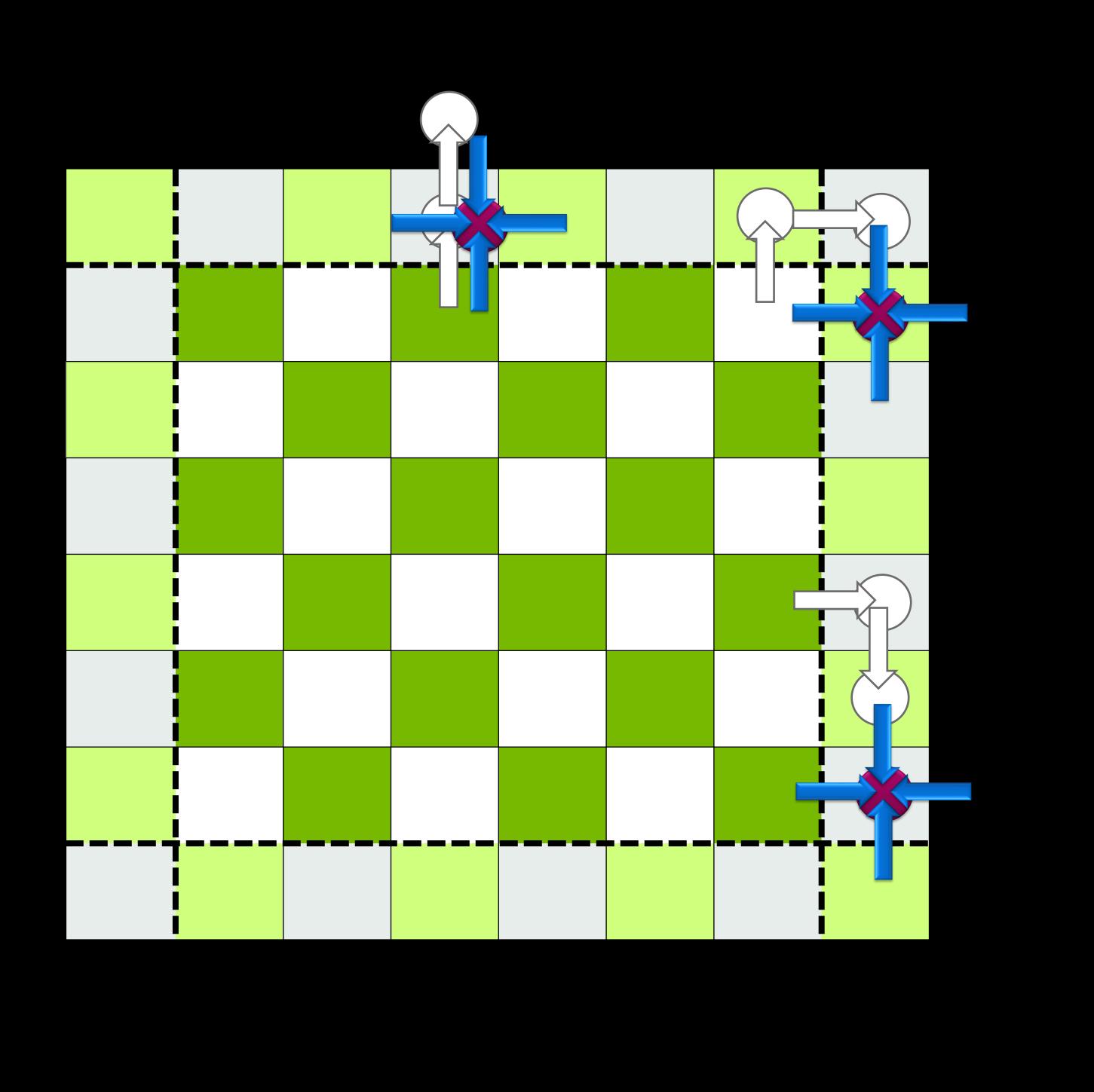
 $D_{oe}^{\dagger} D_{eo}^{\dagger} D_{eo} D_{oe}$ 



- Four steps, one for each operator application
  - **1.**  $D_{oe}$  on  $(L + 2)^4$  volume
  - 2.  $D_{eo}$  on  $(L + 4)^4$  volume
  - 3.  $D_{eo}^{\dagger}$  on  $(L+2)^4$  volume

### **Existing Work Mobius Fermions**

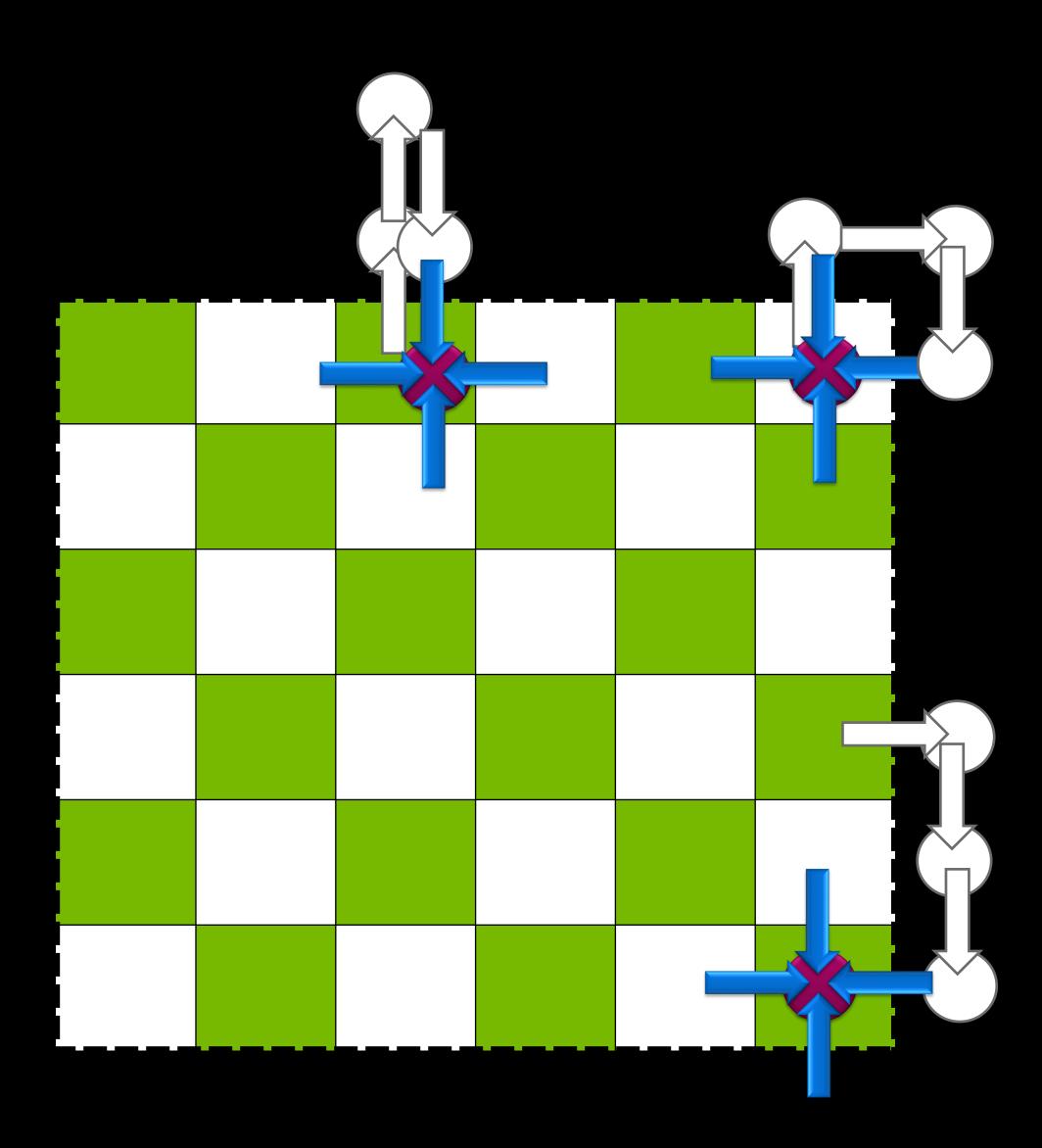
 $D_{oe}^{\dagger} D_{eo}^{\dagger} D_{eo} D_{oe}$ 



- Four steps, one for each operator application
  - **1.**  $D_{oe}$  on  $(L + 2)^4$  volume
  - 2.  $D_{eo}$  on  $(L + 4)^4$  volume
  - 3.  $D_{eo}^{\dagger}$  on  $(L+2)^4$  volume
  - 4.  $D_{oe}^{\dagger}$  on on  $L^4$  volume

### **Existing Work Mobius Fermions**

 $D_{oe}^{\dagger} D_{eo}^{\dagger} D_{eo} D_{oe}$ 



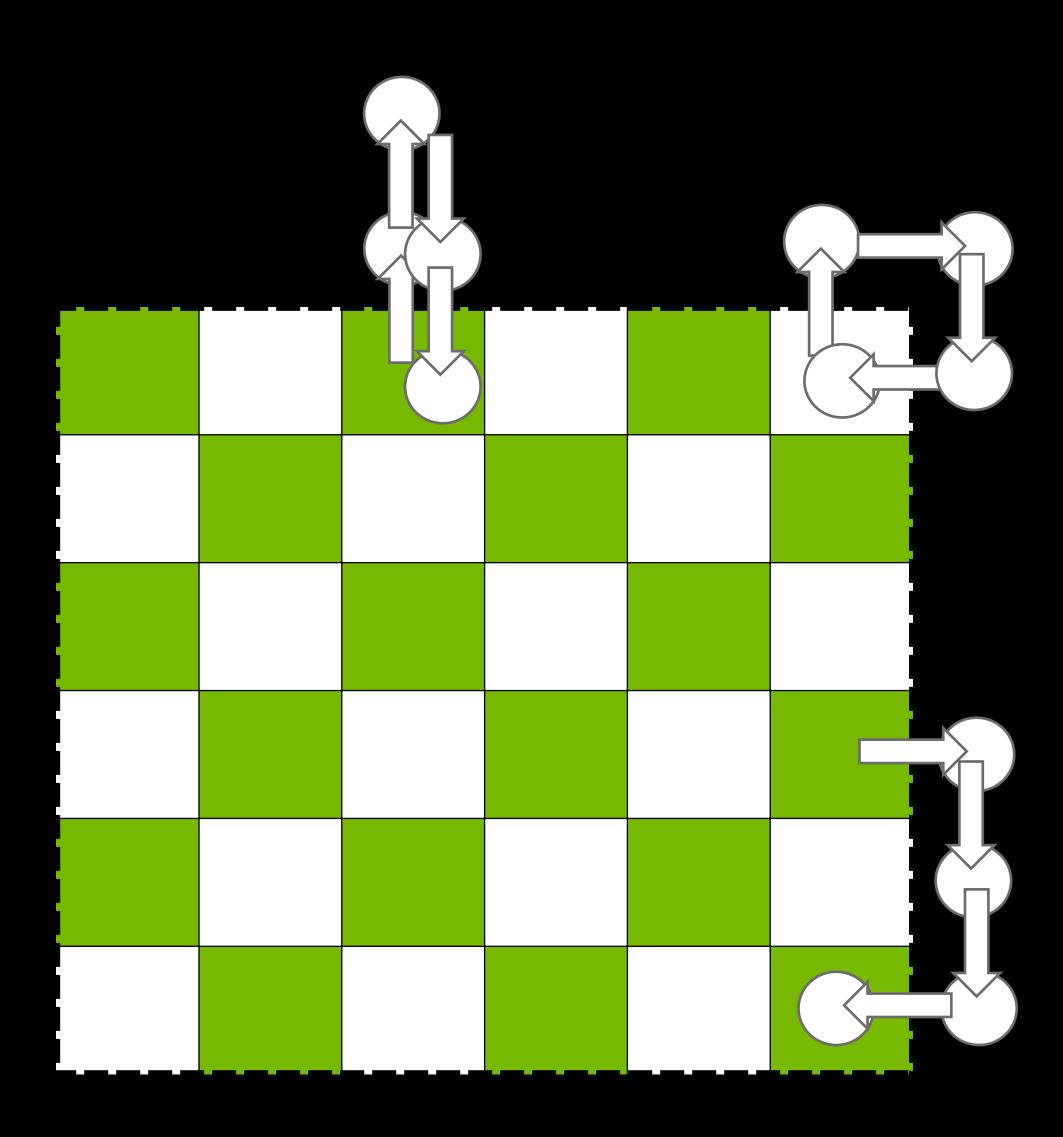


- Four steps, one for each operator application
  - **1**.  $D_{oe}$  on  $(L + 2)^4$  volume
  - 2.  $D_{eo}$  on  $(L + 4)^4$  volume
  - 3.  $D_{eo}^{\dagger}$  on  $(L+2)^4$  volume
  - 4.  $D_{oe}^{\dagger}$  on on  $L^4$  volume
- domains (strong-scaling regime)

### **Existing Work Mobius Fermions**

 $D_{oe}^{\dagger} D_{eo}^{\dagger} D_{eo} D_{oe}$ 

This extra work can be very expensive; non-trivially so for small local



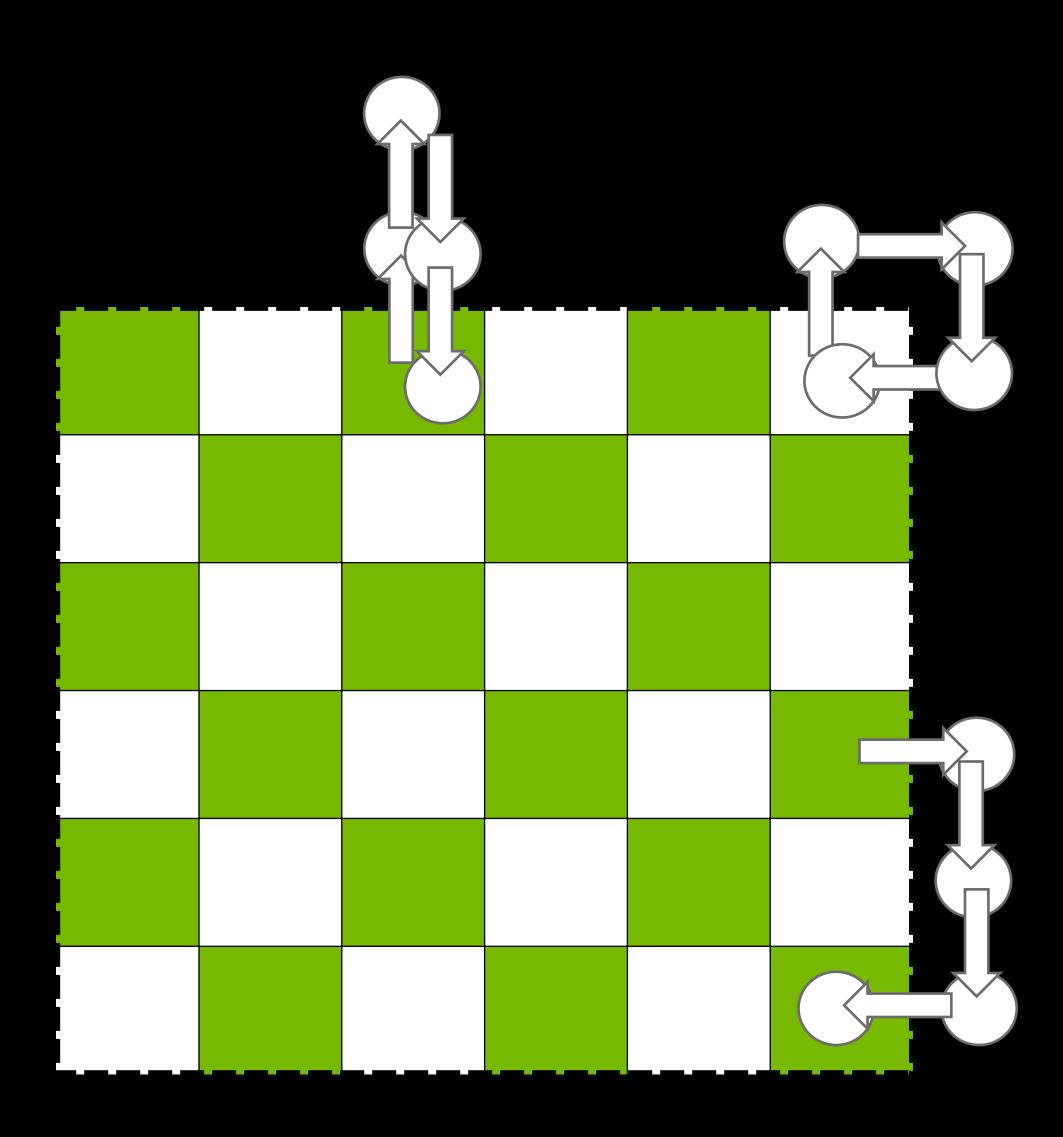


- Four steps, one for each operator application
  - **1**.  $D_{oe}$  on  $(L + 2)^4$  volume
  - 2.  $D_{eo}$  on  $(L + 4)^4$  volume
  - 3.  $D_{eo}^{\dagger}$  on  $(L+2)^4$  volume
  - 4.  $D_{oe}^{\dagger}$  on on  $L^4$  volume
- domains (strong-scaling regime)
- HISQ fermions have relative benefits and challenges
  - Only  $D_{eo}D_{oe}$
  - Need to bookkeep distance-1 and distance-3 terms
  - Distance-3 terms would necessitate an  $(L + 6)^4$  volume

### **Existing Work** Mobius Fermions

 $D_{oe}^{\dagger} D_{eo}^{\dagger} D_{eo} D_{oe}$ 

This extra work can be very expensive; non-trivially so for small local

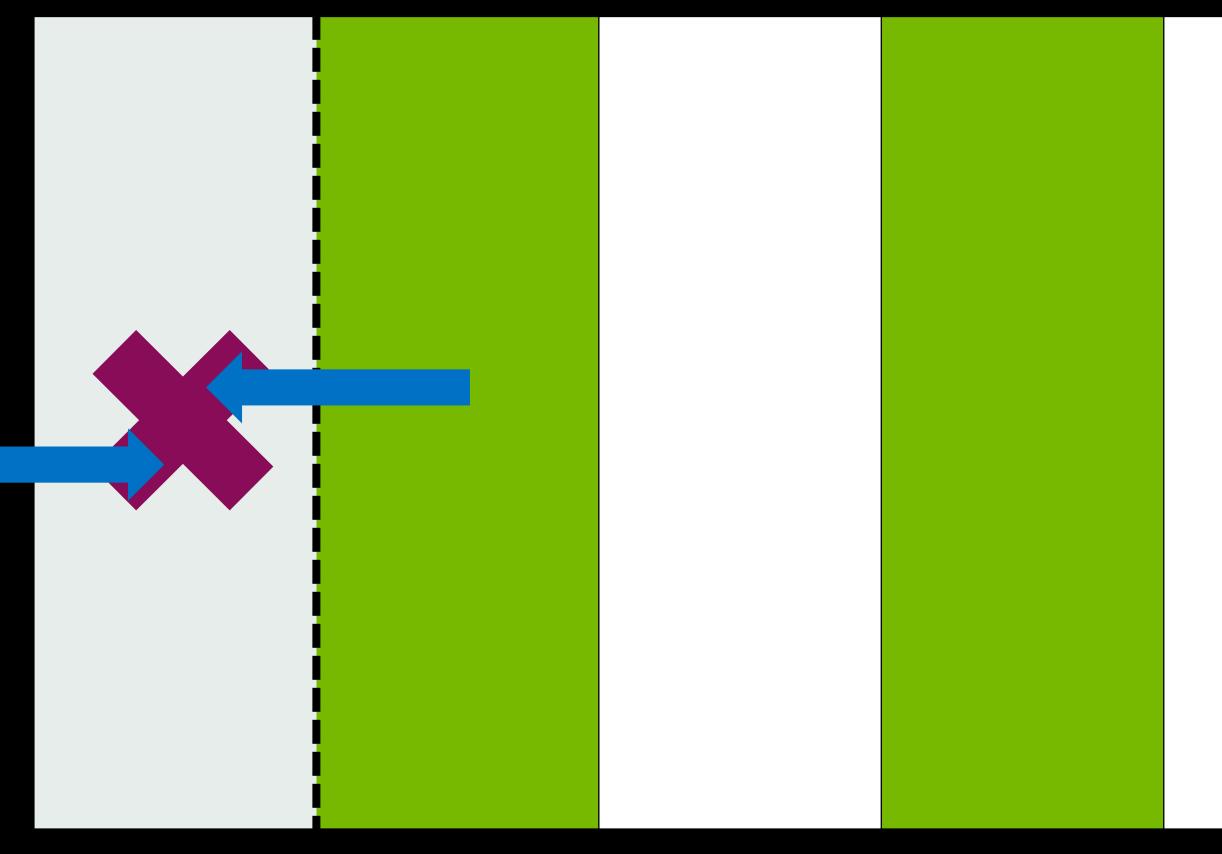




Step one: calculate including the extended domain

### Application to 1-d Staggered Extended domains

### $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger}(x-\hat{\mu}) \delta_{x,y+1} \right]$

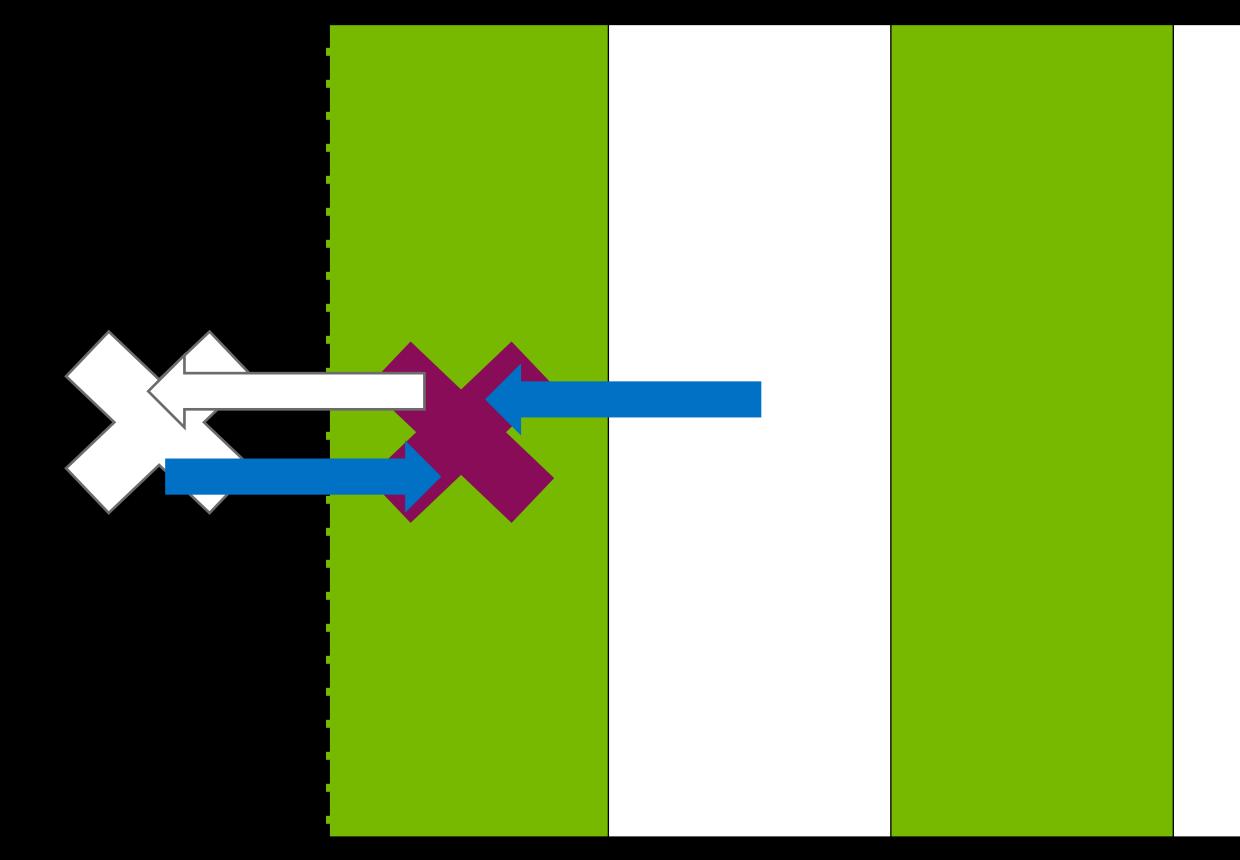




Step two: only calculate within the interior

### Application to 1-d Staggered Extended domains

### $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger}(x-\hat{\mu}) \delta_{x,y+1} \right]$



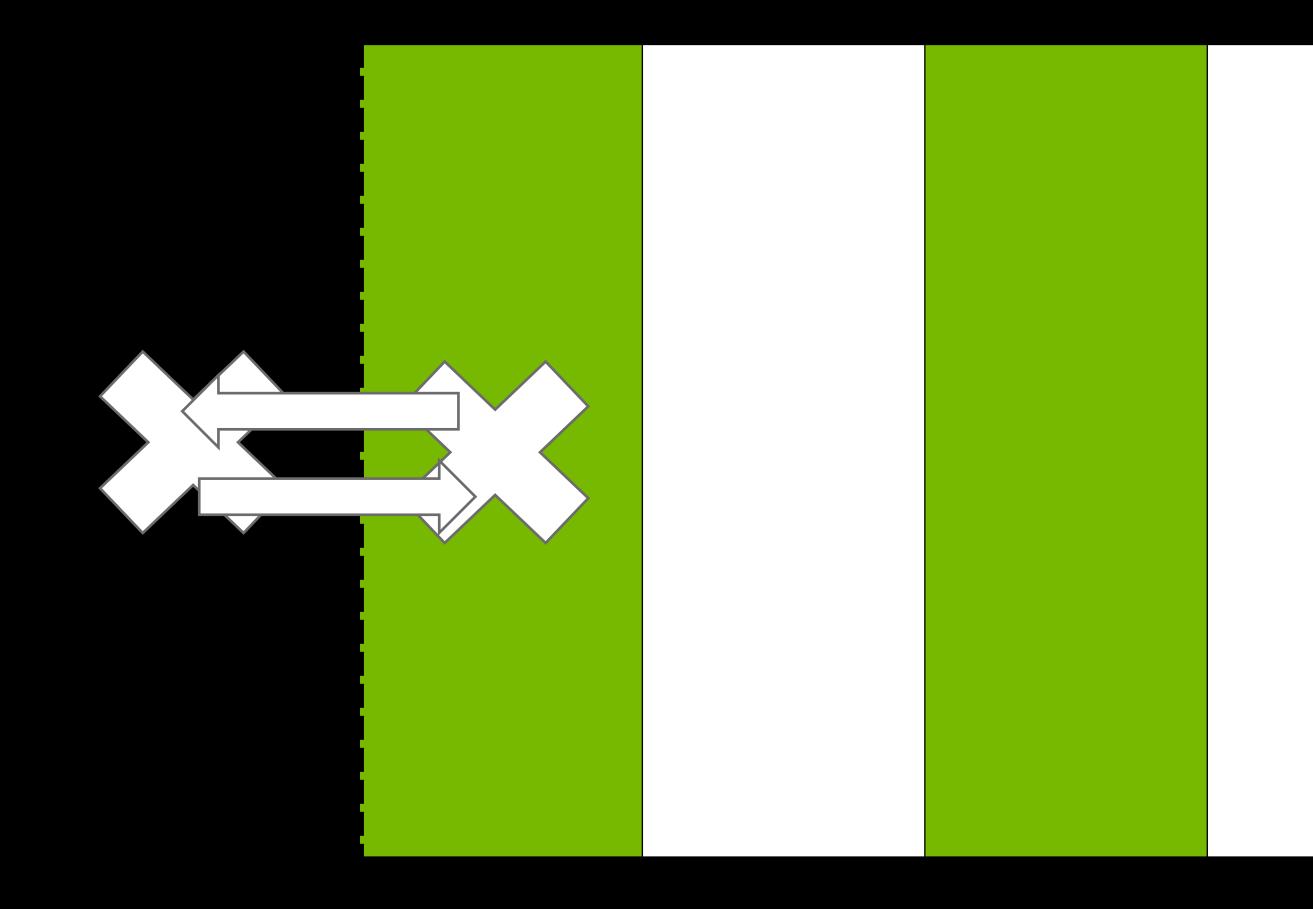


This also gives you the boundary term

### Application to 1-d Staggered

 $D_{x,y}^{stag} \approx \left[ M_{\mu}(x) \delta_{x,y-1} - M_{\mu}^{\dagger}(x-\hat{\mu}) \delta_{x,y+1} \right]$ 

From self





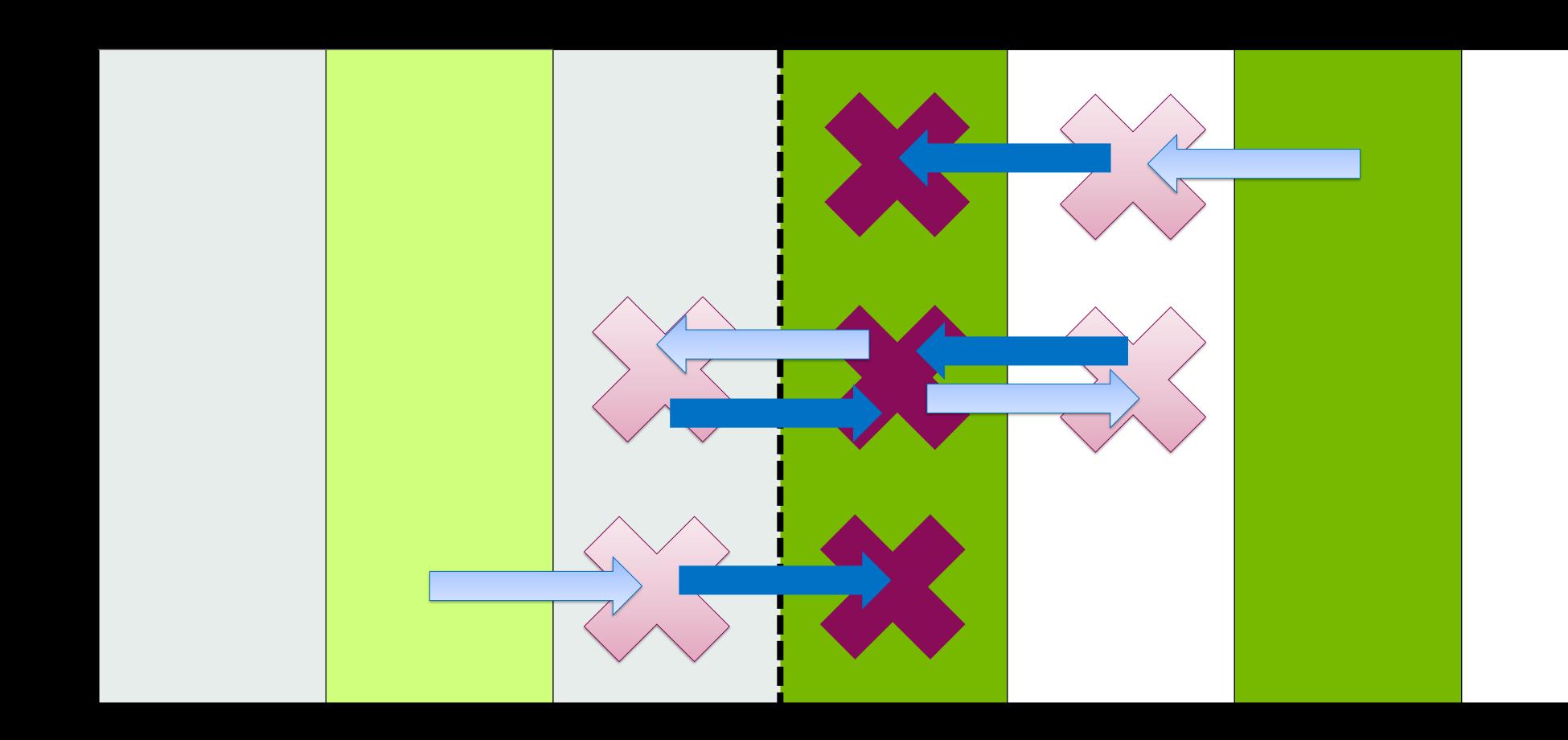
### $\approx M_{\mu}(x)M_{\mu}(x+\hat{\mu})\delta_{x,y-2} - \left[M_{\mu}(x)M_{\mu}^{\dagger}(x) + M_{\mu}(x-\hat{\mu})M_{\mu}^{\dagger}(x-\hat{\mu})\right]\delta_{y,z} + M_{\mu}^{\dagger}(x-\hat{\mu})M_{\mu}^{\dagger}(x-2\hat{\mu})\delta_{x,y+2}$

### From the left









### **Alternative Form: "Boundary Clover"**

 $\approx \underbrace{M_{\mu}(x)M_{\mu}(x+\hat{\mu})\delta_{x,y-2}}_{\mu} - \underbrace{\left[M_{\mu}(x)M_{\mu}^{\dagger}(x) + M_{\mu}(x-\hat{\mu})M_{\mu}^{\dagger}(x-\hat{\mu})\right]\delta_{y,z}}_{\mu} + \underbrace{M_{\mu}^{\dagger}(x-\hat{\mu})M_{\mu}^{\dagger}(x-2\hat{\mu})\delta_{x,y+2}}_{\mu}$ From self

• Alternative approach: what if we "just" calculated the self-contribution ("boundary clover") directly?

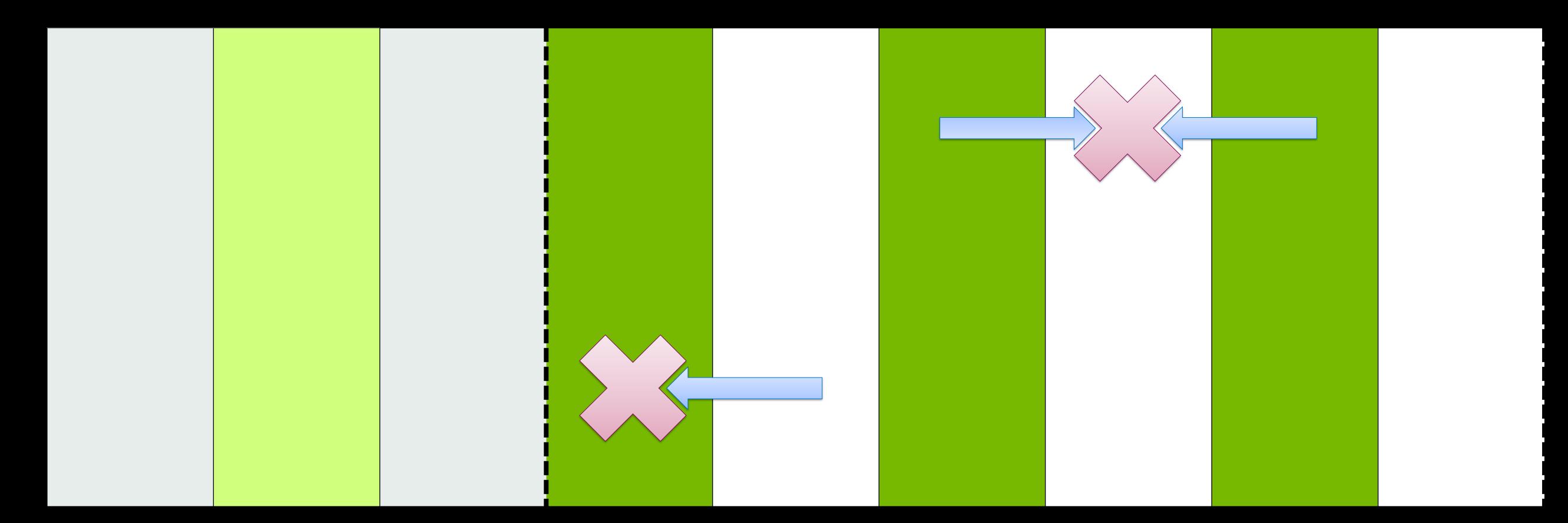
From the left





### Implementing a Boundary Clover Workflow Step 1

- An implementation in two parts:
- Step 1: Apply the operator with Dirichlet boundary conditions
  - For operators in the interior, this is nothing interesting
  - For operators on the boundary, it's a quick snip

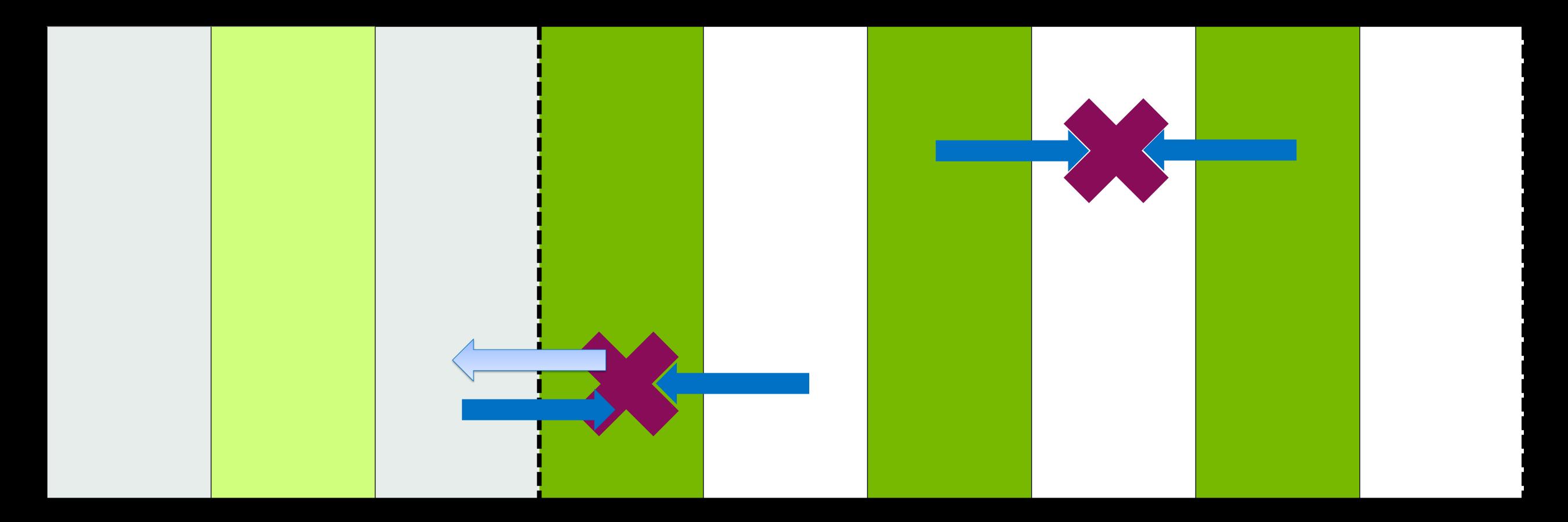


Dirichlet boundary conditions s is nothing interesting it's a quick snip

<



- An implementation in two parts:
- - For operators on the interior, this is nothing special
- Key optimizations:
  - We can reuse the same link for the "out" as the "in"



### **Boundary Clover** Step 2

 Step 2: Apply the operator with "clover" computations on the boundary • For operators on the boundary, in the direction of the boundary, compute the full hop "out and in"

We could create a custom field with this pre-computed to avoid the multiplication

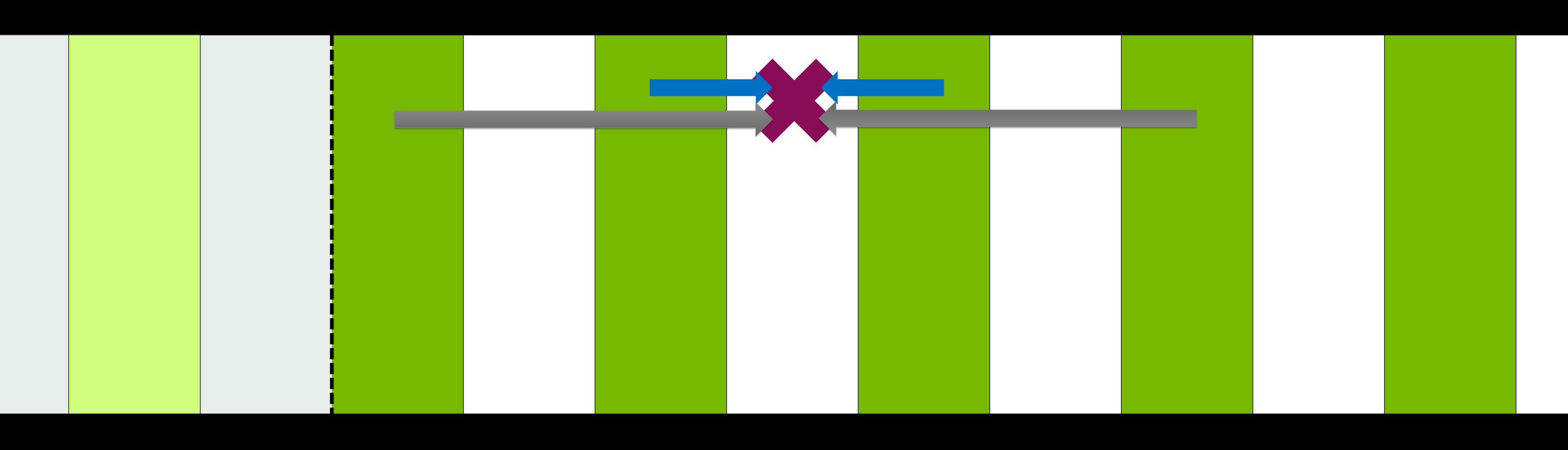


# Application to HISQ



$$D_{x,y}^{HISQ} \approx \sum_{\mu=0}^{3} \eta_{\mu}(x) \Big[ \big(F_{\mu}(x) \big) \Big] \Big]$$

Here, F is the distance 1 "fat link" and L is the distance 3 "long link"



### **Review: HISQ Stencil** Three hops this time

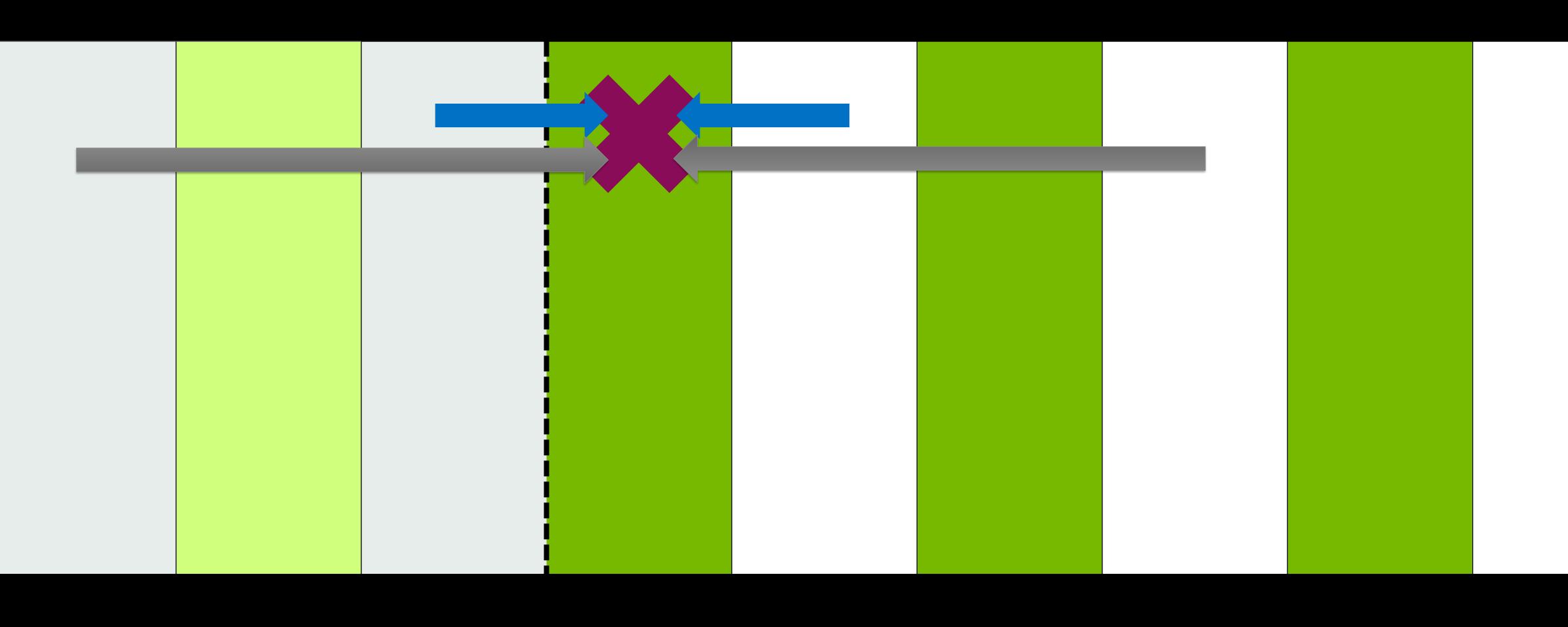
• On face value, the HISQ stencil has no complications relative to the naïve staggered example

$$(x)\delta_{x,y-1} - F_{\mu}^{\dagger}(x-\hat{\mu})\delta_{x,y+1}) + (L_{\mu}(x)\delta_{x,y-3} - L_{\mu}^{\dagger})$$

 $\frac{\dagger}{\mu}(x-3\hat{\mu})\delta_{x,y+3})] + 2m\delta_{x,y}$ 

$$D_{x,y}^{HISQ} \approx \sum_{\mu=0}^{3} \eta_{\mu}(x) \Big[ \big(F_{\mu}(x)\big) \Big] \Big]$$

- This *does* lead to extra bookkeeping at the boundary



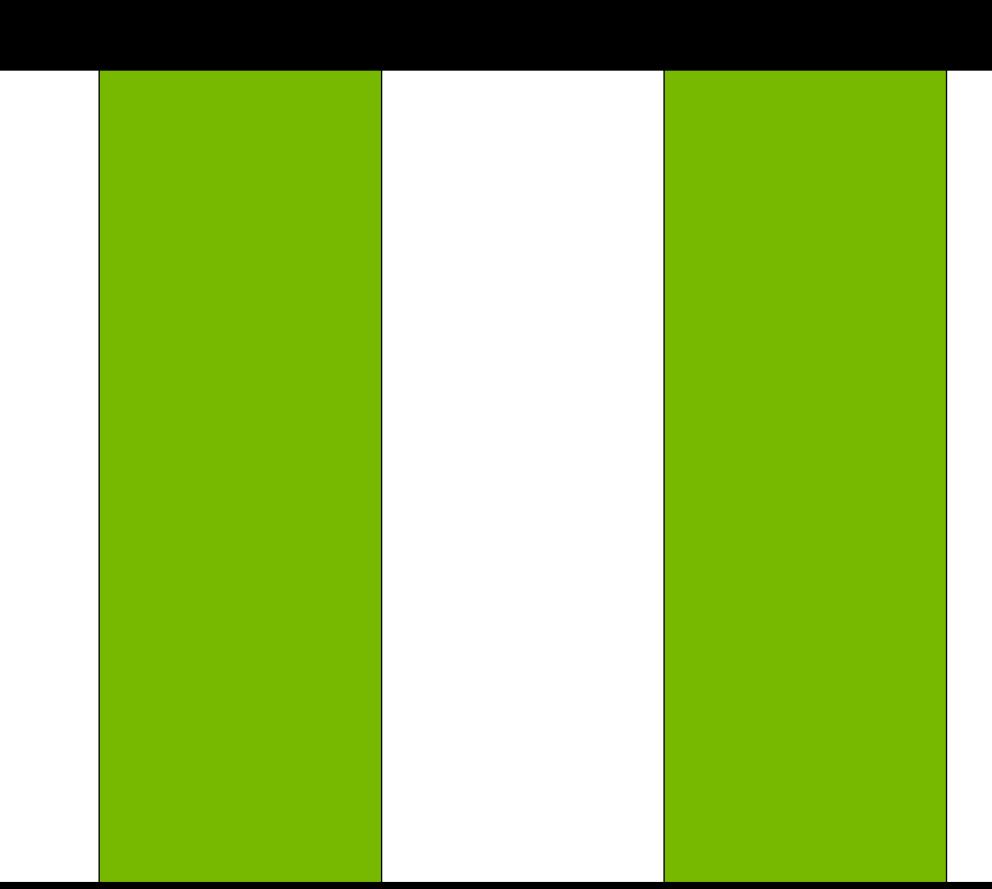
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$$(x)\delta_{x,y-1} - F_{\mu}^{\dagger}(x-\hat{\mu})\delta_{x,y+1}) + (L_{\mu}(x)\delta_{x,y-3} - L_{\mu}^{\dagger})$$

Here, F is the distance 1 "fat link" and L is the distance 3 "long link"

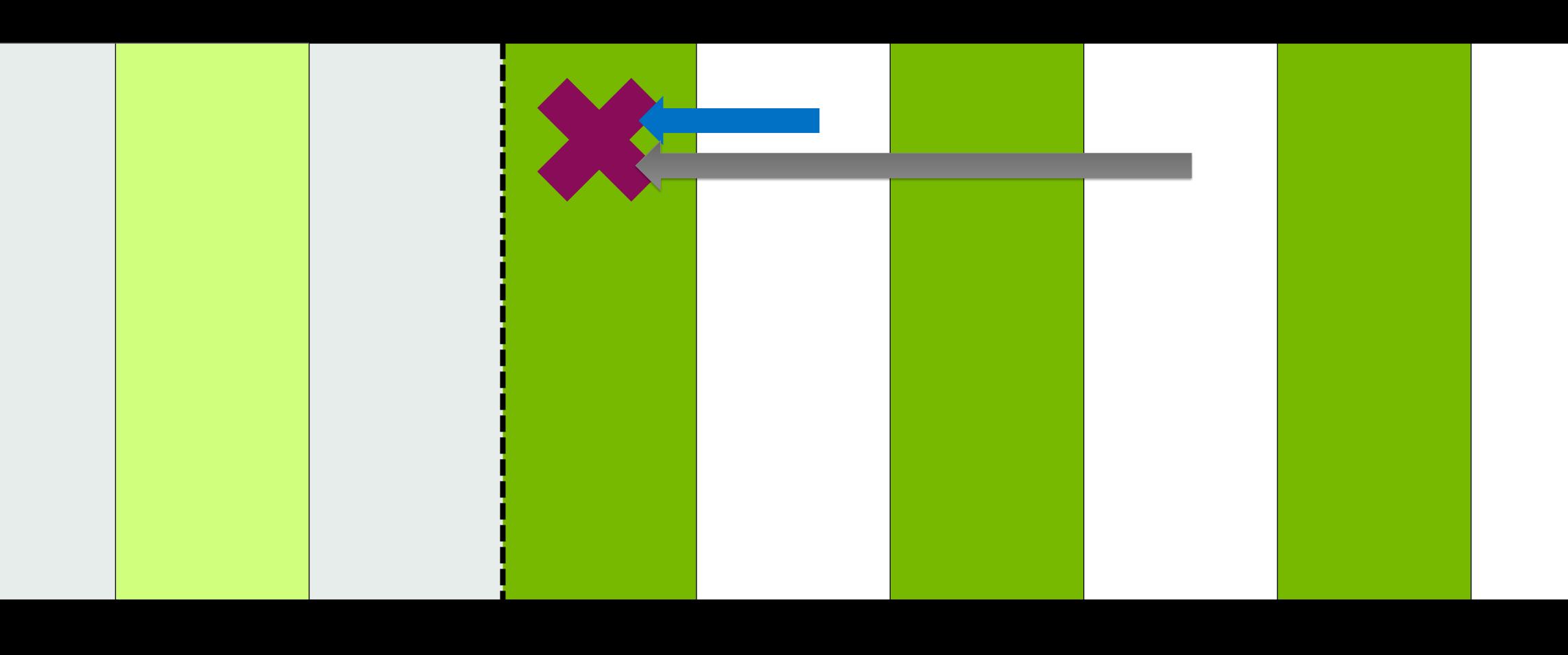
• Site at [0]: There are neither fat nor long link contributions from the "left": outside the domain

 $\int_{u}^{\dagger} (x - 3\hat{\mu})\delta_{x,y+3} \Big] + 2m\delta_{x,y}$ 



$$D_{x,y}^{HISQ} \approx \sum_{\mu=0}^{3} \eta_{\mu}(x) \Big[ \big(F_{\mu}(x)\big) \Big] \Big]$$

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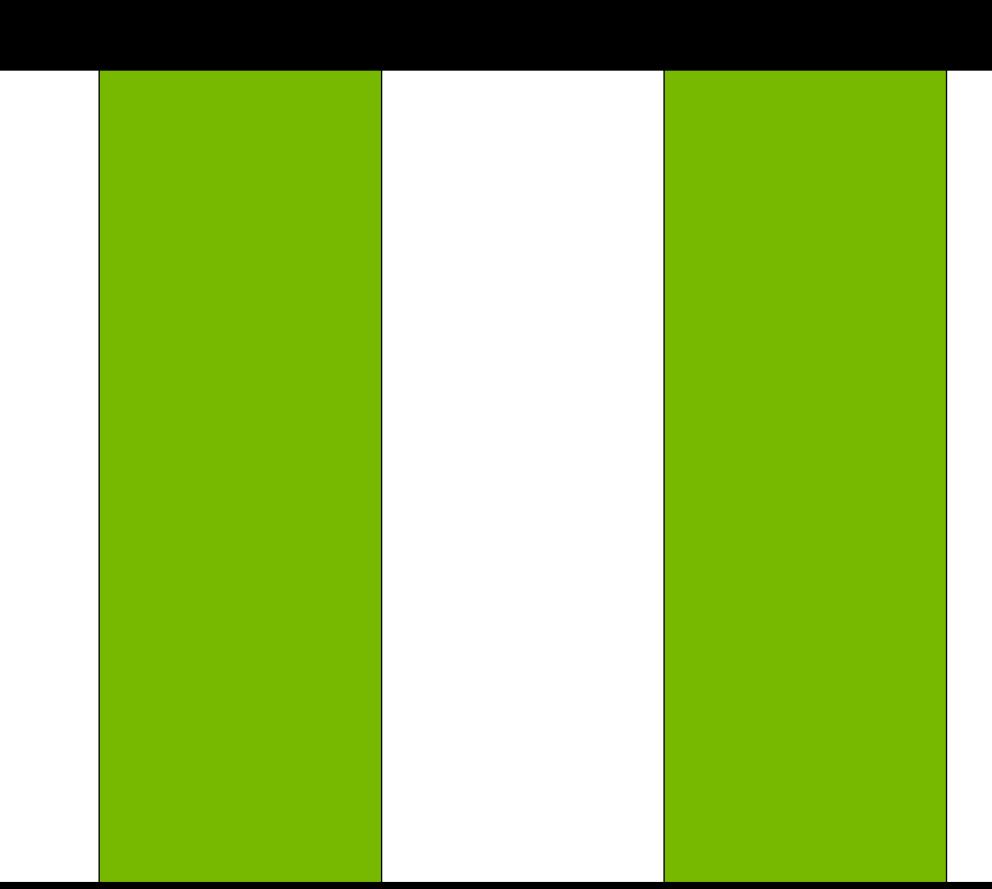
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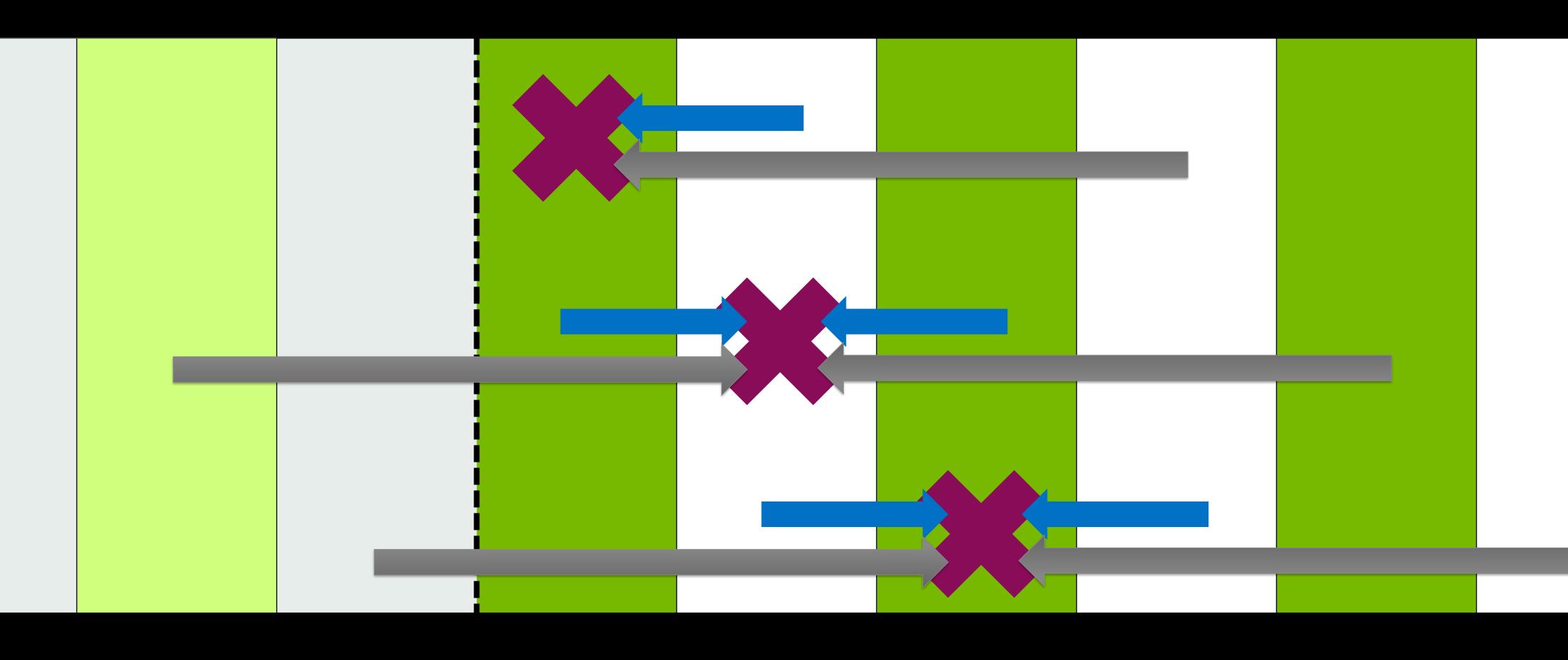
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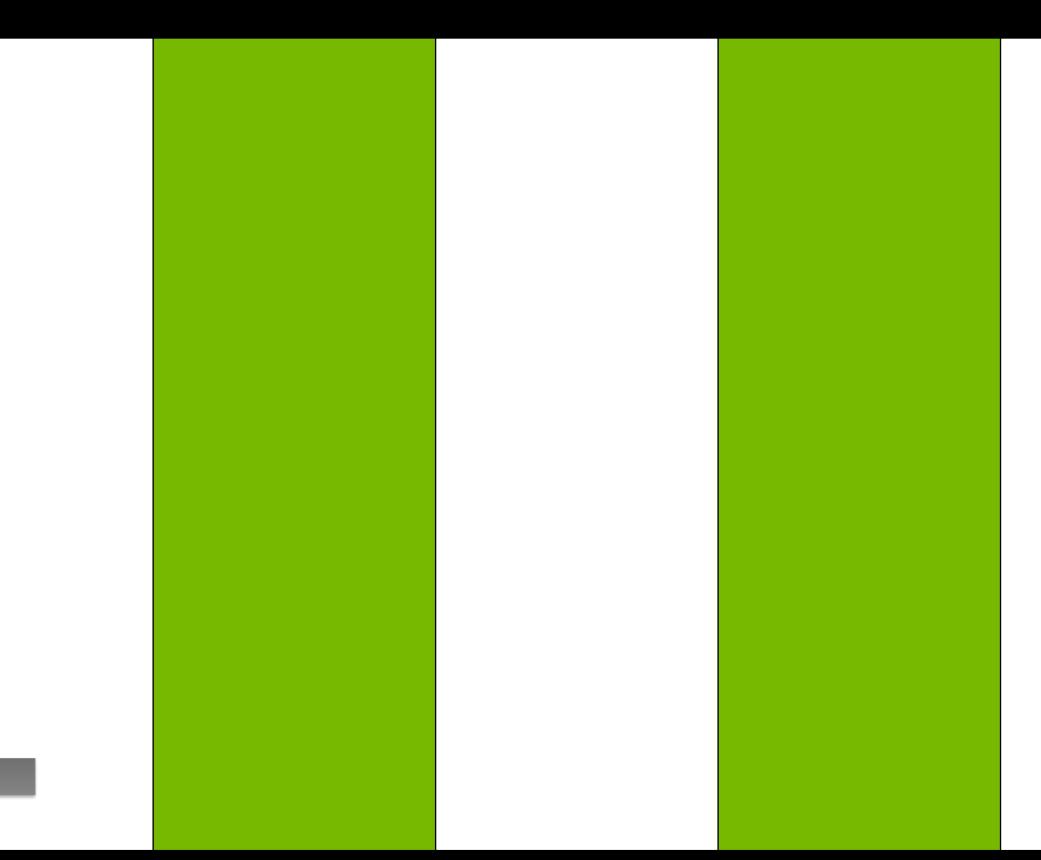
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$$(x)\delta_{x,y-1} - F_{\mu}^{\dagger}(x-\hat{\mu})\delta_{x,y+1}) + (L_{\mu}(x)\delta_{x,y-3} - L_{\mu}^{\dagger})$$

Here, F is the distance 1 "fat link" and L is the distance 3 "long link"

• Site at [0]: There are neither fat nor long link contributions from the "left": outside the domain • Sites at [1] or [2]: There is no long link contribution from the "left", but there's still a fat link contribution!

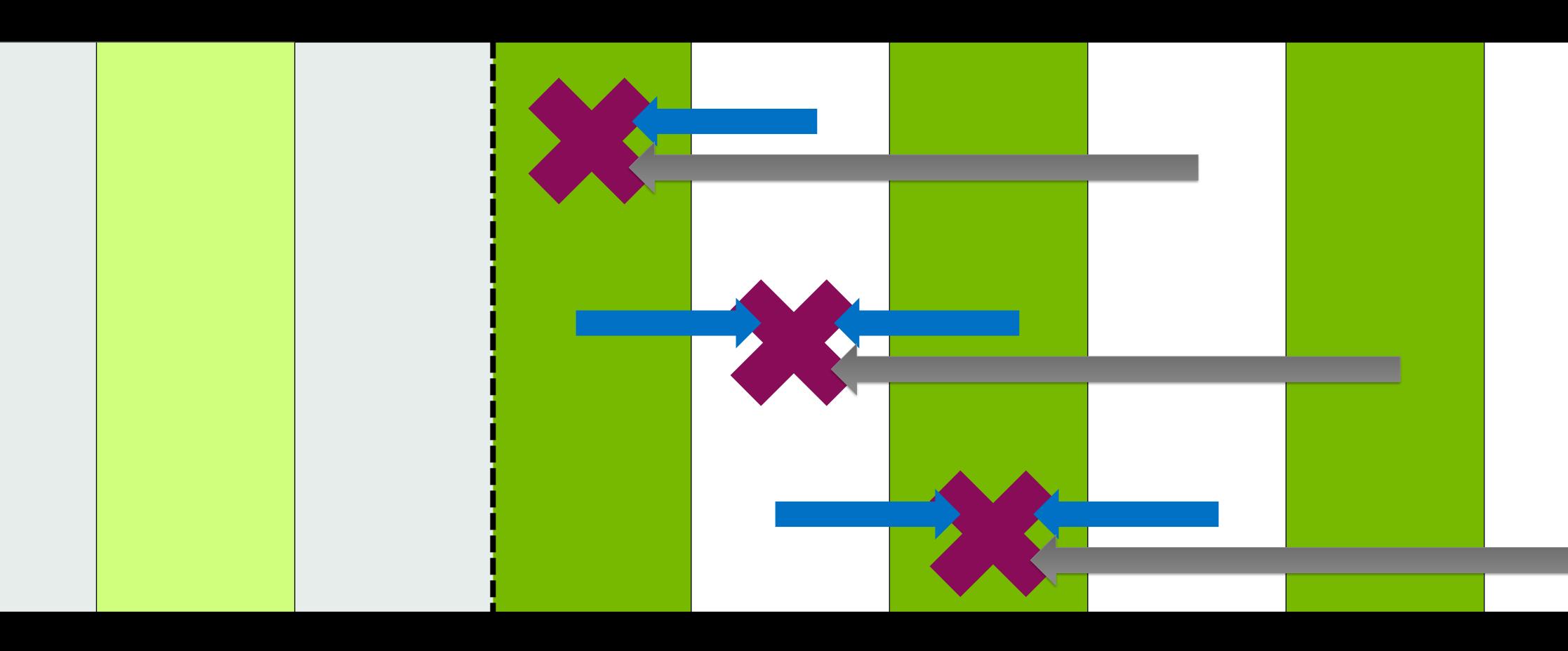
 $\int_{u}^{\dagger} (x - 3\hat{\mu}) \delta_{x,y+3} \Big] + 2m\delta_{x,y}$ 



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$$D_{x,y}^{HISQ} \approx \sum_{\mu=0}^{3} \eta_{\mu}(x) \Big[ \big(F_{\mu}(x)\big) \Big] \Big]$$

- This does lead to extra bookkeeping at the boundary



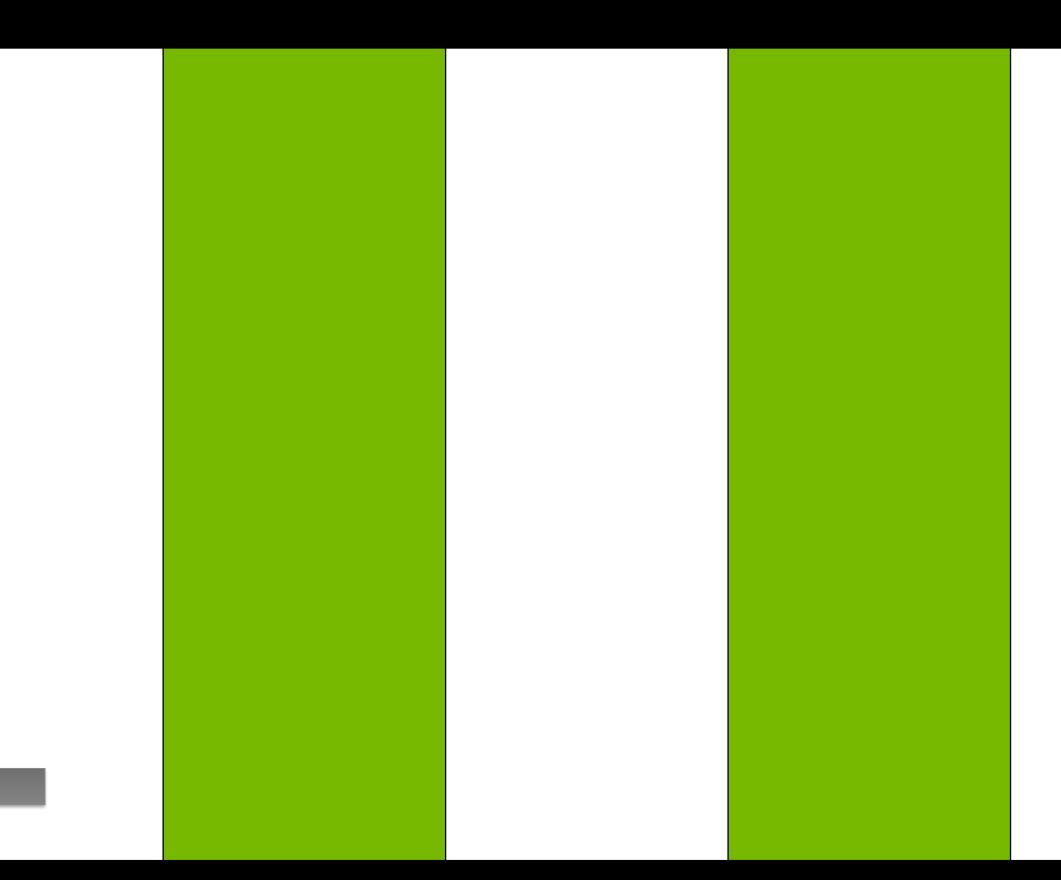
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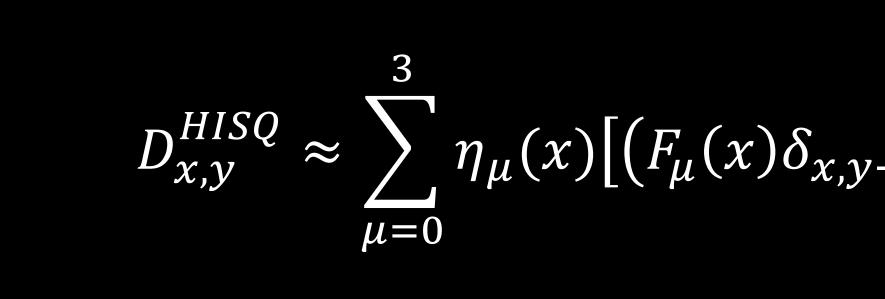
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 $\int_{u}^{\dagger} (x - 3\hat{\mu}) \delta_{x,y+3} \Big] + 2m\delta_{x,y}$ 



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### • The "real" goal is the even/odd preconditioned operator:



# $(4m^2 - D_{eo}D_{oe})x_e = 2mb_e - D_{eo}b_o$

The type of bookkeeping noted in the previous slide causes new headaches

### Schur Going to Have a Tough Time Three hops this time

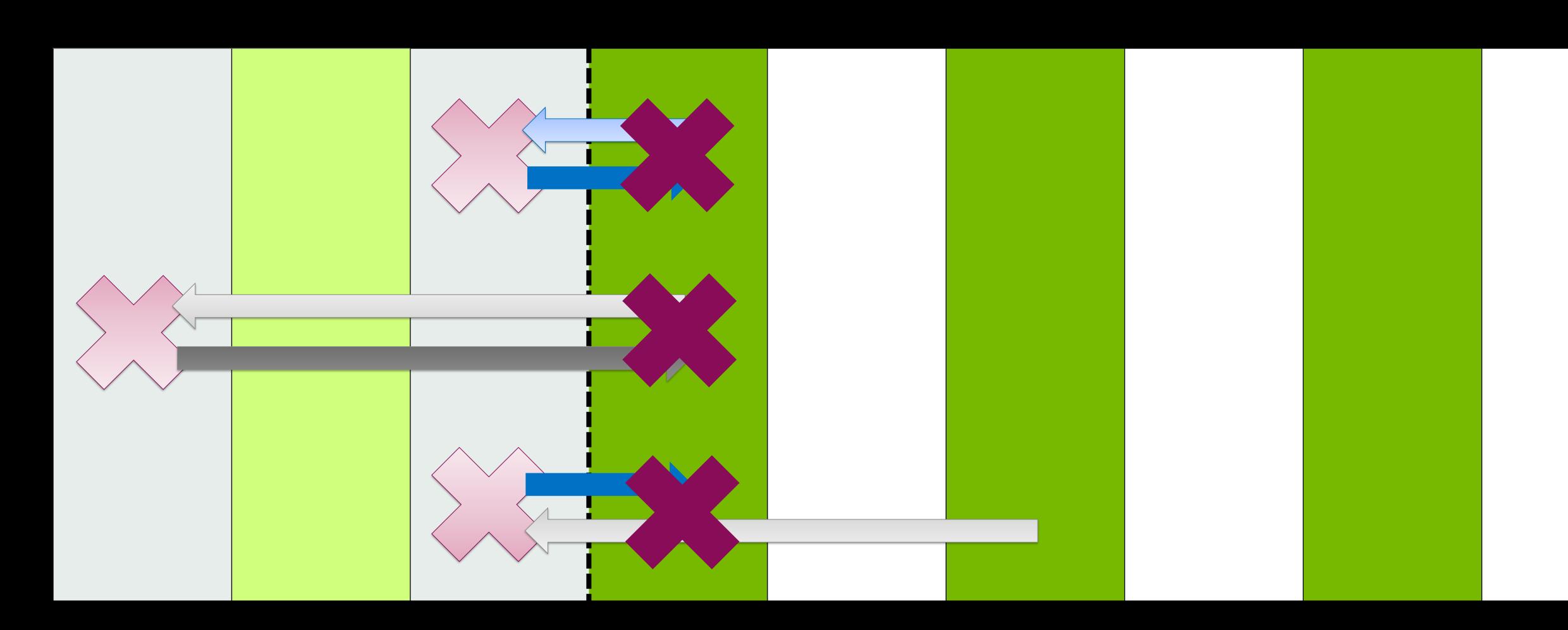
$$(x)\delta_{x,y-1} - F_{\mu}^{\dagger}(x-\hat{\mu})\delta_{x,y+1}) + (L_{\mu}(x)\delta_{x,y-3} - L_{\mu}^{\dagger})$$

# $\begin{bmatrix} 2m & D_{eo} \\ D_{oe} & 2m \end{bmatrix} \begin{bmatrix} x_e \\ x_o \end{bmatrix} = \begin{bmatrix} b_e \\ b_o \end{bmatrix}$

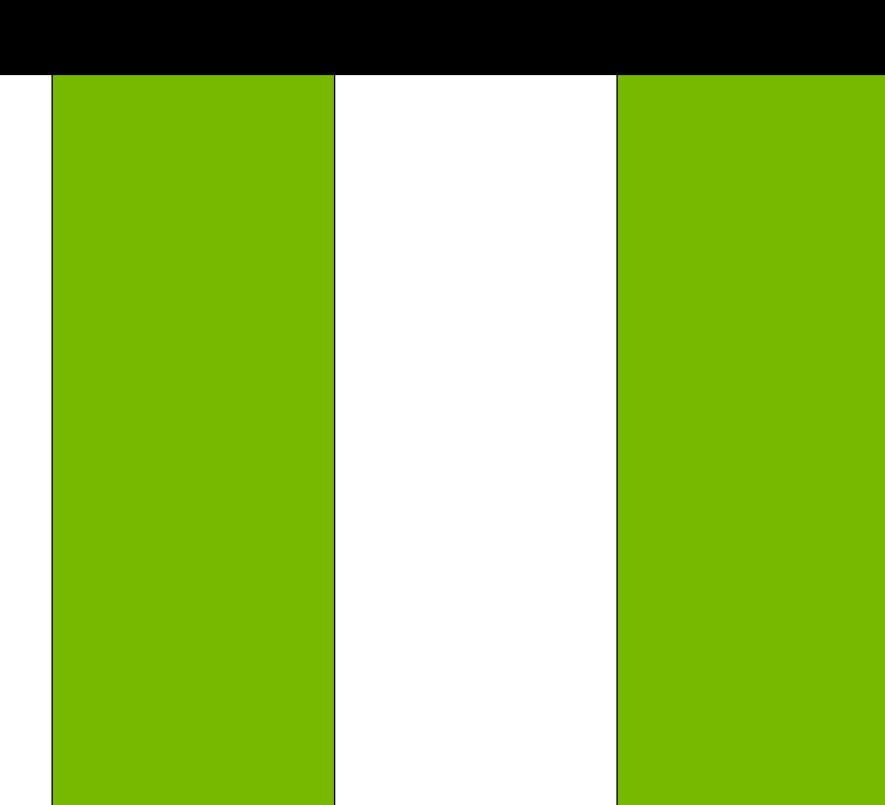
 $\frac{1}{\mu}(x-3\hat{\mu})\delta_{x,y+3}] + 2m\delta_{x,y}$ 



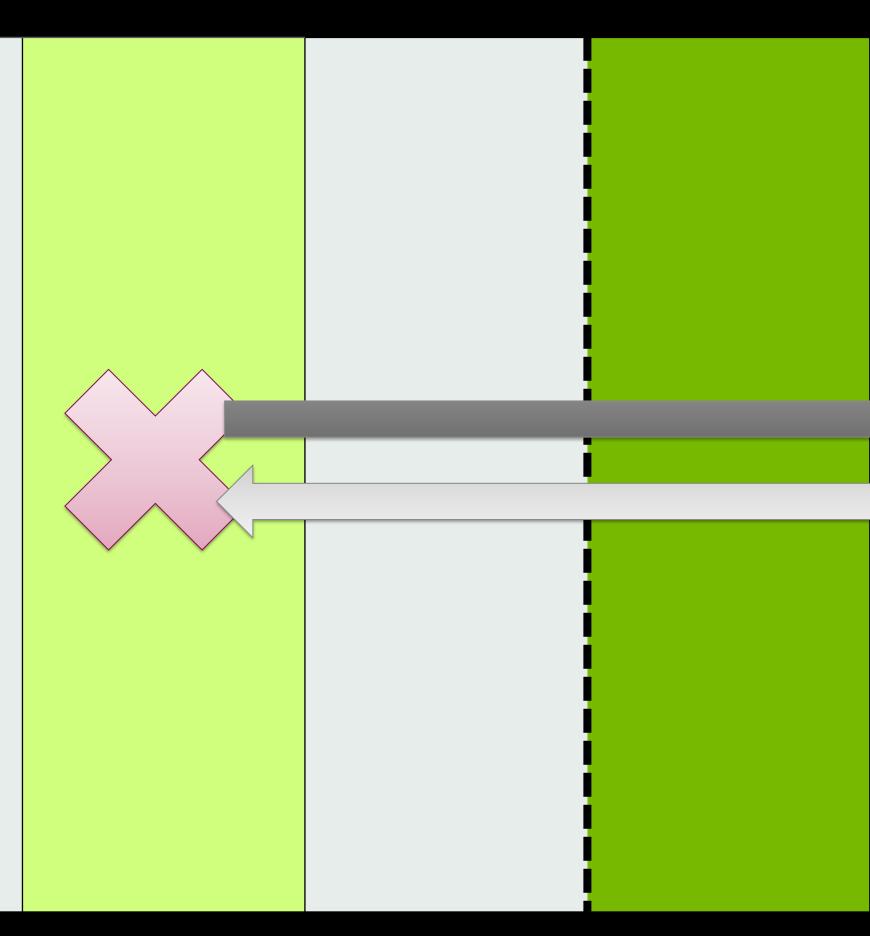
- Let's first consider the site at [0]
- There are three "boundary" contributions:
  - Start at [0]: fat link left, fat link right
  - Start at [0]: long link left, long link right
  - Start at [2]: long link left, fat link right



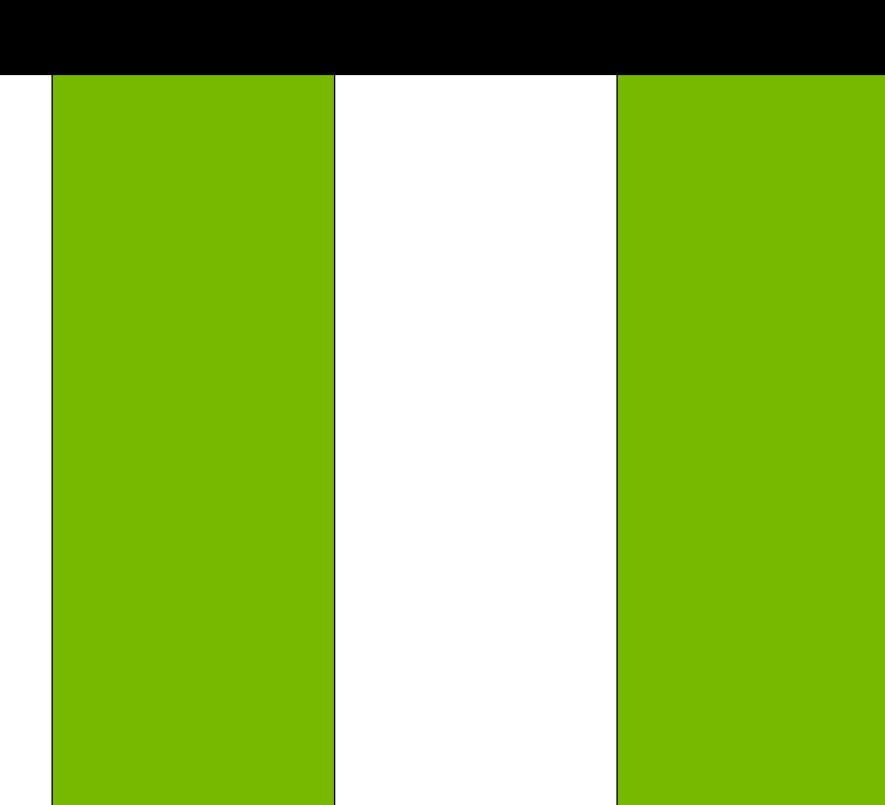
### Site Zero Three hops this time



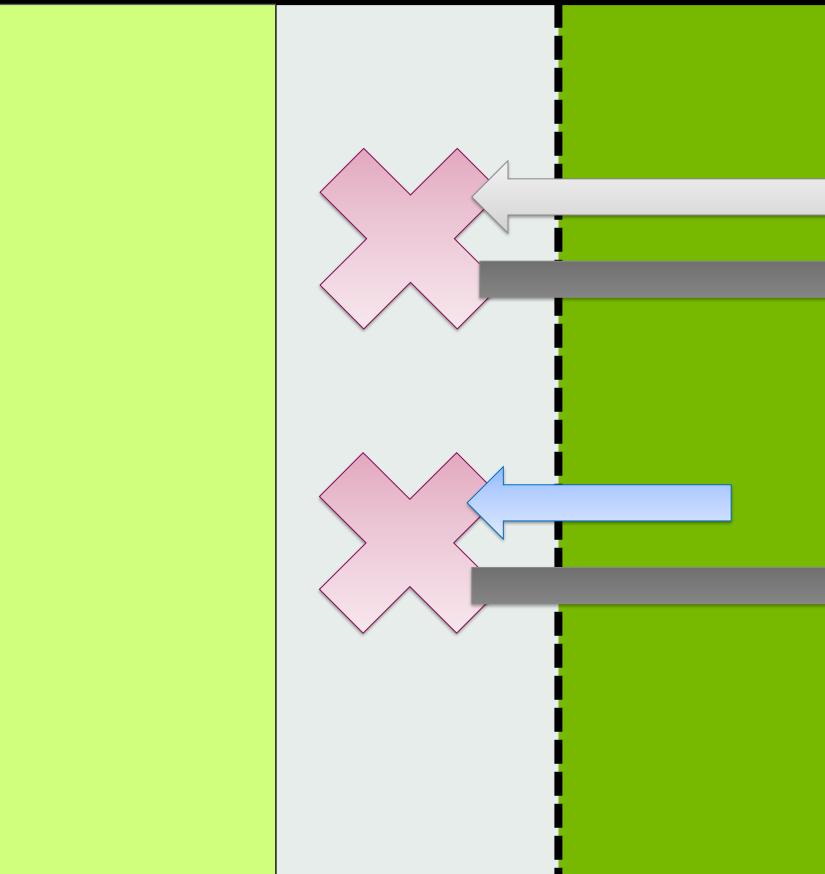
- Let's first consider the site at [1]
- There is only one boundary condition: Start at [1]: long link left, long link right



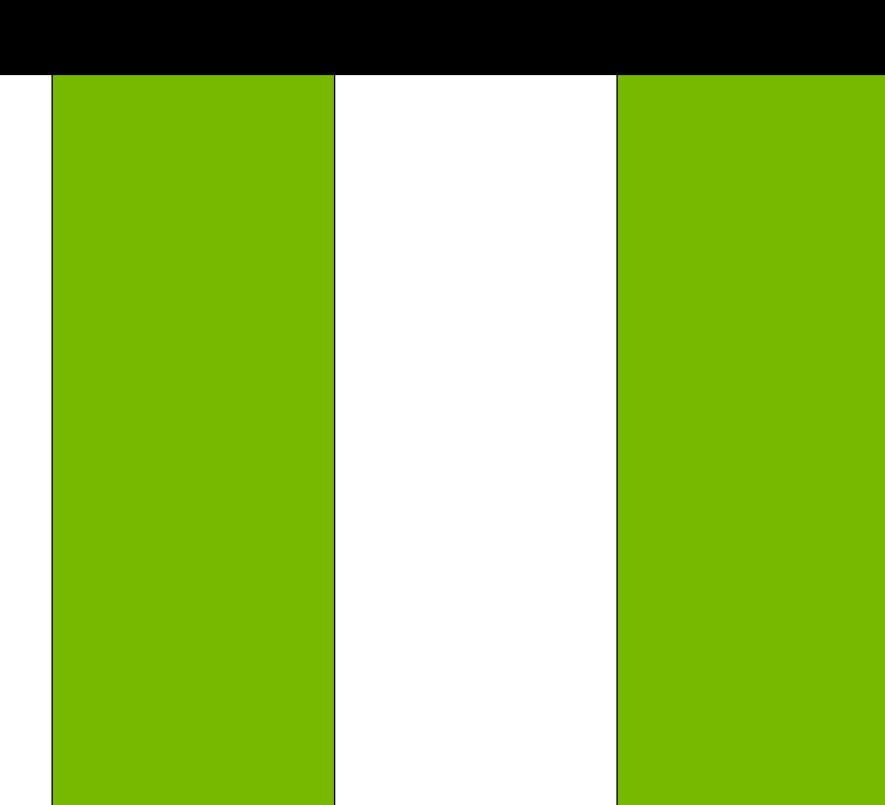
### Site One Three hops this time



- Last, we'll consider the term at [2]
- There are two boundary contributions:
  - Start at [2]: long link left, long link right
  - Start at [0]!: fat link left, long link right



### Site Two Three hops this time



• For the non-preconditioned solve, we use mixed-precision conjugate gradient (CG) with gauge link reconstruction

### Solver Workflow Solving at the speed of sound



- Reconstruction reminder:
  - The fat links are general 3x3 matrices

• For the non-preconditioned solve, we use mixed-precision conjugate gradient (CG) with gauge link reconstruction

• The long links are (proportional to) U(3) matrices, which can be represented as 9 or 13 reals



- Reconstruction reminder:
  - The fat links are general 3x3 matrices
- Mixed precision solve:
  - Outer operator: Double precision; reconstruct-13 for long links

• For the non-preconditioned solve, we use mixed-precision conjugate gradient (CG) with gauge link reconstruction

• The long links are (proportional to) U(3) matrices, which can be represented as 9 or 13 reals

• Sloppy operator: "Half" precision (QUDA's 16-bit fixed point format); reconstruct-9 for long links



- Reconstruction reminder:
  - The fat links are general 3x3 matrices
- Mixed precision solve:
  - Outer operator: Double precision; reconstruct-13 for long links
- For the preconditioned solver:
  - We use preconditioned CG (PCG) as the outer solve
  - We use fixed-iteration CG as the inner solve

• For the non-preconditioned solve, we use mixed-precision conjugate gradient (CG) with gauge link reconstruction

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- Reconstruction reminder:
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- Mixed precision solve:
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- For the preconditioned solver:
  - We use preconditioned CG (PCG) as the outer solve
  - We use fixed-iteration CG as the inner solve
- Note: PCG on paper requires a stationary preconditioner...

  - ...and seems to work well enough

• For the non-preconditioned solve, we use mixed-precision conjugate gradient (CG) with gauge link reconstruction

• The long links are (proportional to) U(3) matrices, which can be represented as 9 or 13 reals

• Sloppy operator: "Half" precision (QUDA's 16-bit fixed point format); reconstruct-9 for long links

• But with a Polak–Ribière correction, CG is "no worse than" Gradient Descent...



- Configuration:
  - NERSC Large configuration
  - Volume: 72<sup>3</sup>x144
  - Bare light mass am = 0.001





- Configuration:
  - NERSC Large configuration
  - Volume: 72<sup>3</sup>x144
  - Bare light mass am = 0.001
- Machine: Selene
  - DGX-A100-80GB nodes
  - Use 4xGPUs per node
  - 1:1 NIC ratio; HDR 200 (25 GB/s bi-directional)





- Configuration:
  - NERSC Large configuration
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  - Use 4xGPUs per node
  - 1:1 NIC ratio; HDR 200 (25 GB/s bi-directional)
- We consider multiple strong scaling problem sizes

	Nodes
8	
16	
32	
64	
128	



GPUs	Local Domain
32	364
64	36 <sup>3</sup> x18
128	36 <sup>2</sup> x18 <sup>2</sup>
256	36x18 <sup>3</sup>
512	184



- Configuration:
  - NERSC Large configuration
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  - DGX-A100-80GB nodes
  - Use 4xGPUs per node
  - 1:1 NIC ratio; HDR 200 (25 GB/s bi-directional)
- We consider multiple strong scaling problem sizes
- For networks:

  - 4:1 GPU:NIC bindings with staging through the CPU

Nodes	GPUs	Local Domain
8	32	364
16	64	36 <sup>3</sup> x18
32	128	36 <sup>2</sup> x18 <sup>2</sup>
64	256	36x18 <sup>3</sup>
128	512	184

2:1 and 1:1 direct GPU:NIC bindings to emulate different network bandwidths





- Configuration:
  - NERSC Large configuration
  - Volume: 72<sup>3</sup>x144
  - Bare light mass am = 0.001
- Machine: Selene
  - DGX-A100-80GB nodes
  - Use 4xGPUs per node
  - 1:1 NIC ratio; HDR 200 (25 GB/s bi-directional)
- We consider multiple strong scaling problem sizes
- For networks:

  - 4:1 GPU:NIC bindings with staging through the CPU
- All tests utilize NVSHMEM, implementations of the HISQ kernel
  - Device-driven communications

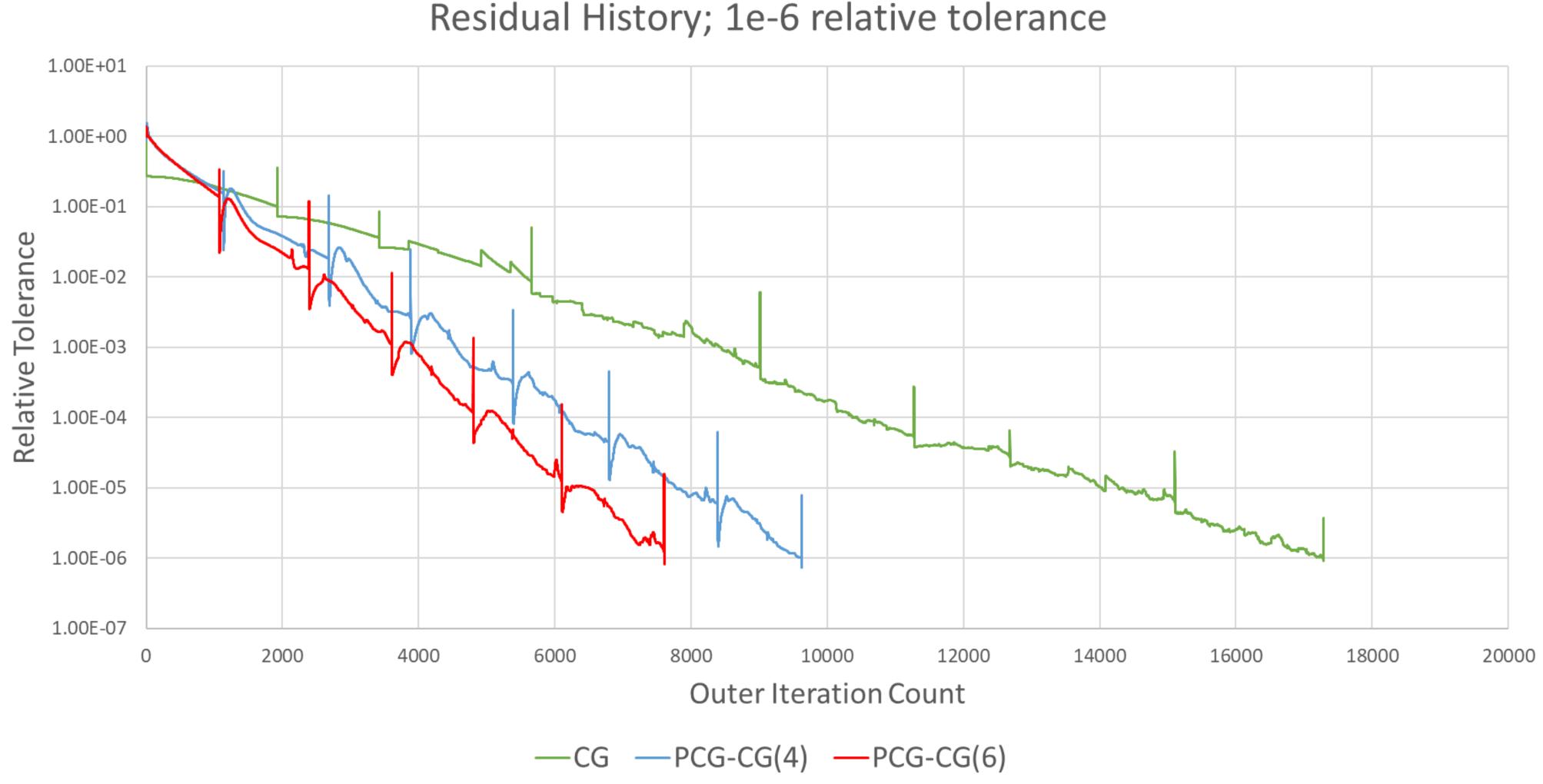
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2:1 and 1:1 direct GPU:NIC bindings to emulate different network bandwidths

• Reduces latency: no separate packing kernel, no overhead of MPI calls, gets the host out of the way







- CG and PCG each converge in a stable fashion
- (P)CG solve

### **Convergence History** An unstable algorithm is pointless

NERSC Large: 72<sup>3</sup>x144, m=0.001 128 Selene nodes, 4xGPU/node Residual History; 1e-6 relative tolerance

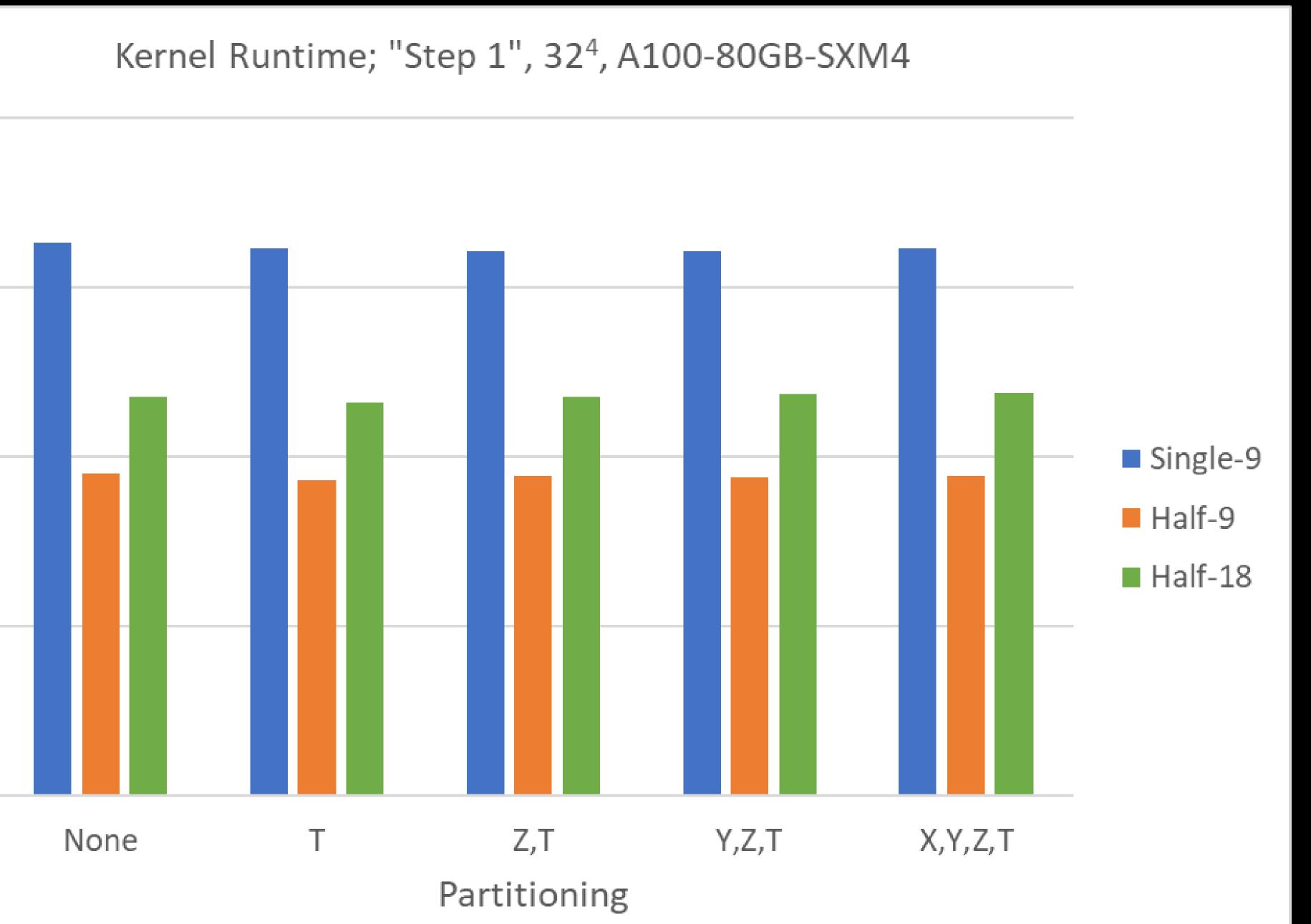
• The "spikes" are due to residual updates: "every so often" we recompute the exact residual and re-inject it into the





- 4.00E-04
- 3.00E-04
- (s) Time 2.00E-04
  - 1.00E-04
  - 0.00E+00

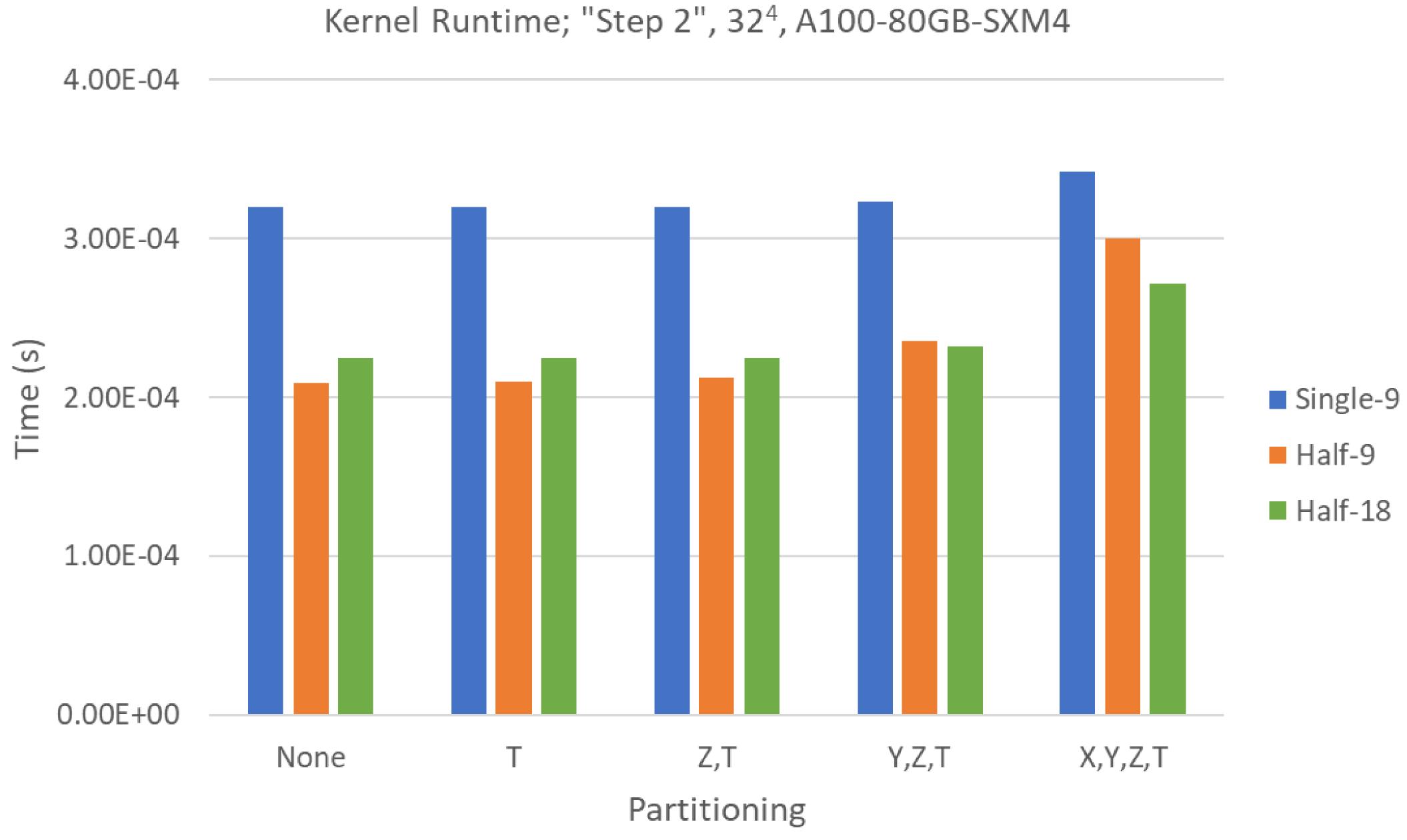
# **Operator Performance: Zero Boundary Conditions**



Performance is essentially independent of the partitioning This makes sense: all we're doing is "snipping" away work



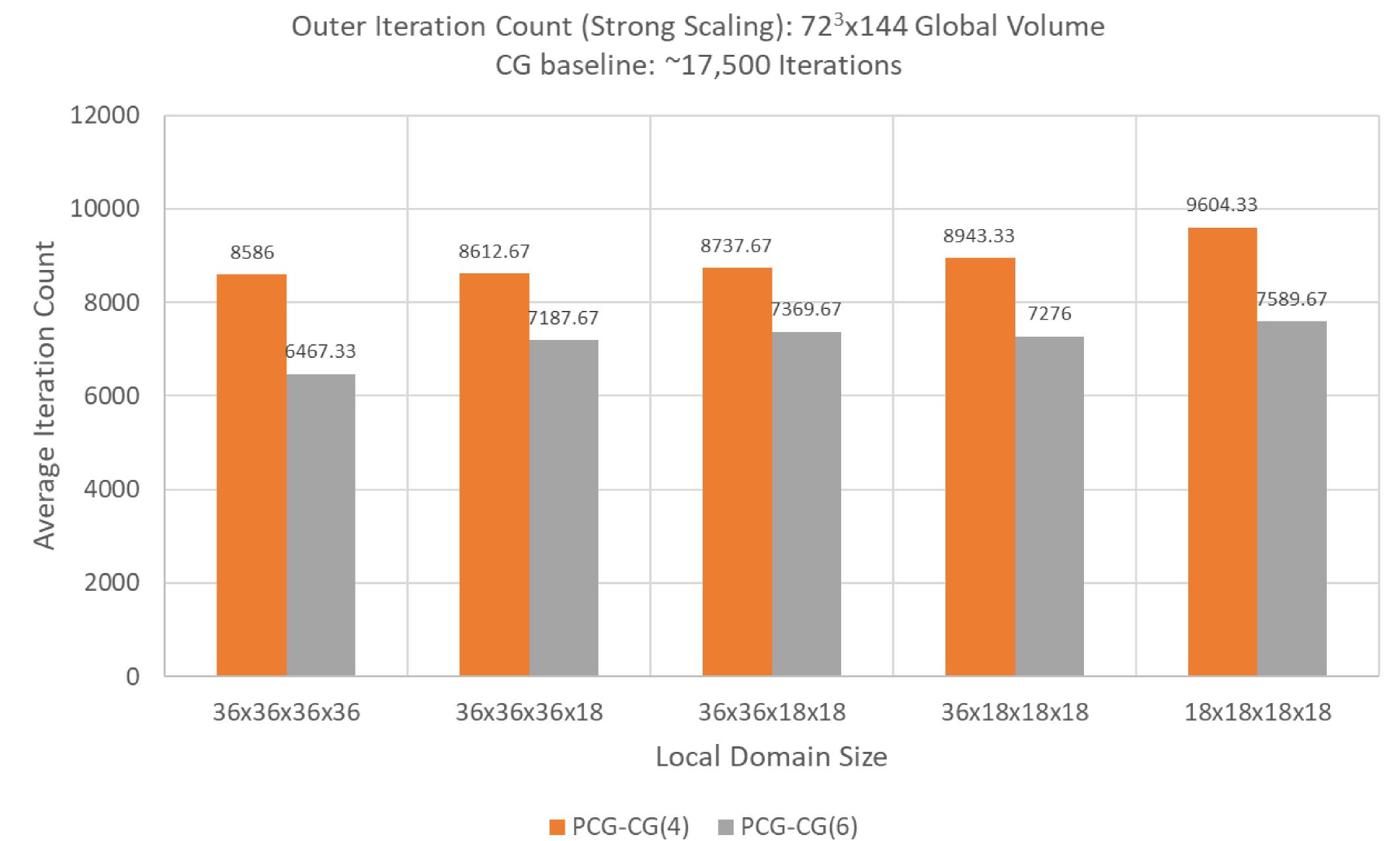
# **Operator Performance: Boundary Clovers**



Performance decreases with partitioning This makes sense: we're adding (divergent) work Extra note: reconstruct becomes a *detriment:* extra instructions hold up threads



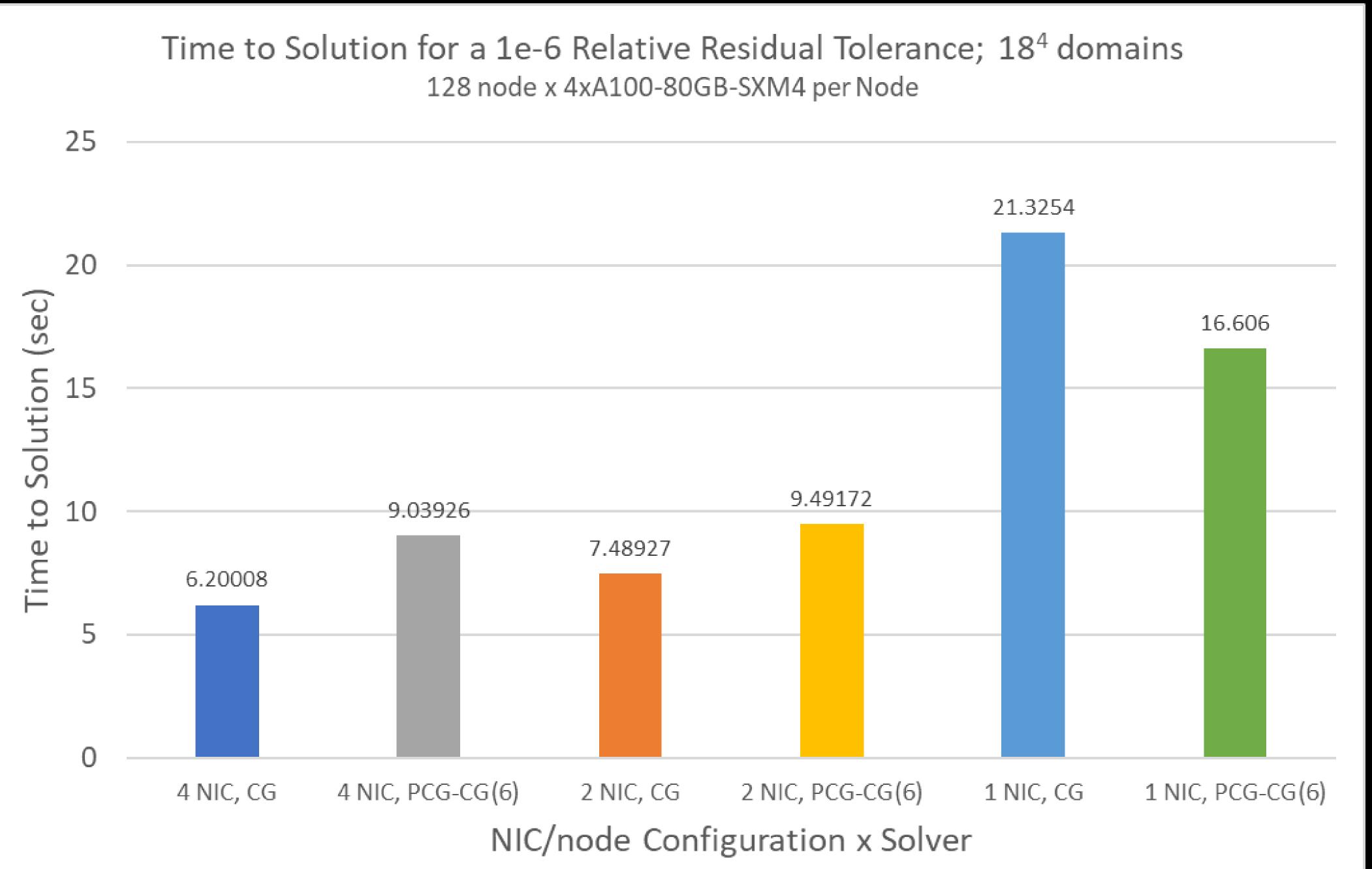
# **Iteration Counts for each Preconditioner**



More preconditioner iterations -> fewer outer iterations (to a point) Diminishing benefit with smaller partition sizes -> domain is a lower-quality approximation of full domain



# Time to Solution (which is all that matters)



Note: 1xNIC includes CPU staging for two GPUs to access a NIC! There's still outstanding work to be done when the network is strong (25 GB/s bi-directional per NIC)... ...but we also see that the preconditioner is beneficial when the network is slow



# Future Work



## HISQ Force: no further optimizations

- latencies
- HISQ MG + Schwarz Preconditioner:

  - implementations
- •...192<sup>3</sup>x384 ensemble

### **Future HISQy Business** Same old song and dance

Schwarz Preconditioner: Pre-computed matrix products to reduce

 Use the local operator as a smoother on all levels • Outer HISQ and Kahler-Dirac preconditioned operator have GPU code

Even/odd preconditioned coarse operators do not





