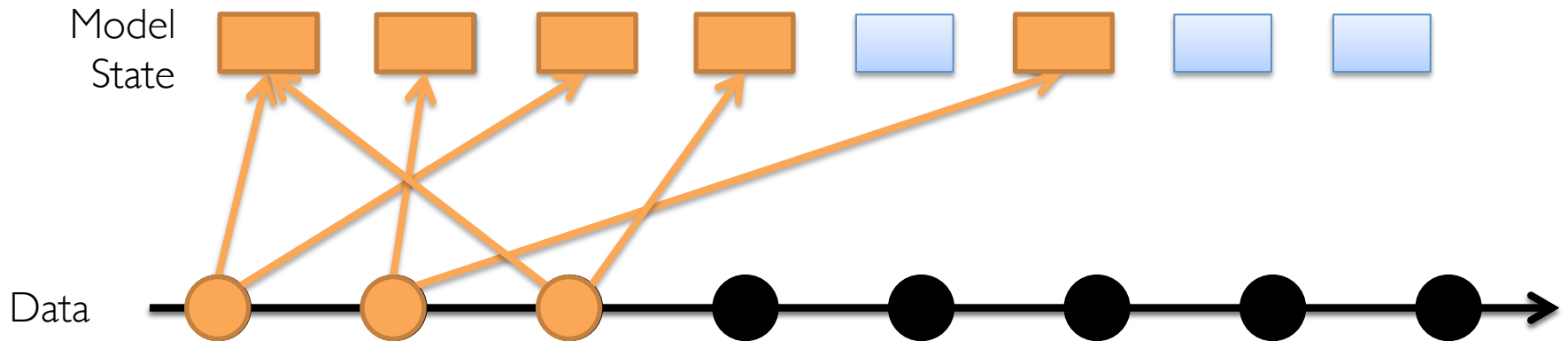


Optimistic Concurrency Control in the Design and Analysis of Parallel Learning Algorithms

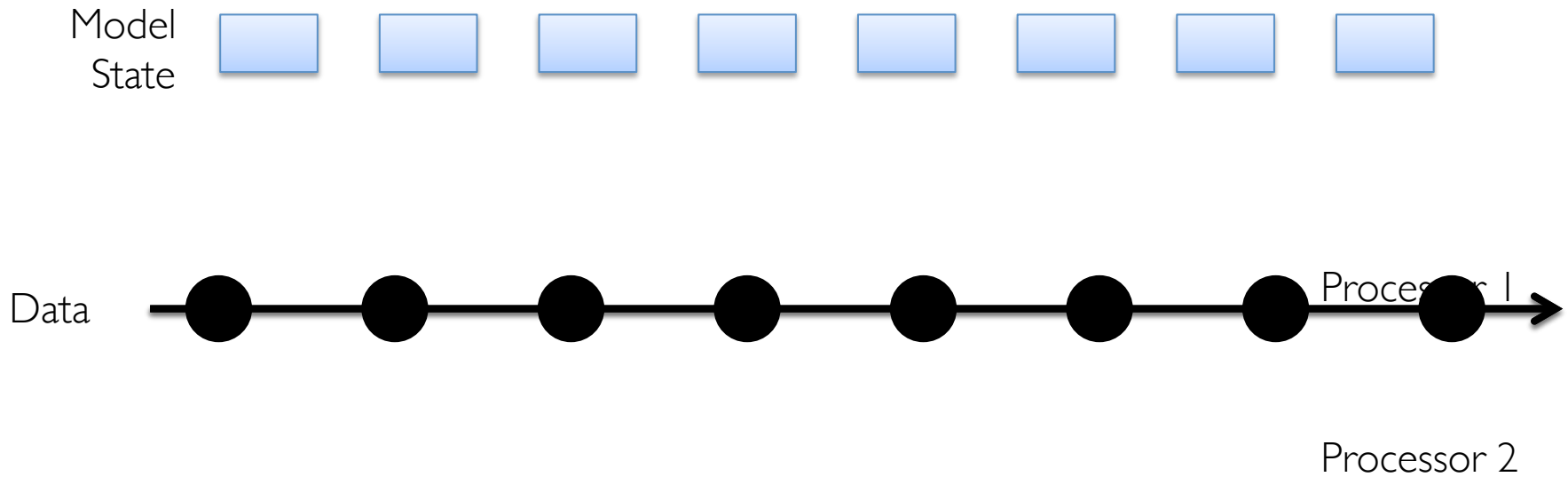


Joseph E. Gonzalez
Postdoc, UC Berkeley AMPLab
Co-founder, GraphLab Inc.
jegonzal@eecs.berkeley.edu

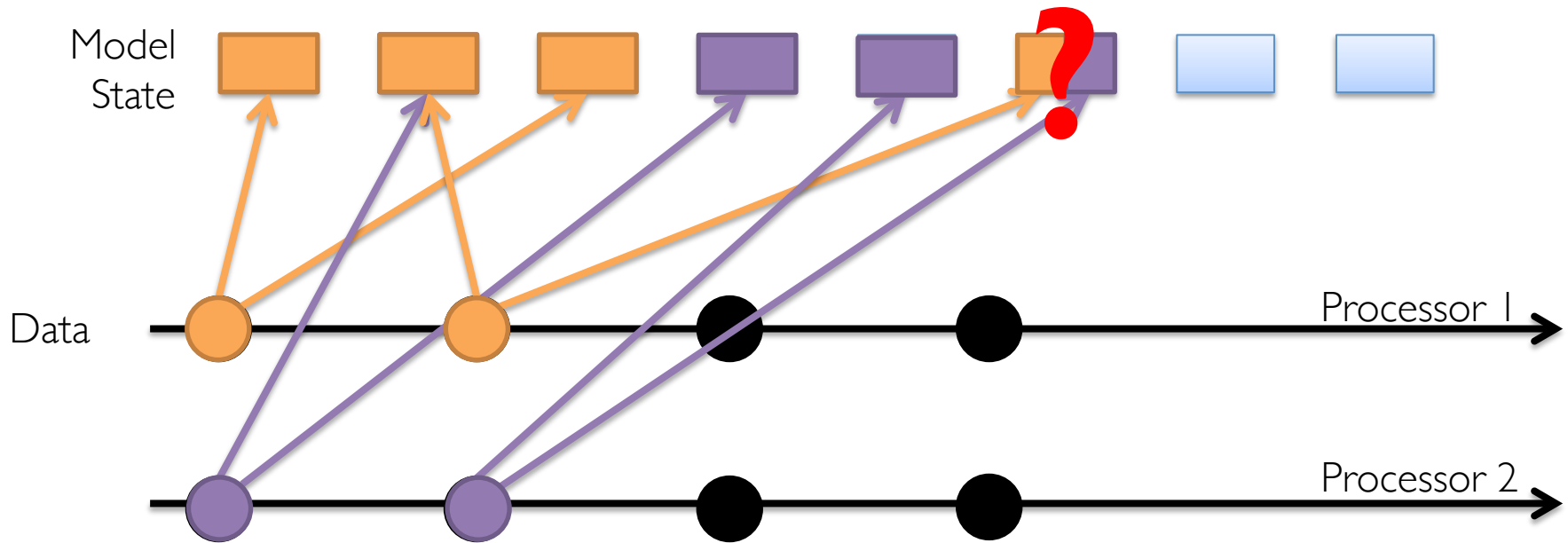
Serial Inference



Parallel Inference



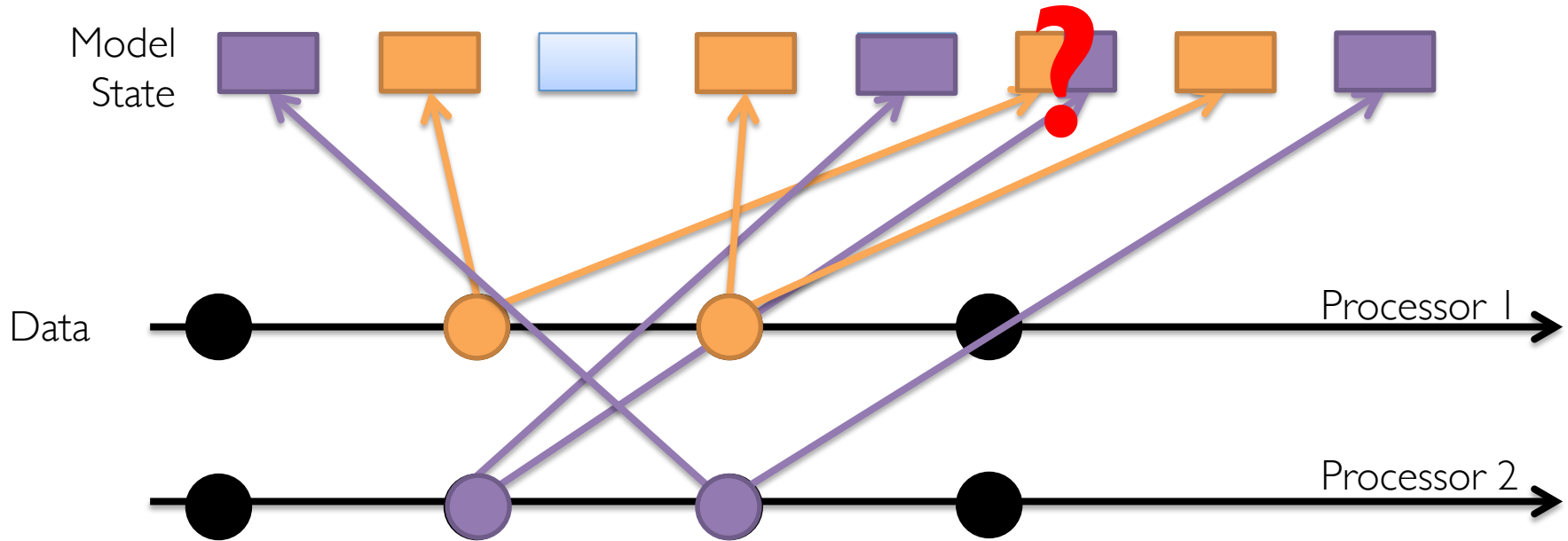
Parallel Inference



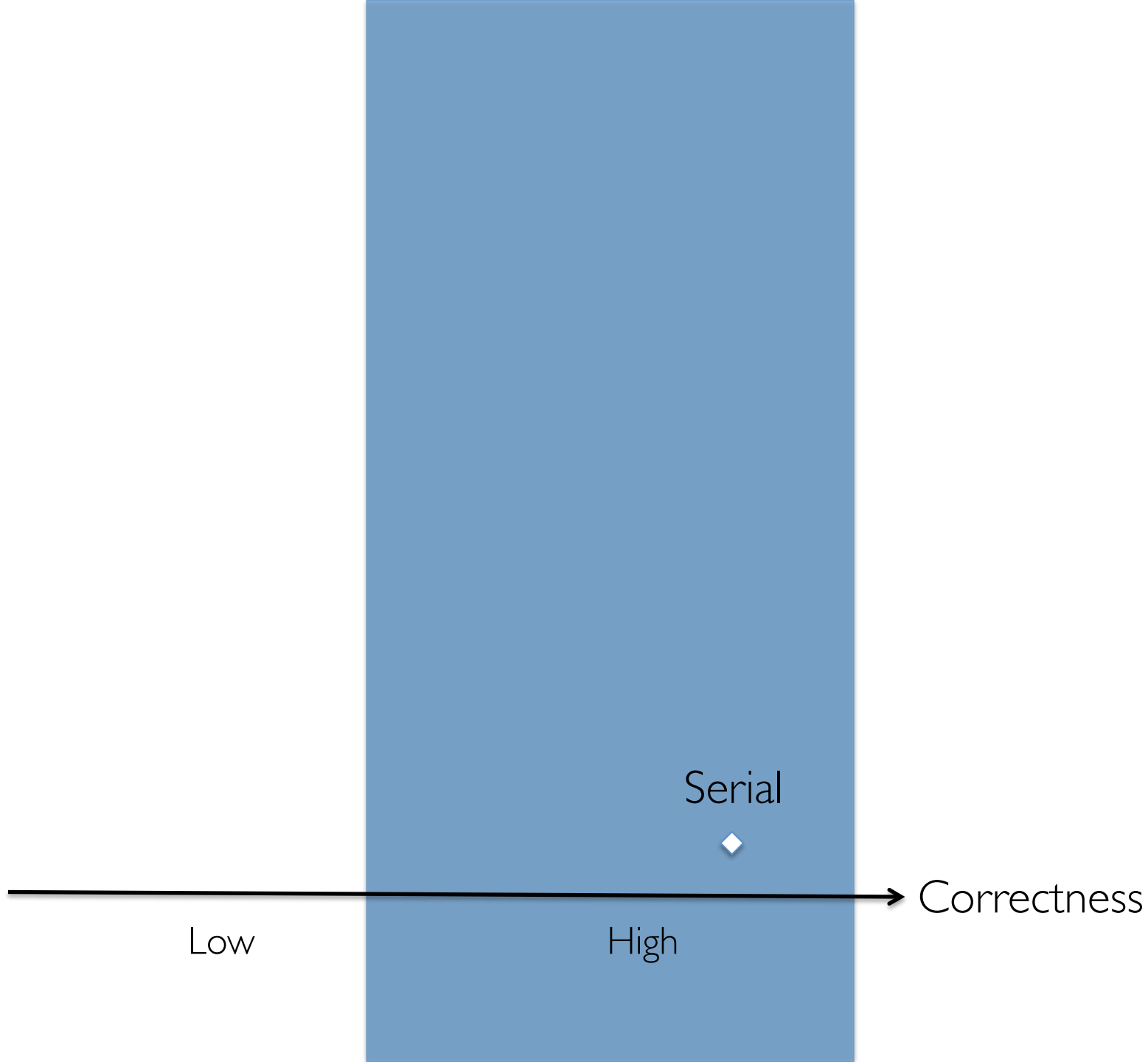
Correctness:
serial equivalence

Concurrency:
more machines = less time

Coordination Free Parallel Inference



Correctness and Consistency:
Depends on Assumptions (almost) free



Concurrency

High

Low

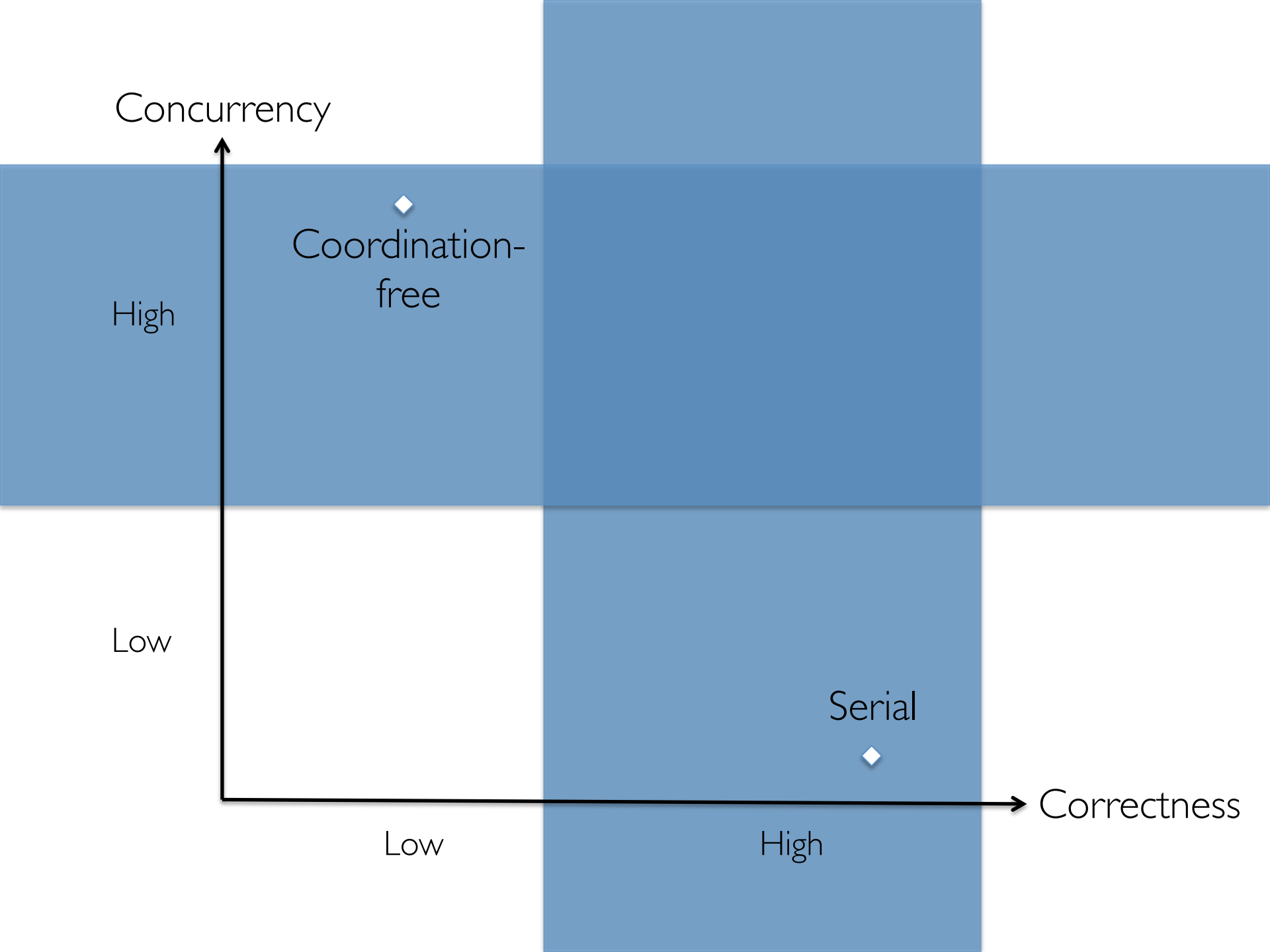
Coordination-
free

Low

High

Serial

Correctness



Concurrency

High

Low

Coordination-free

Concurrency Control

Database mechanisms

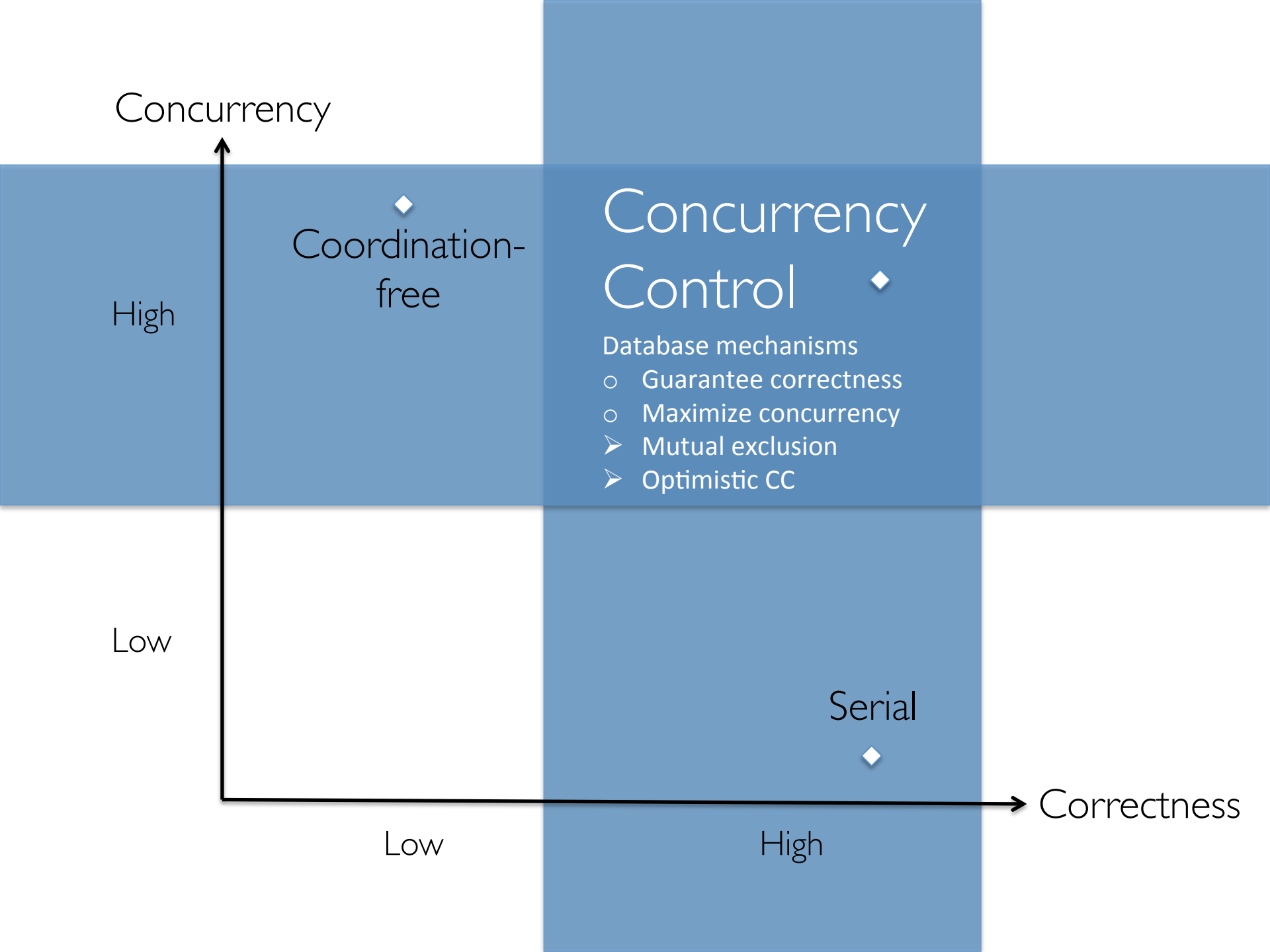
- Guarantee correctness
- Maximize concurrency
- Mutual exclusion
- Optimistic CC

Serial

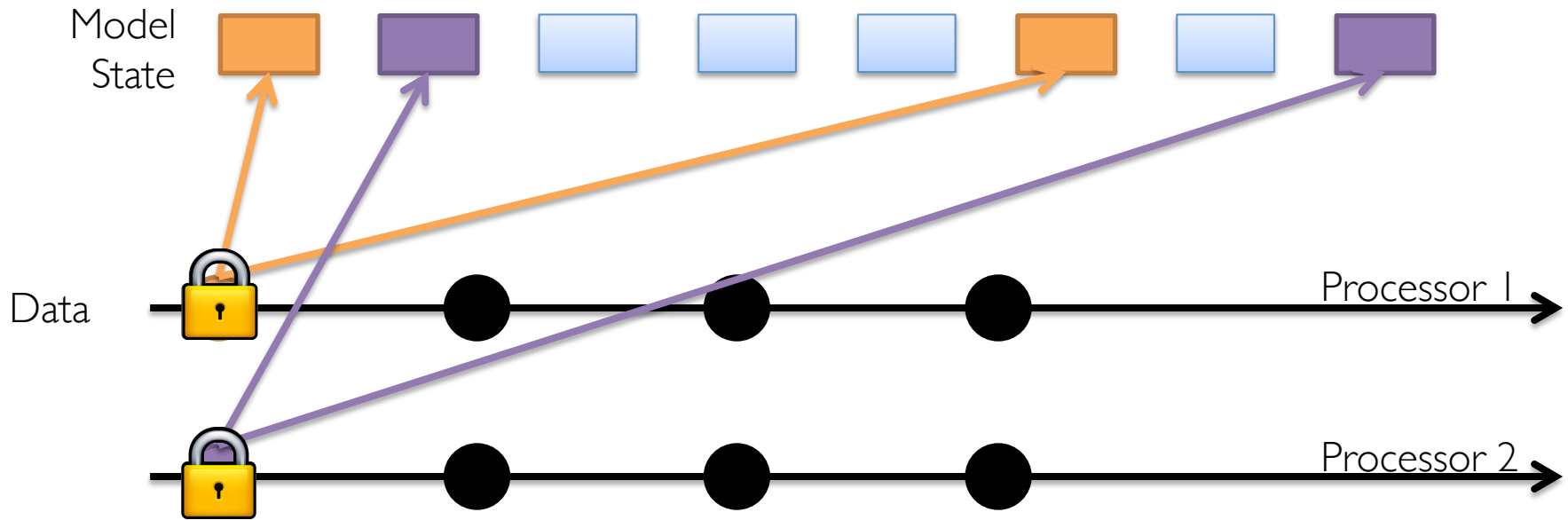
Low

High

Correctness

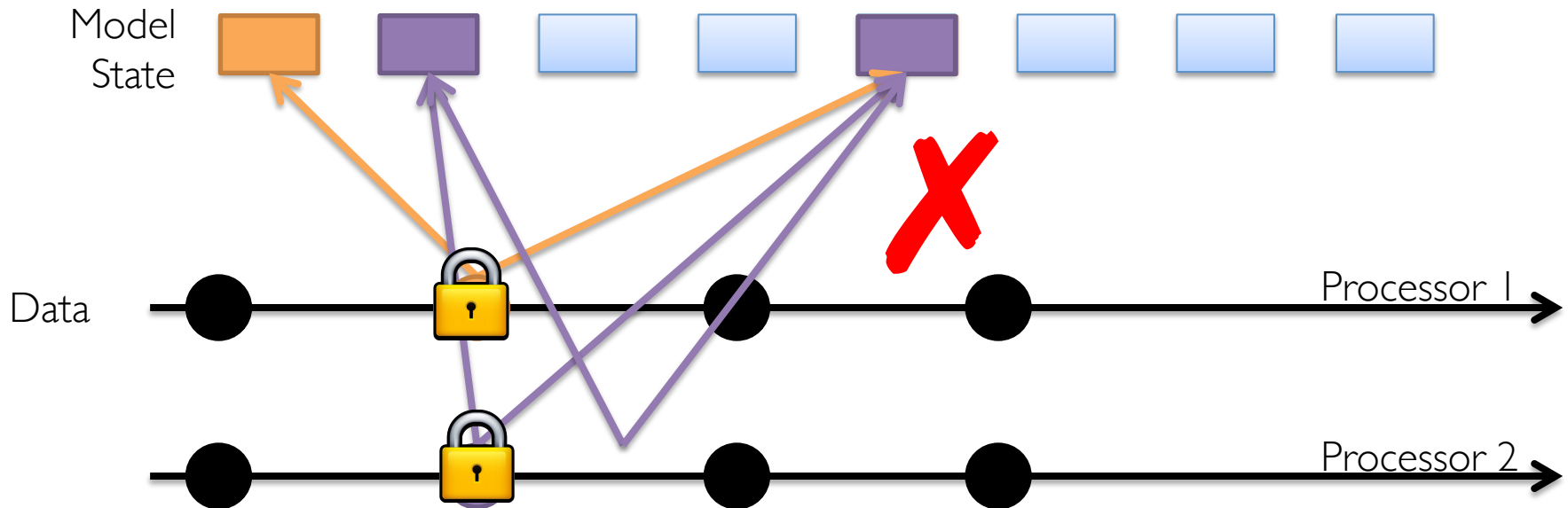


Mutual Exclusion Through Locking



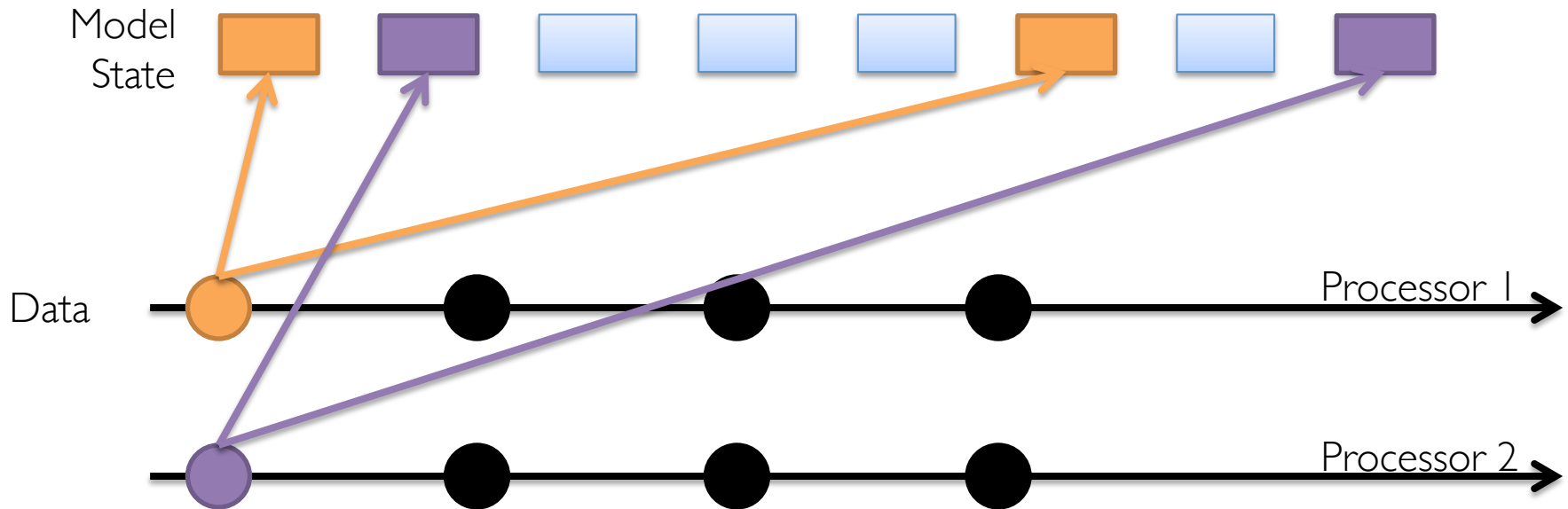
Introducing locking (scheduling) protocols to prevent potential conflicts.

Mutual Exclusion Through Locking



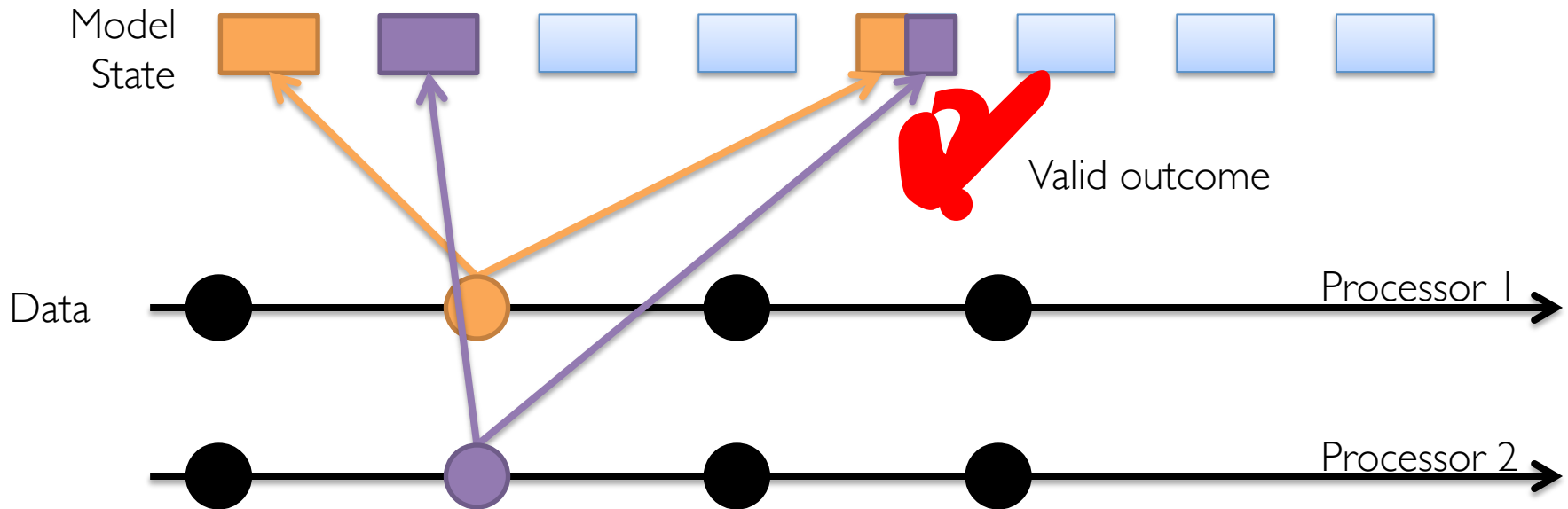
Enforce serialization of computation that could conflict.

Optimistic Concurrency Control



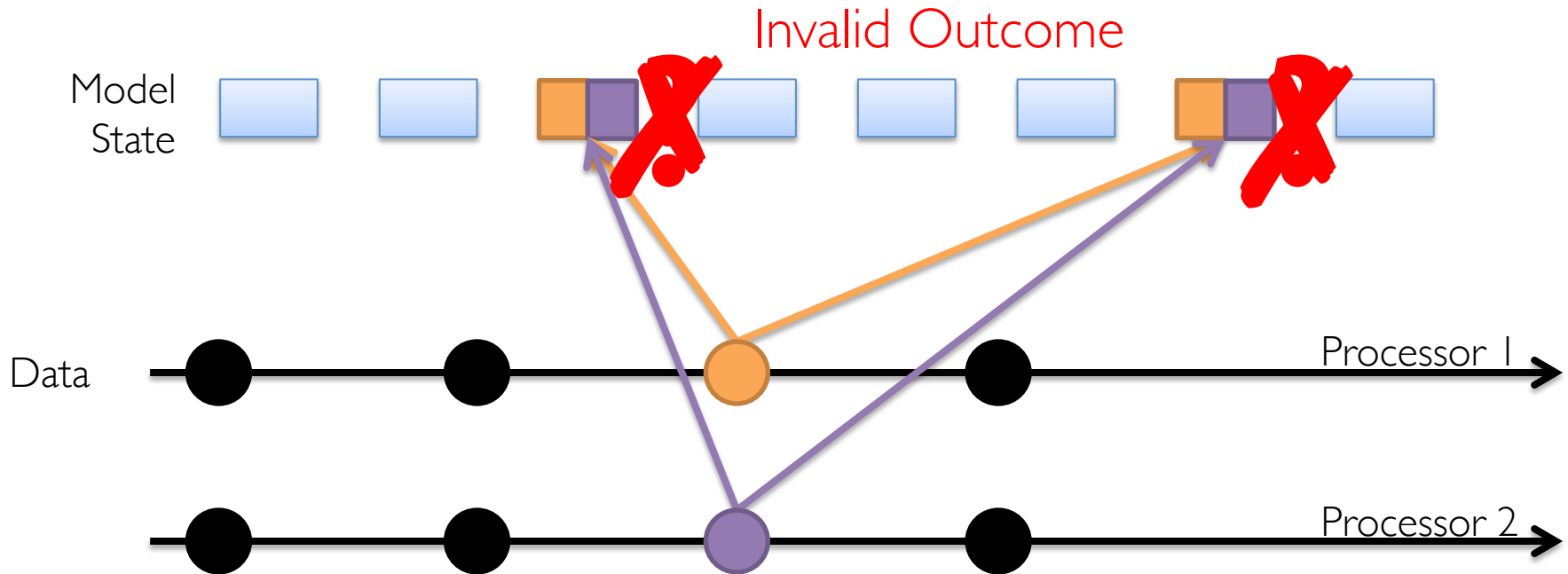
Allow computation to proceed without blocking.

Optimistic Concurrency Control



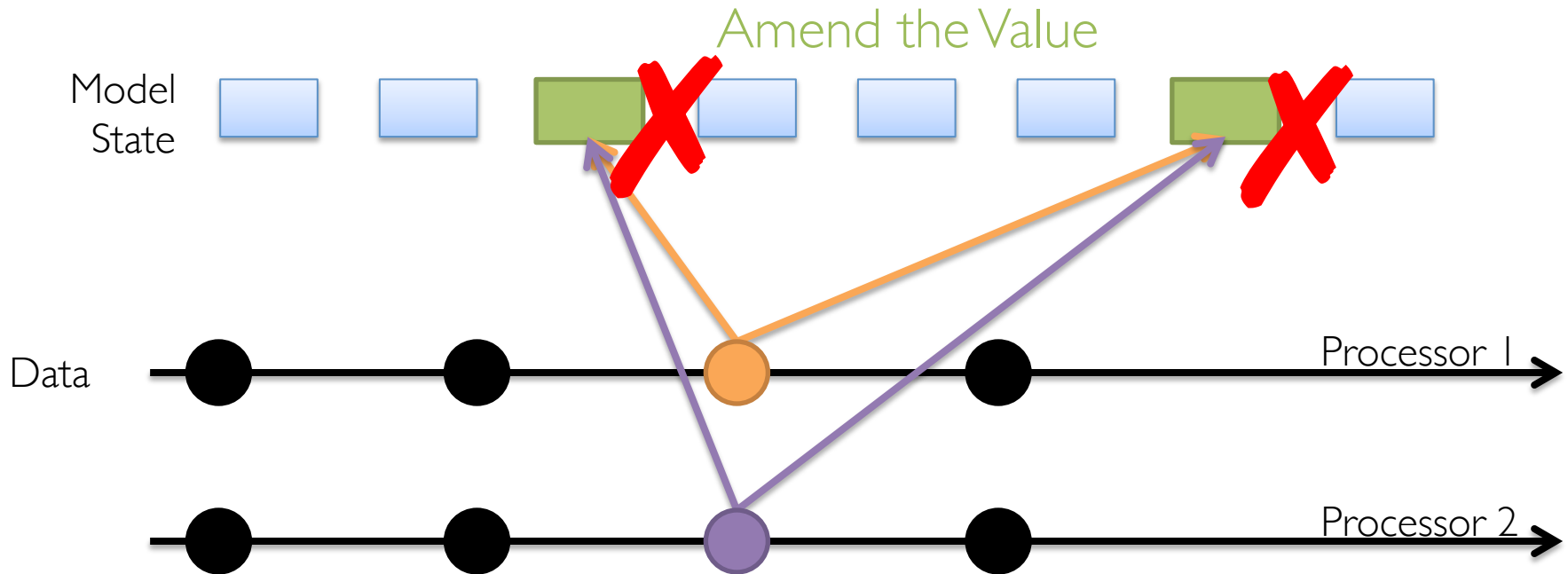
Validate potential conflicts.

Optimistic Concurrency Control



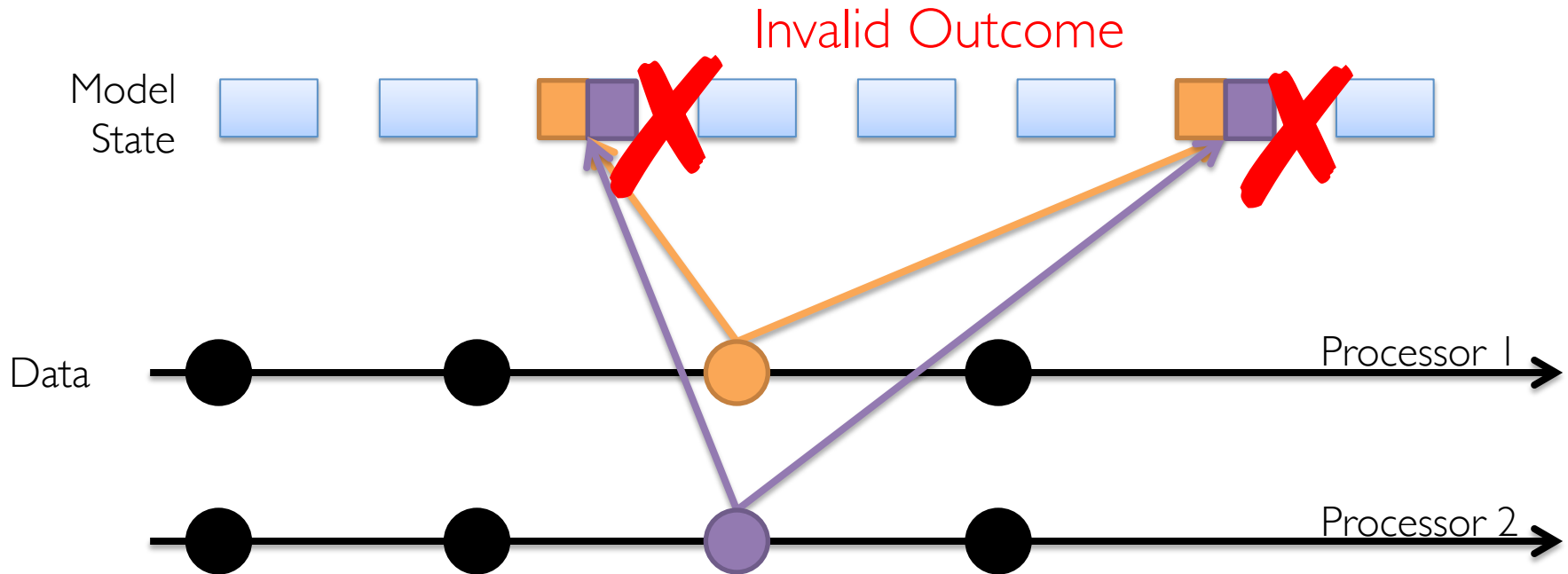
Validate potential conflicts.

Optimistic Concurrency Control



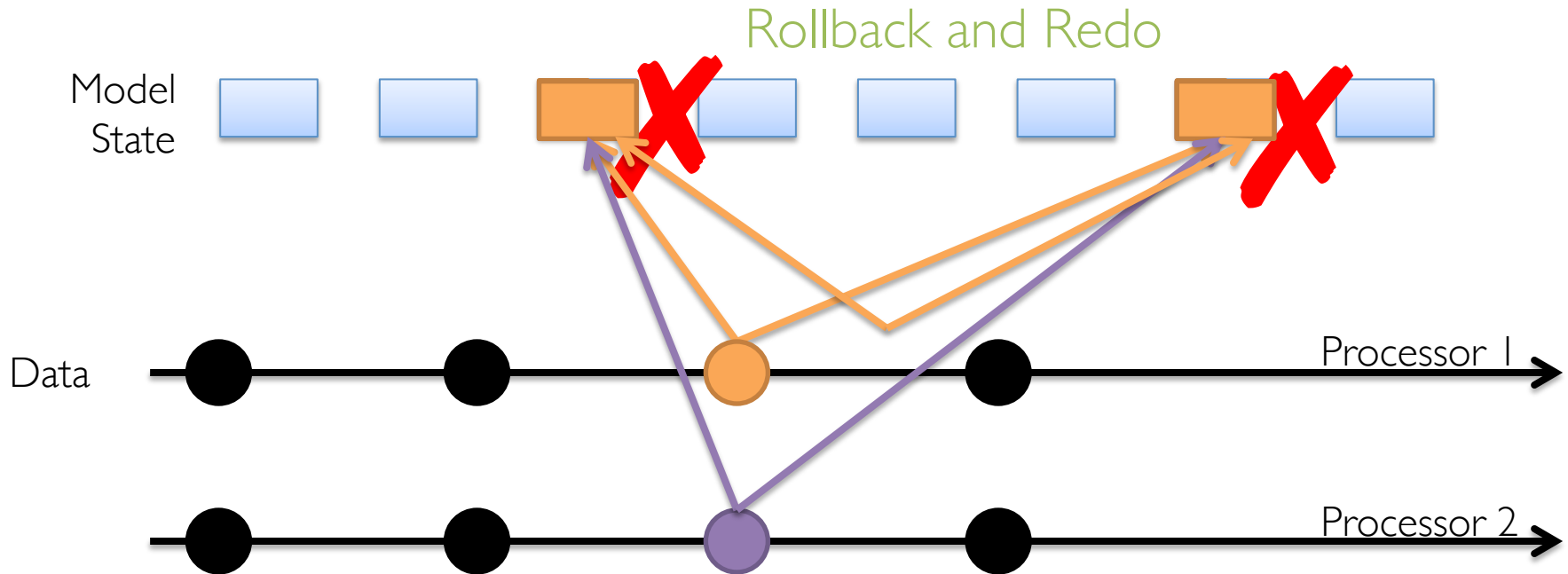
Take a compensating action.

Optimistic Concurrency Control



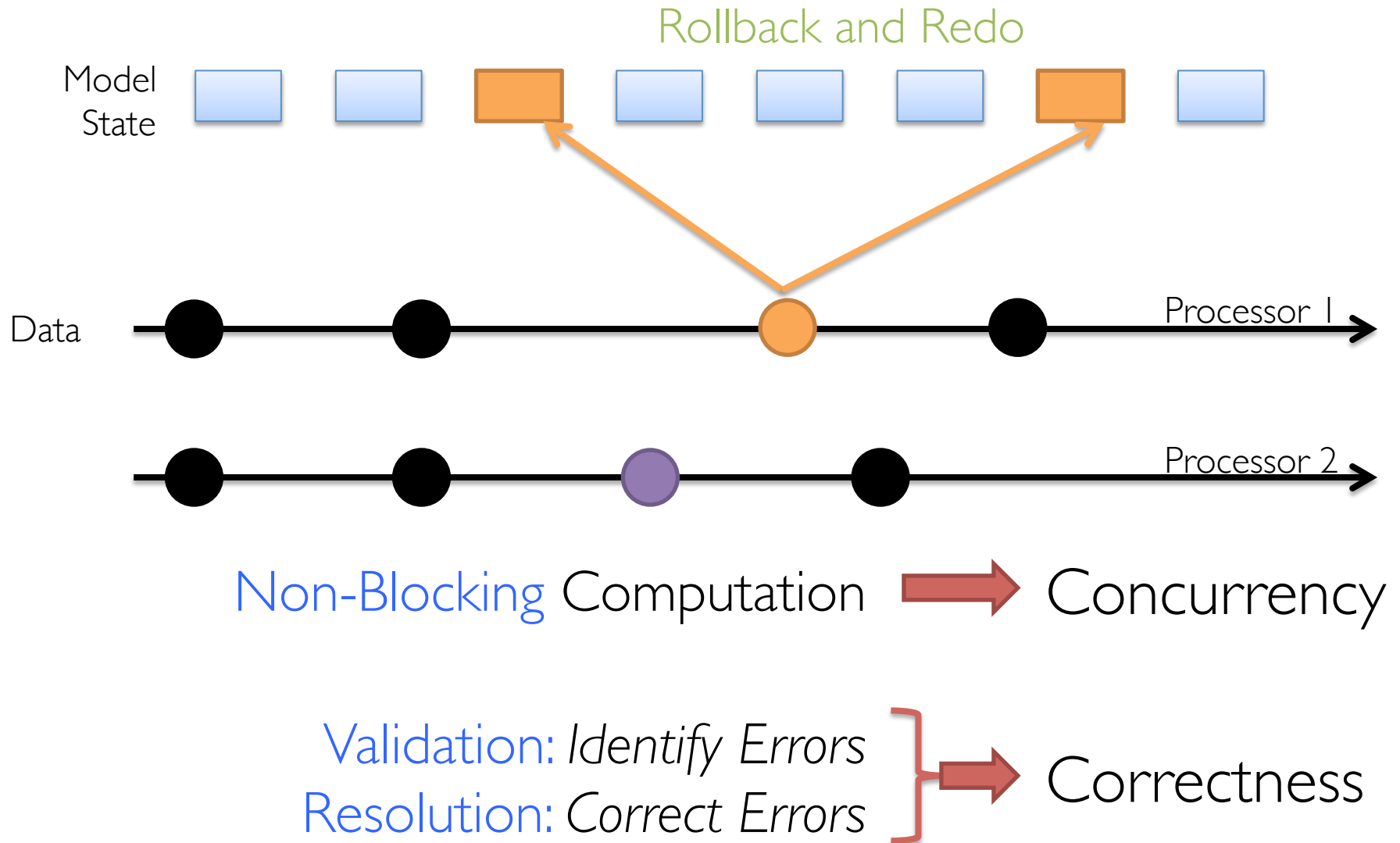
Validate potential conflicts.

Optimistic Concurrency Control



Take a compensating action.

Optimistic Concurrency Control

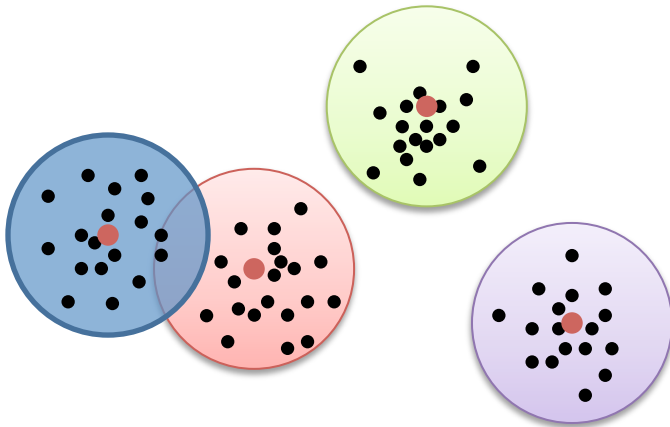


Optimistic Concurrency Control for Machine Learning

Non-parametric Clustering

Distributed DP-Means

[NIPS'13]



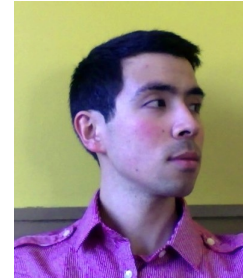
Submodular Optimization

Double Greedy Submodular Maximization

[NIPS'14]



Optimistic Concurrency Control for Submodular Maximization



Xinghao Pan, Stefanie Jegelka, Joseph Gonzalez, Joseph Bradley, Michael I. Jordan

Submodular Set Functions

Diminishing Returns Property

$F: 2^V \rightarrow \mathbb{R}$, such that for all $A \subset B \subseteq \mathcal{V}$ and $e \notin B$

$$F(A \cup e) - F(A) \geq F(B \cup e) - F(B)$$



Submodular Examples

Sensing



$F(S) = \text{area covered by } S$

$F(S) = I(Y; X_S)$

= reduction in uncertainty

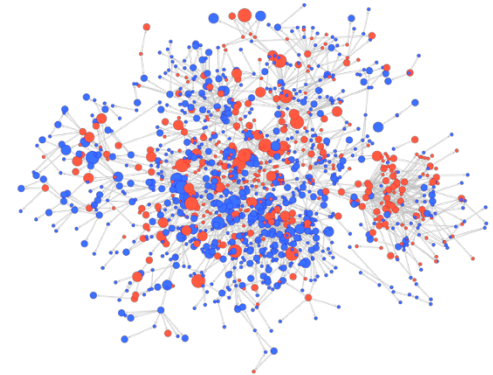
(Krause & Guestrin 2005)

Document Summarization



(Lin & Bilmes 2011)

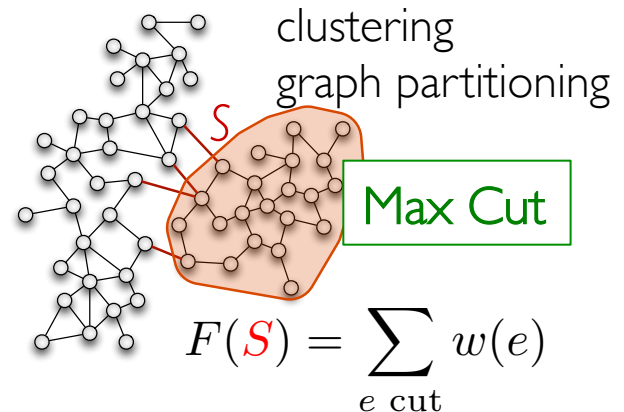
Network Analysis



$F(S) = \mathbb{E}[\# \text{ active nodes at end}]$

(Kempe, Kleinberg, Tardos 2003,
Mossel & Roch 2007)

Graph Algorithms

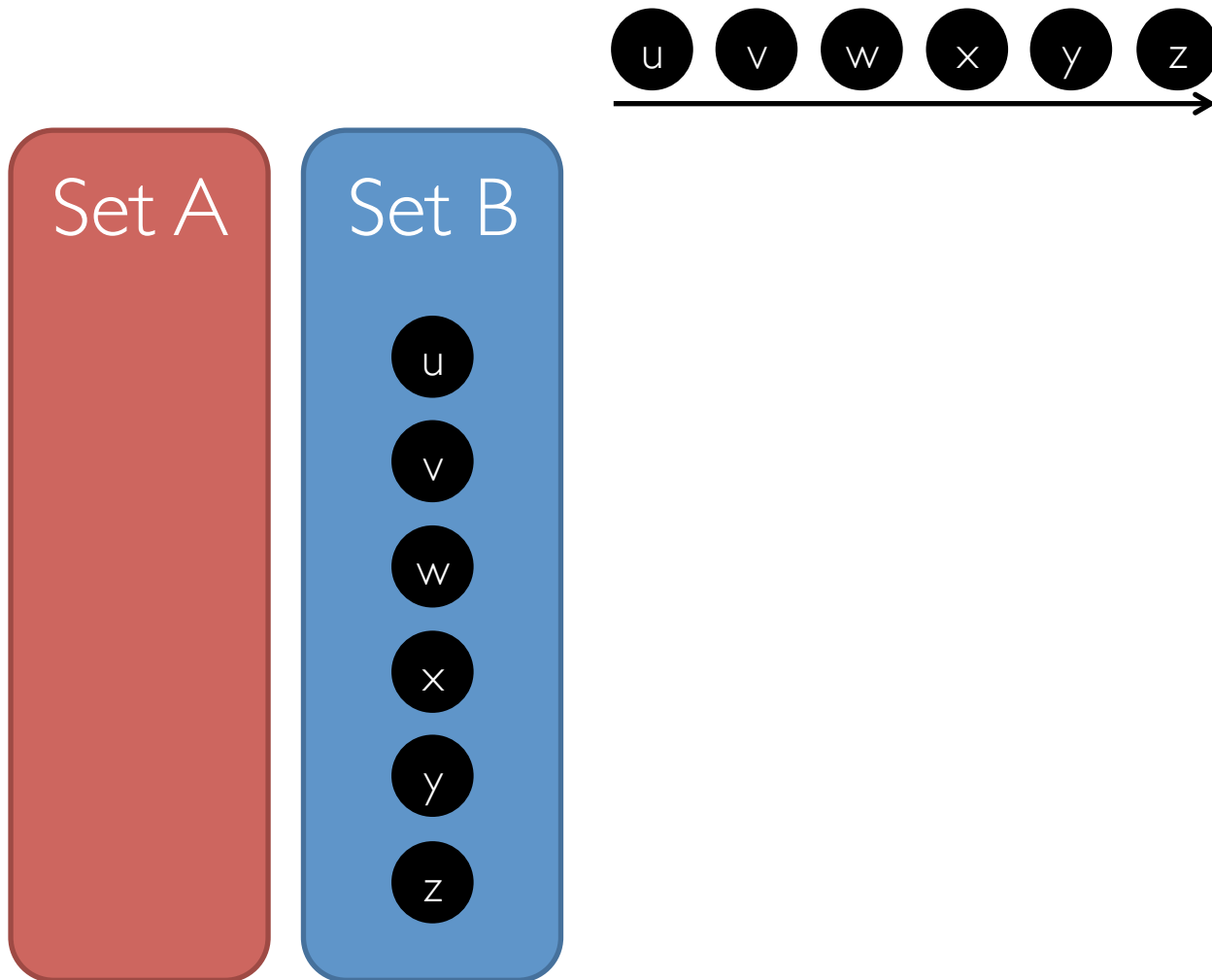


Submodular Maximization

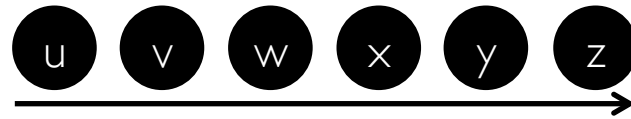
$$\max F(A), A \subseteq V$$

		Monotone (increasing) functions [Positive marginal gains]	Non-monotone functions
Sequential / Parallel / Distributed	Sequential	<i>Greedy (Nemhauser et al, 1978)</i> (1 - 1/e) - approximation Optimal polytime	<i>Double Greedy (Buchbinder et al, 2012)</i> 1/2 - approximation Optimal polytime
	Parallel / Distributed	<i>GREEDI (Mirzasoleiman et al, 2013)</i> (1 - 1/e) ² / p – approximation 1 MapReduce round <i>(Kumar et al, 2013)</i> 1 / (2 + ε) – approximation O(1 / ε) MapReduce rounds	<i>Concurrency Control Double Greedy</i> Optimal 1/2 - approximation Bounded overhead <i>Coordination Free Double Greedy</i> Bounded error Minimal overhead

Double Greedy Algorithm



Double Greedy Algorithm



Set A

Set B

u

v

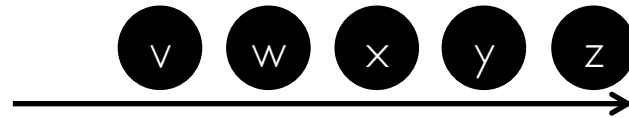
w

x

y

z

Double Greedy Algorithm



Set A

Set B

u
v
w
x
y
z

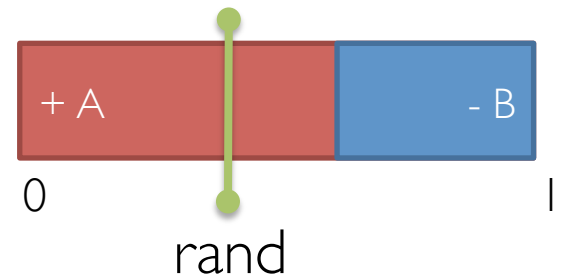
u

Marginal gains

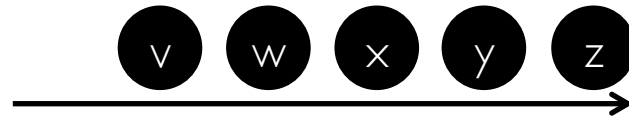
$$\Delta_+(u|A) = F(A \cup u) - F(A),$$

$$\Delta_-(u|B) = F(B \setminus u) - F(B).$$

$$p(u | A, B) =$$



Double Greedy Algorithm



Set A

Set B

u
v
w
x
y
z

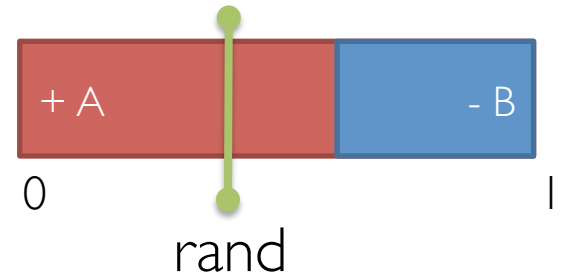
u

Marginal gains

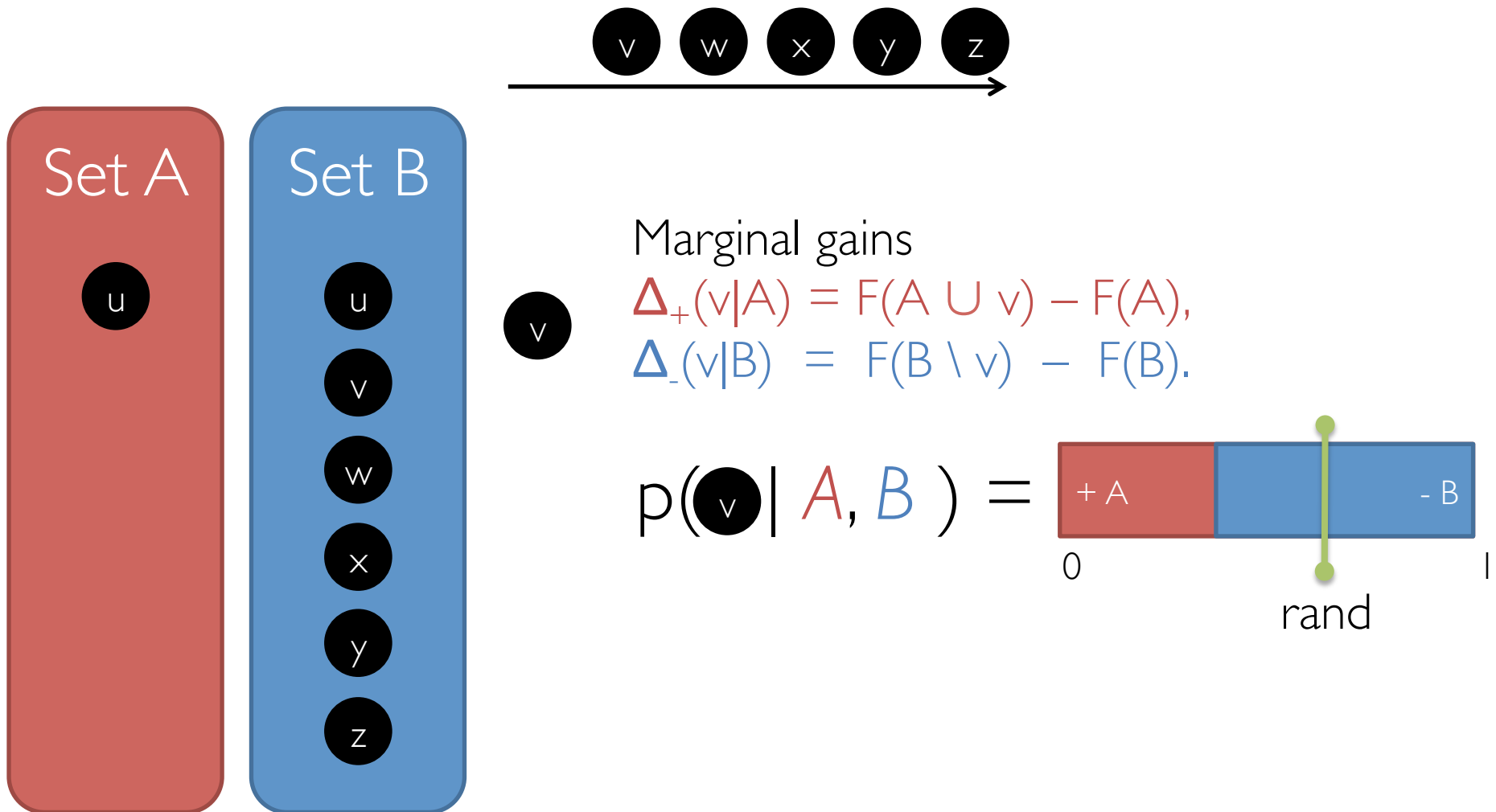
$$\Delta_+(u|A) = F(A \cup u) - F(A),$$

$$\Delta_-(u|B) = F(B \setminus u) - F(B).$$

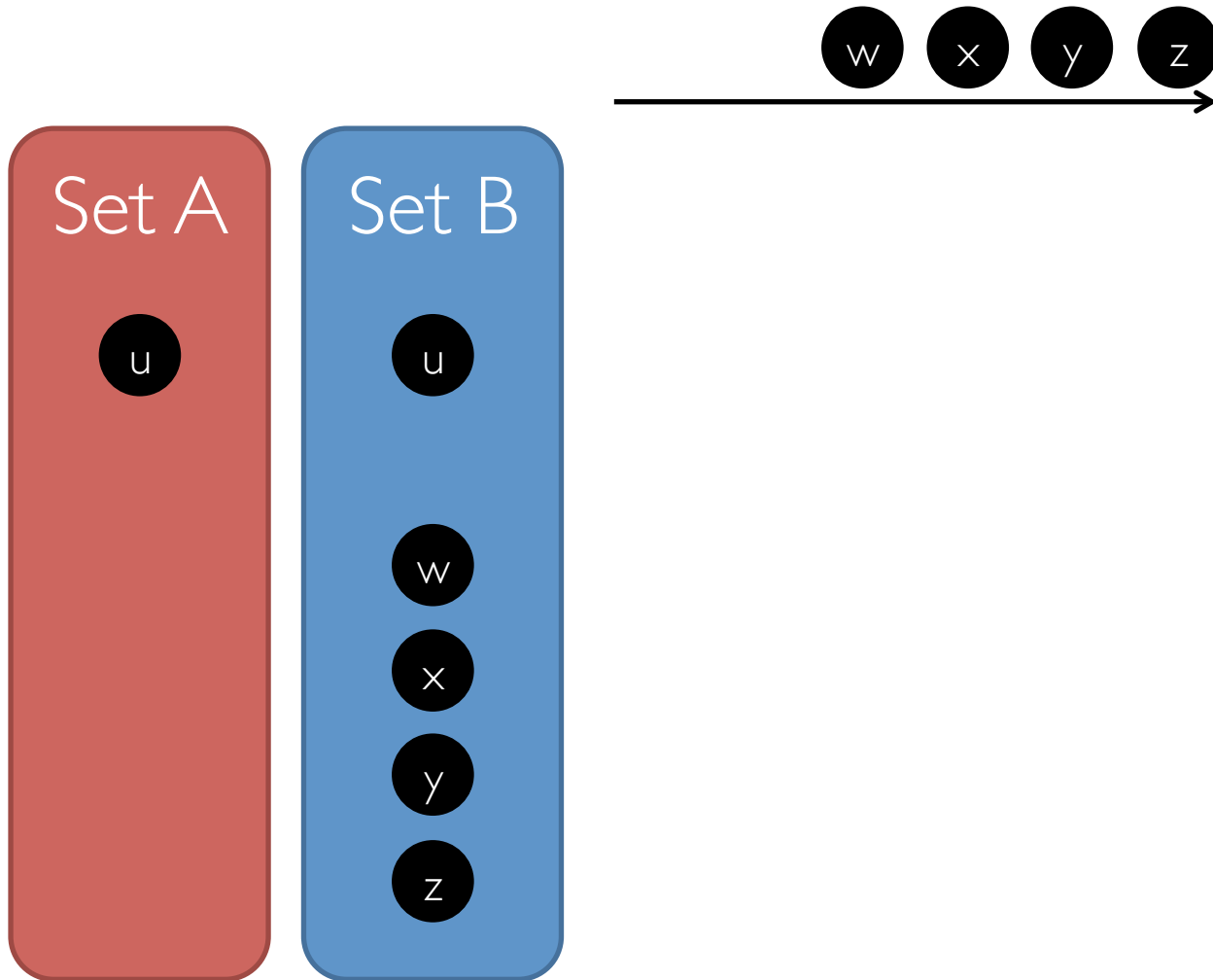
$$p(u | A, B) =$$



Double Greedy Algorithm

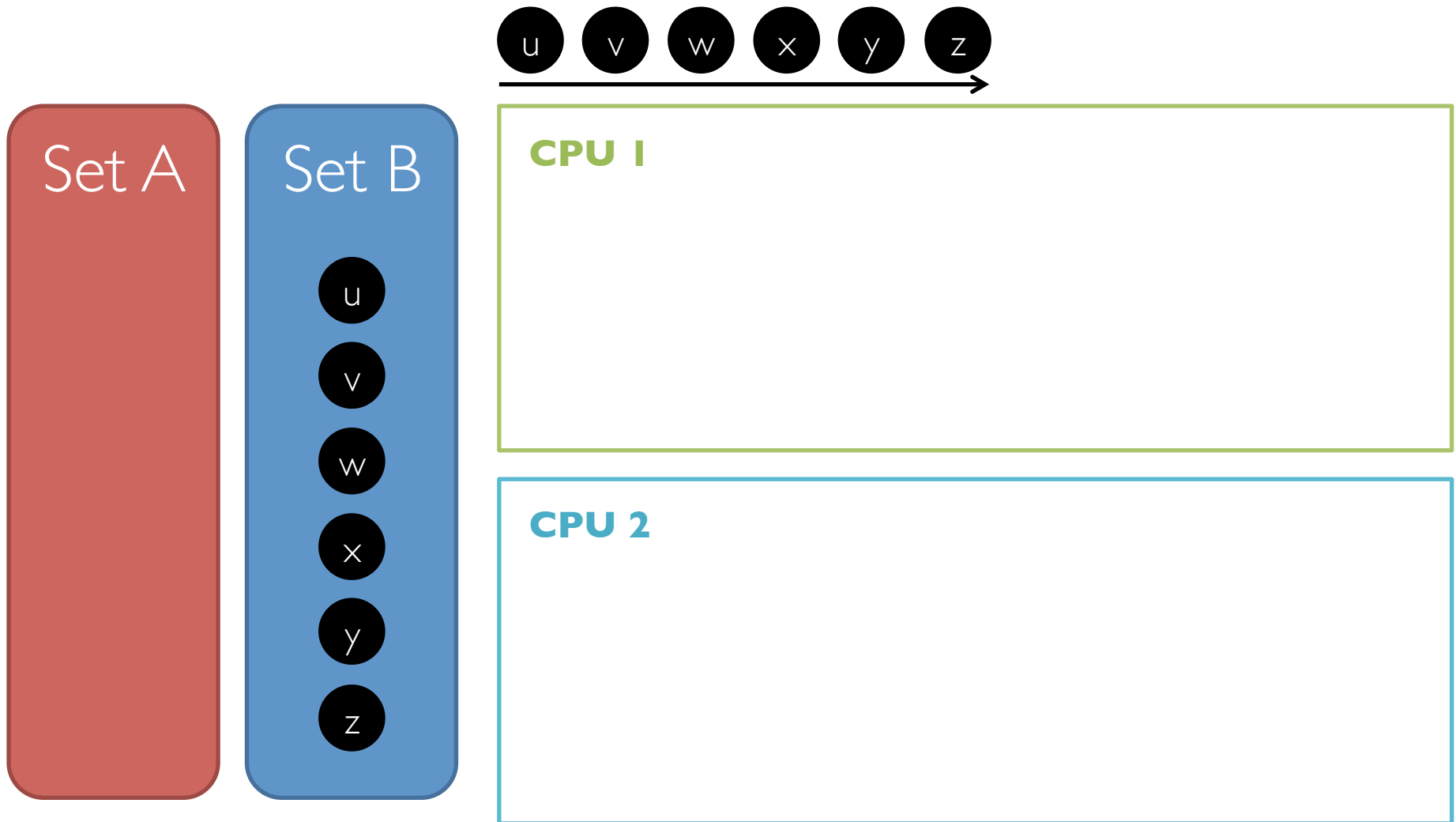


Double Greedy Algorithm



Return A

Parallel Double Greedy Algorithm



Parallel Double Greedy Algorithm

Set A



Set B



w

x

y

z

CPU 1



$$\Delta_+(u | ?) = ?$$

$$\Delta_-(u | ?) = ?$$

CPU 2

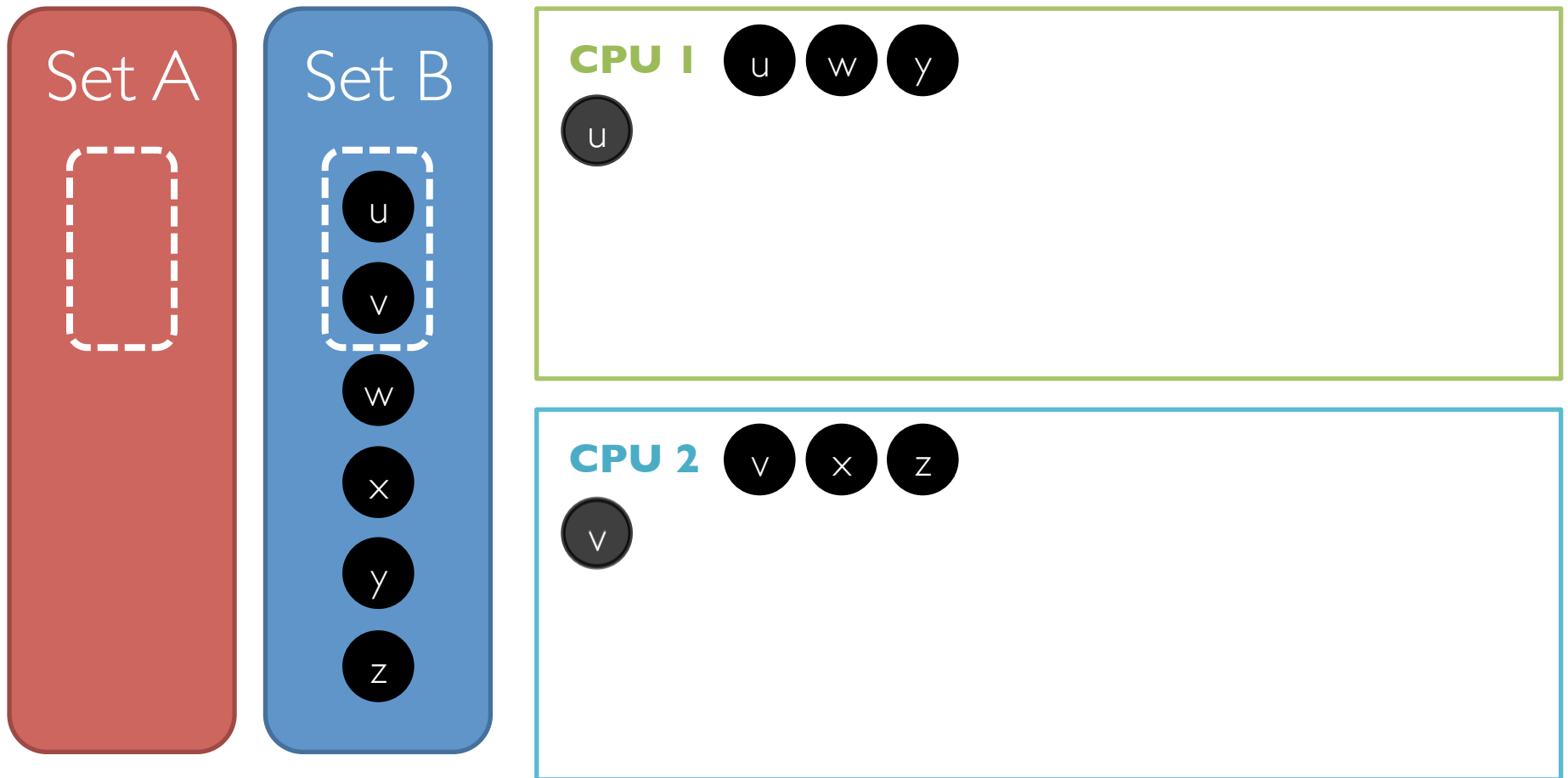


$$\Delta_+(v | ?) = ?$$

$$\Delta_-(v | ?) = ?$$

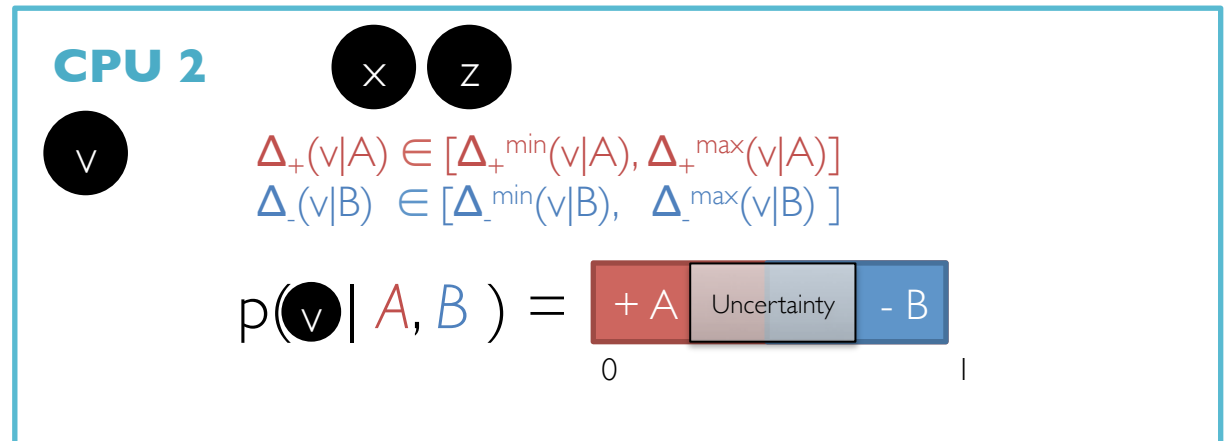
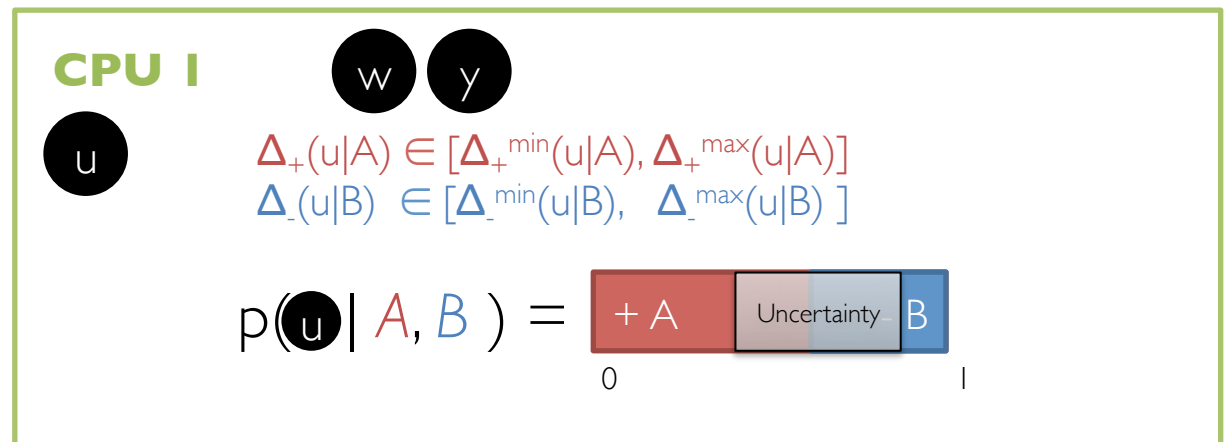
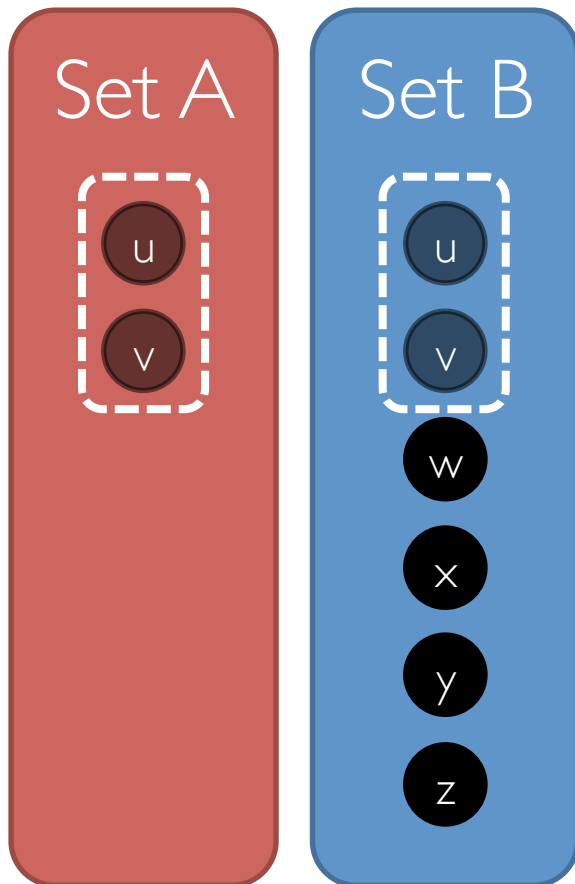
Concurrency Control Double Greedy

Maintain bounds on A, B → Enable threads to make decisions locally



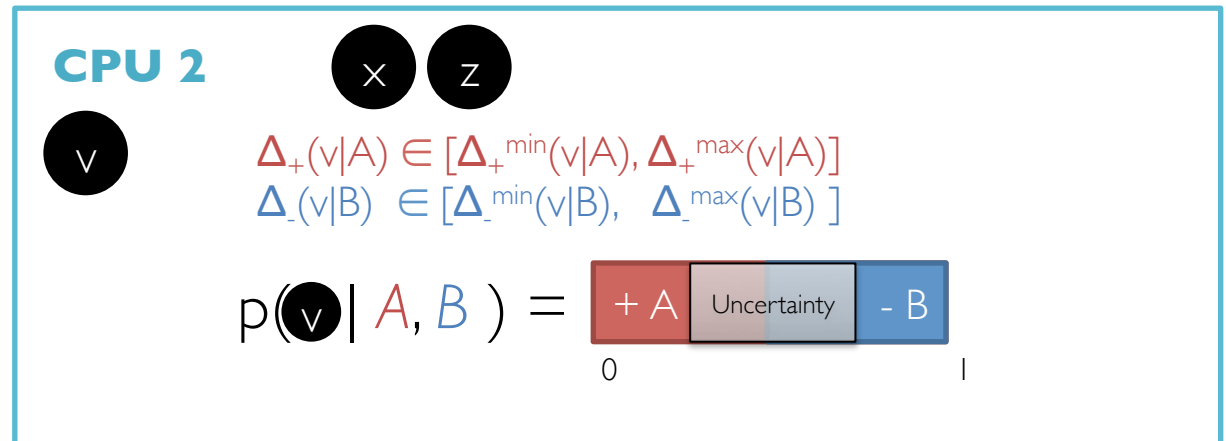
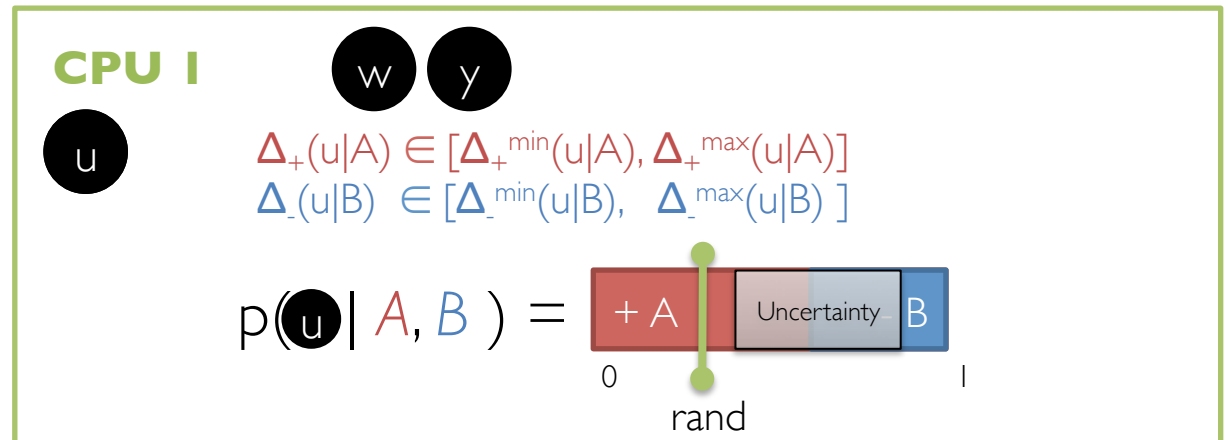
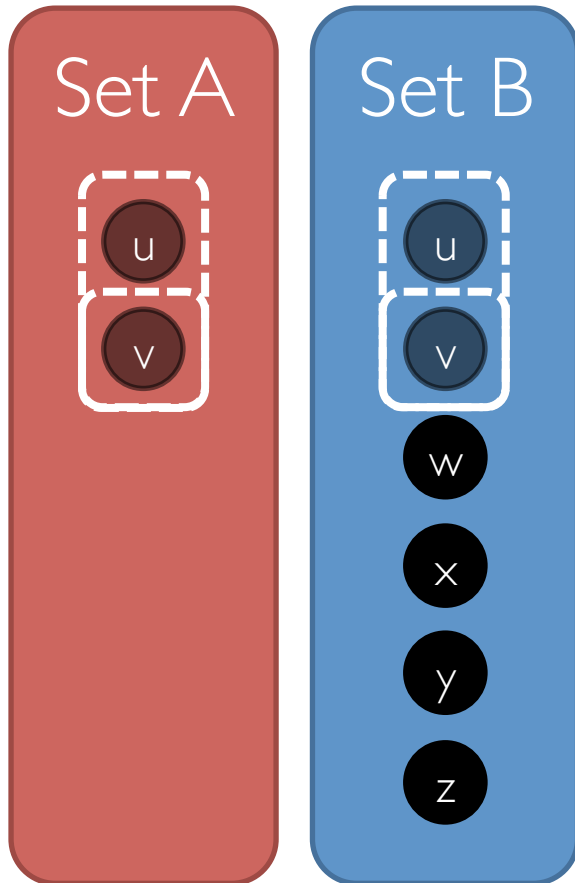
Concurrency Control Double Greedy

Maintain bounds on A, B \rightarrow Enable threads to make decisions locally



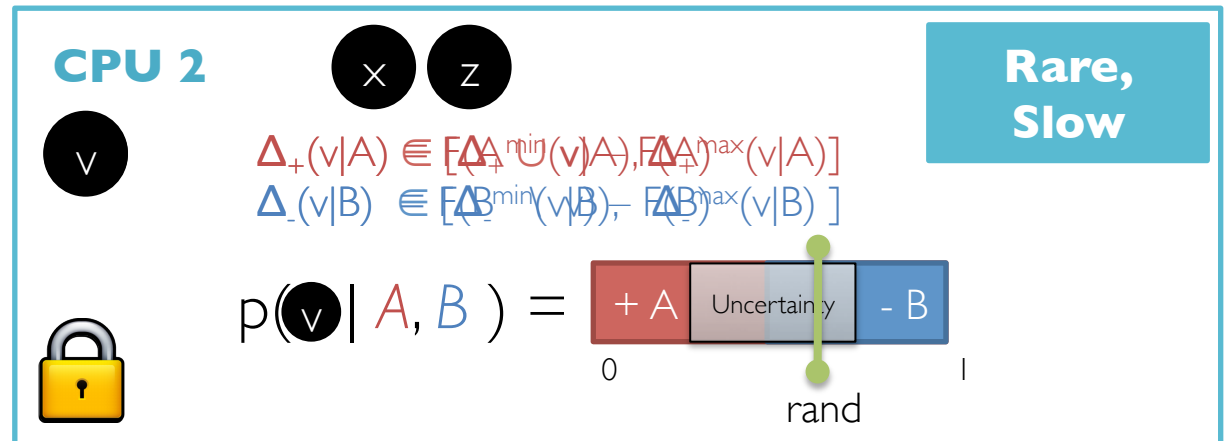
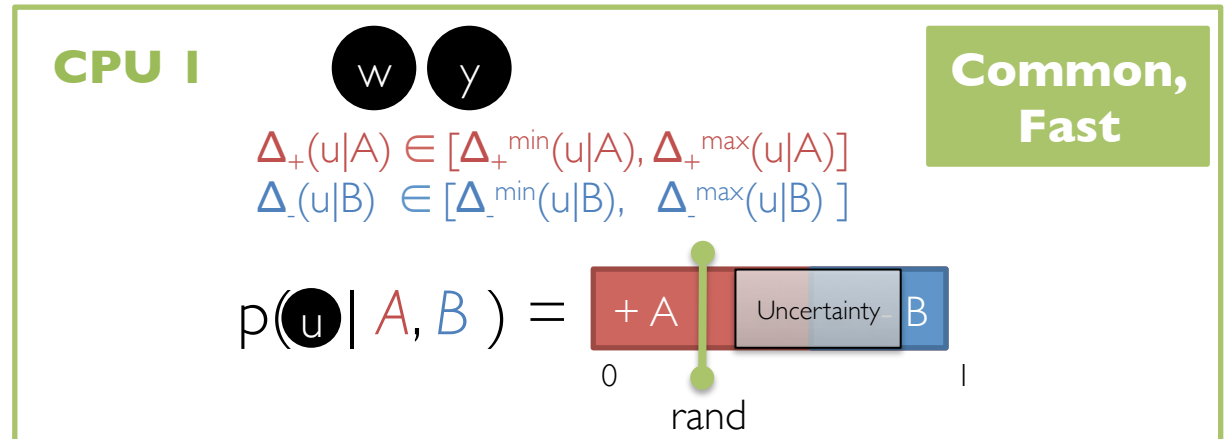
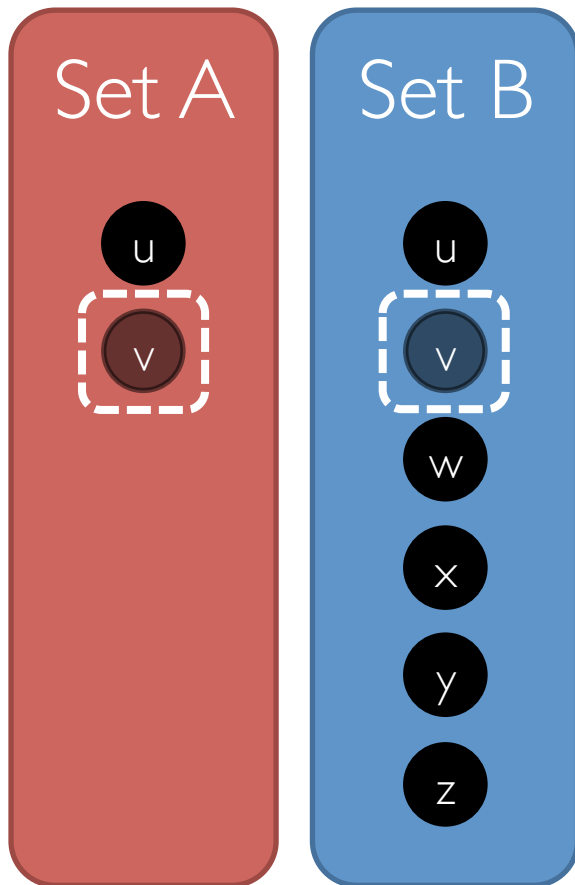
Concurrency Control Double Greedy

Maintain bounds on A, B \rightarrow Enable threads to make decisions locally



Concurrency Control Double Greedy

Maintain bounds on A, B → Enable threads to make decisions locally



Properties of CC Double Greedy

Correctness

Theorem: CC double greedy is serializable.

Corollary: CC double greedy preserves optimal approximation guarantee of $\frac{1}{2}\text{OPT}$.

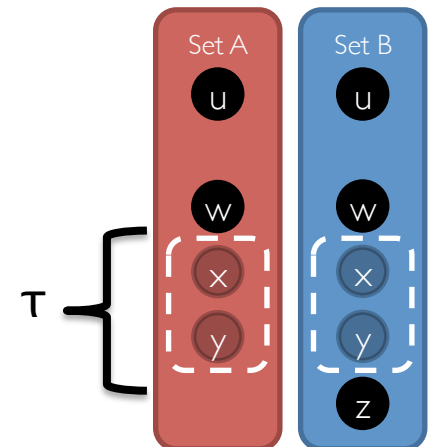
Concurrency

Lemma: CC has bounded overhead.

Expected number of blocked elements

set cover with costs: $< 2\tau$

sparse max cut: $< 2\tau |E| / |V|$



Change in Analysis

Coordination Free:

Provably fast and correct under key assumptions.

Concurrency Control:

Provably correct and fast under key assumptions.

Correctness

Easy Proof

Scalability

Challenging Proof

Empirical Validation

Multicore up to 16 threads

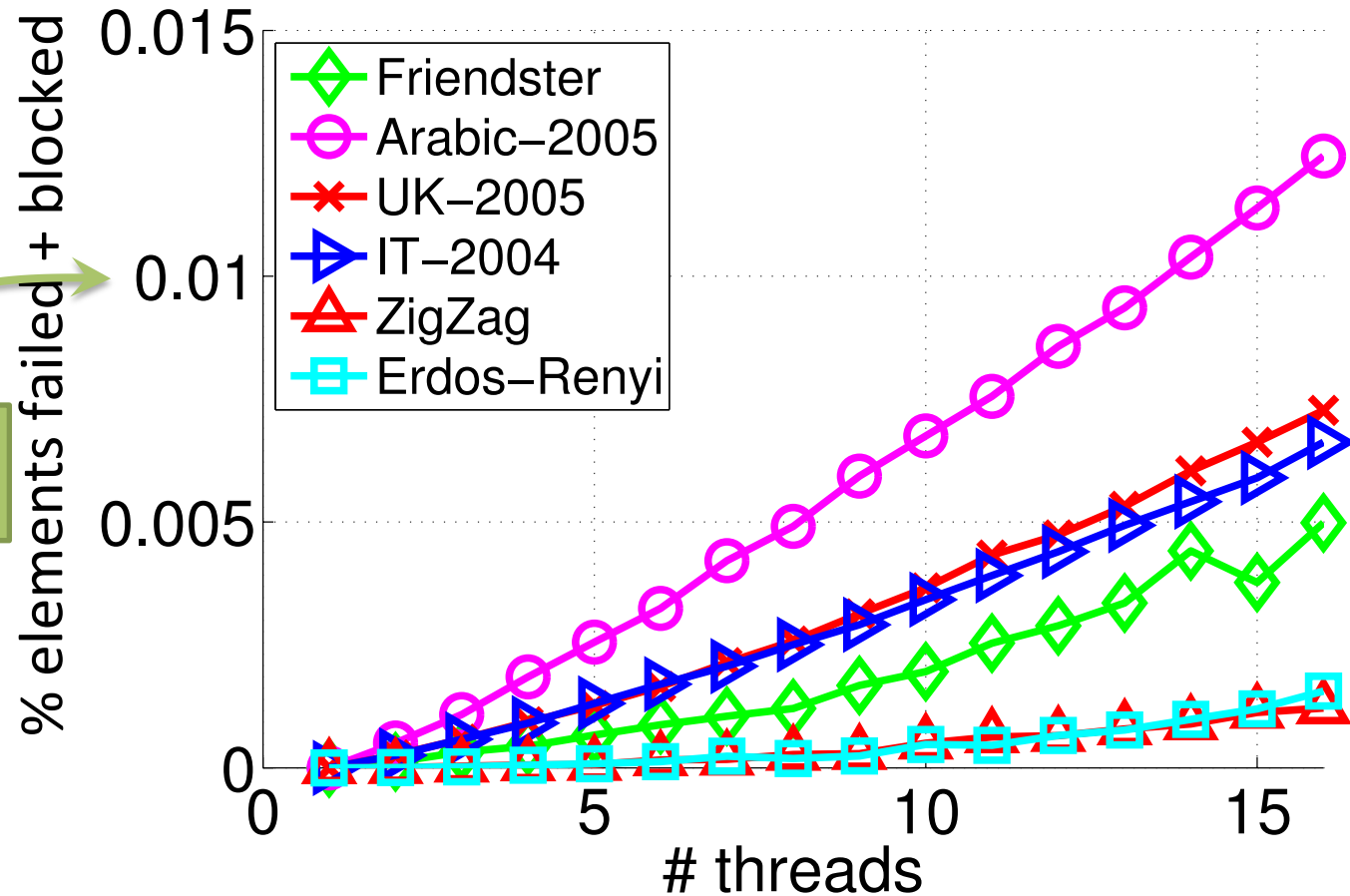
Set cover, Max graph cut

Real and synthetic graphs

Graph		Vertices	Edges
IT-2004	Italian web-graph	41 Million	1.1 Billion
UK-2005	UK web-graph	39 Million	0.9 Billion
Arabic-2005	Arabic web-graph	22 Million	0.6 Billion
Friendster	Social sub-network	10 Million	0.6 Billion
Erdos-Renyi	Synthetic random	20 Million	2.0 Billion
ZigZag	Synthetic expander	25 Million	2.0 Billion

CC Double Greedy Coordination

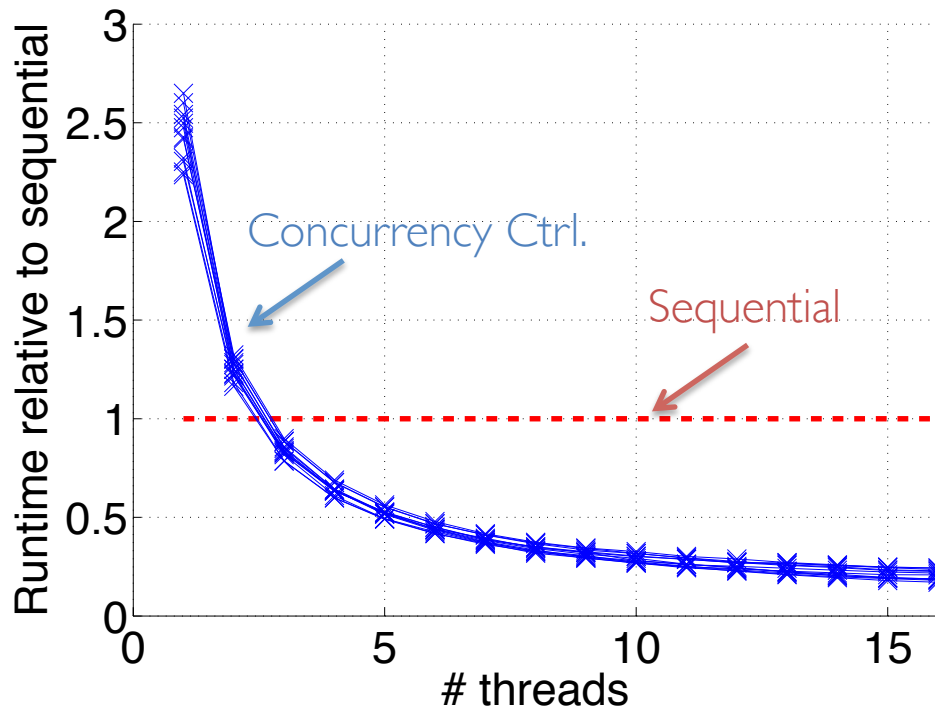
Increase in Coordination



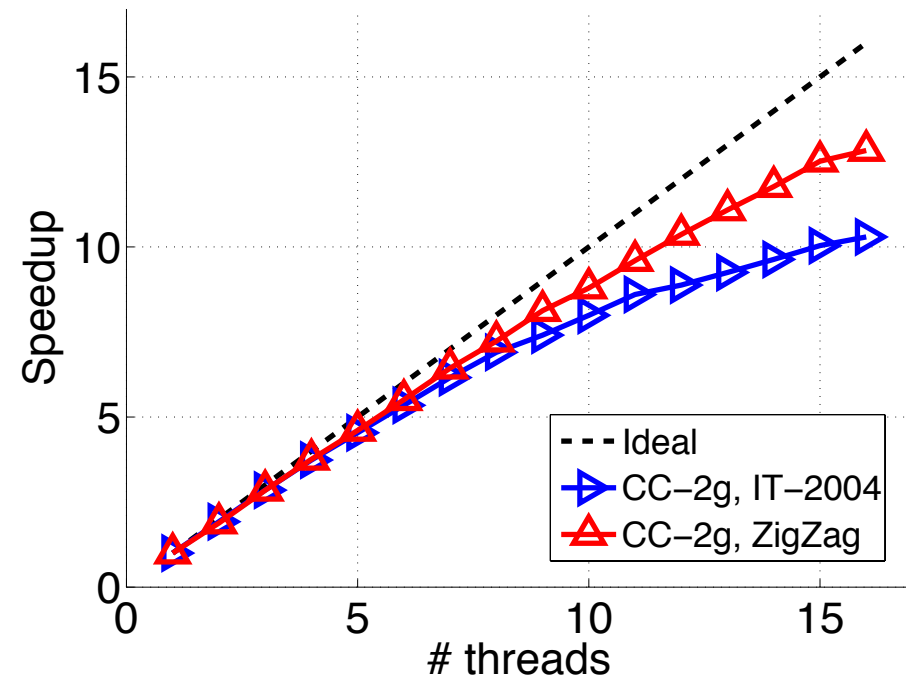
Only 0.01%
needs blocking!

Runtime and Strong-Scaling

Runtime, relative to sequential



Speedup for Max Graph Cut



Conclusion

	Scalability	Approximation
Sequential Double Greedy	Always slow	Always optimal
Concurrency Control Double Greedy	Usually fast	Always optimal
Coordination Free Double Greedy	Always fast	Near optimal

Paper @ NIPS 2014:

Parallel Double Greedy Submodular Maximization.

BACKUP SLIDES

Concurrency Control

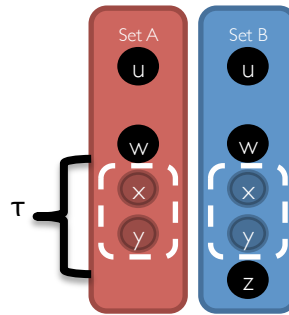
Correctness

Theorem: serializable.
preserves optimal
approximation bound
 $\frac{1}{2}\text{OPT}$.

Concurrency

Lemma:
bounded overhead.

set cover with costs: 2τ
sparse maxcut: $2|E|\tau / |V|$



from same dependencies
(uncertainty region)

Coordination Free

Lemma:

approximation bound
 $\frac{1}{2}\text{OPT} - \text{error}$

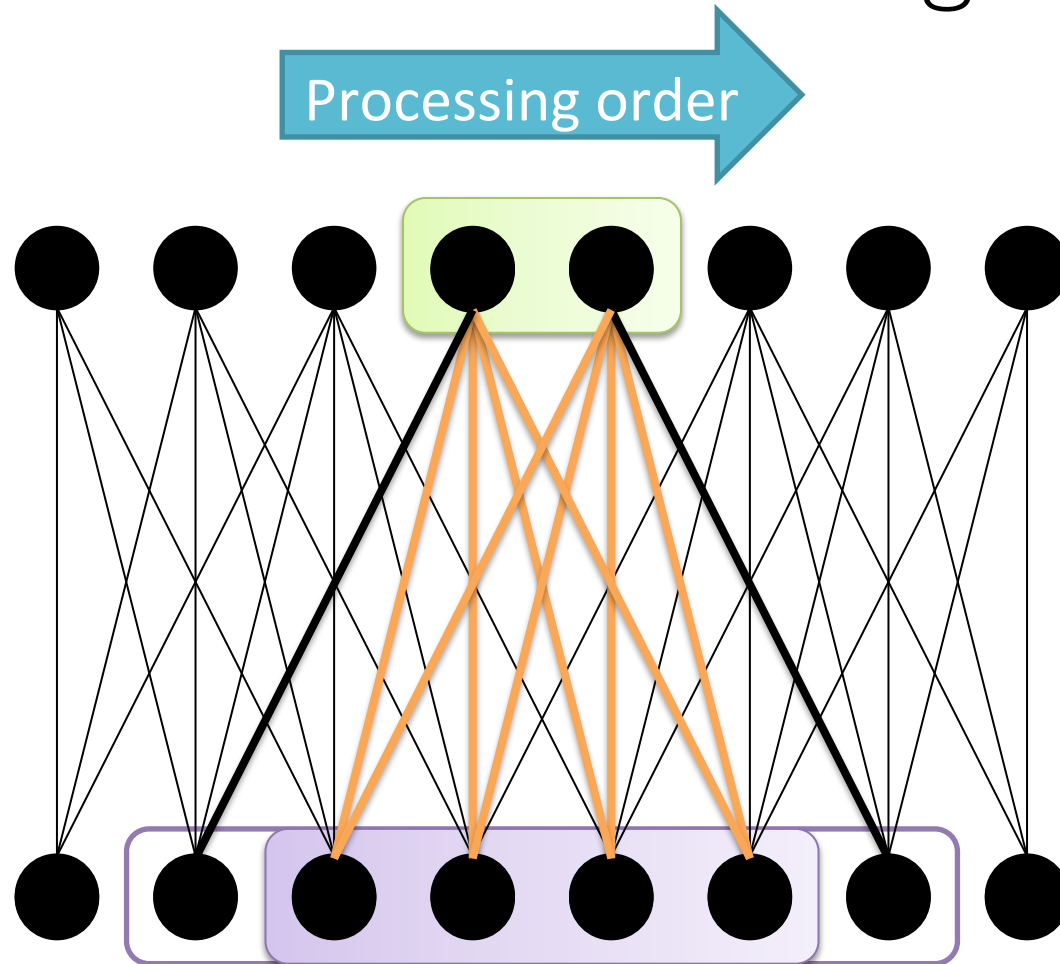
set cover with costs: $\geq \tau$
sparse maxcut: $|E|\tau / 2|V|$

no
overhead

Correctness

Concurrency

Adversarial Setting

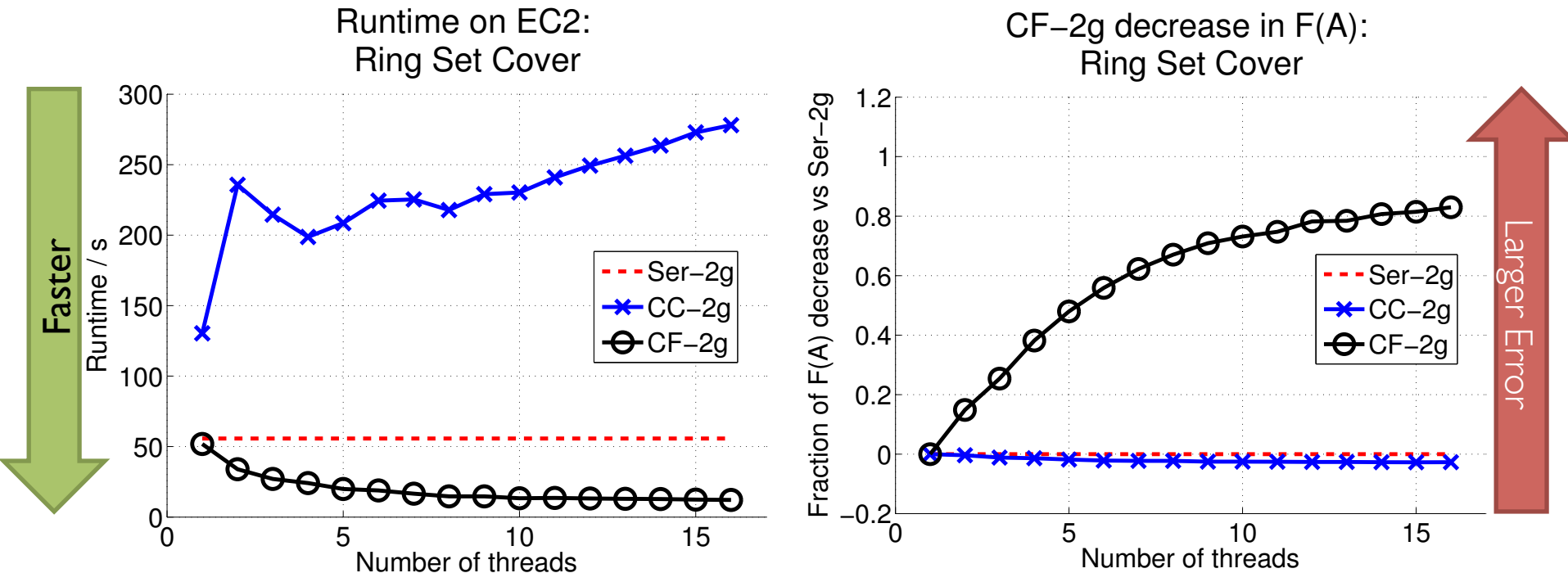


Overlapping covers



Increased coordination

Adversarial Setting – Ring Set Cover



Coord Free

Always fast

Possibly wrong

Conc Ctrl

Possibly slow

Always optimal