

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

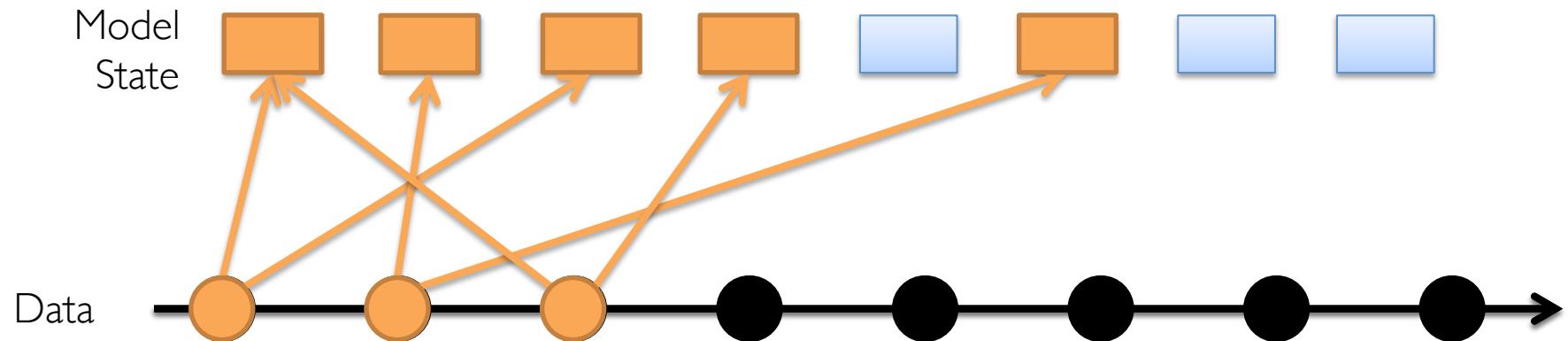


-amplab 

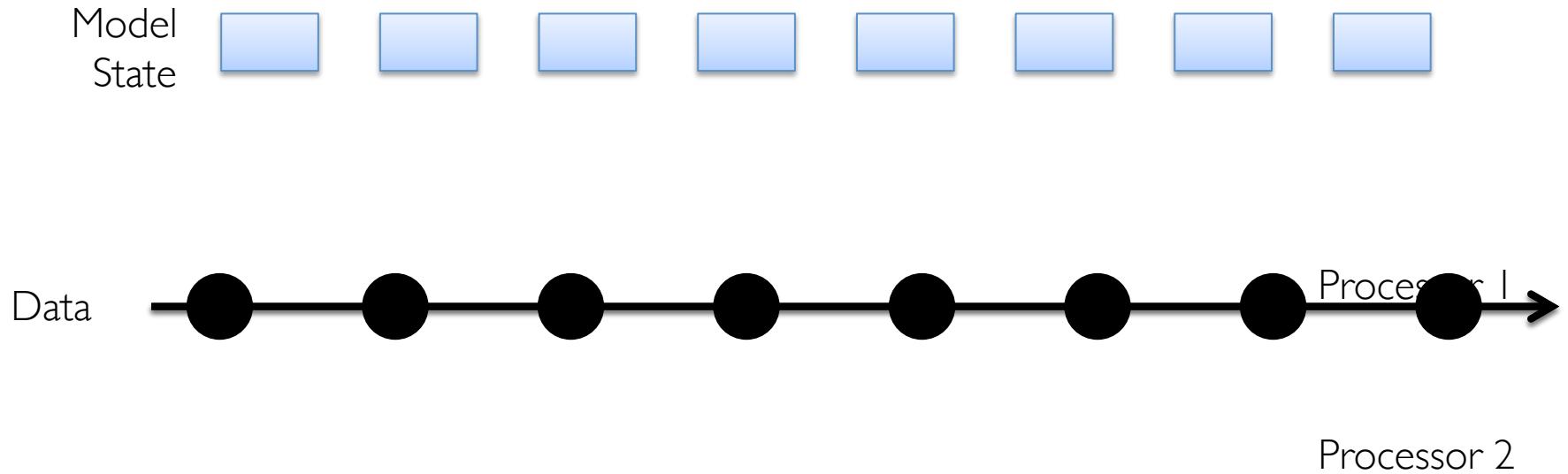
The logo consists of the word "amplab" in a lowercase sans-serif font, with the "a" and "m" in orange and the rest in dark blue. A small orange wavy line graphic follows the "l".

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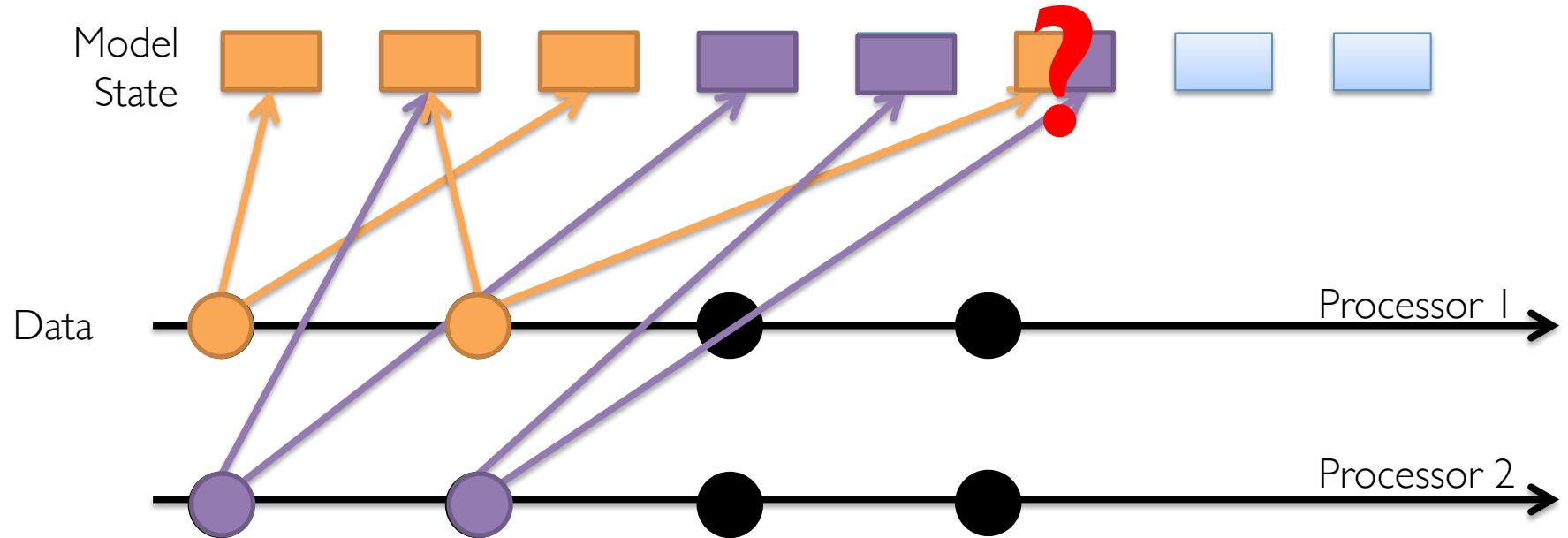
Serial Inference



Parallel Inference



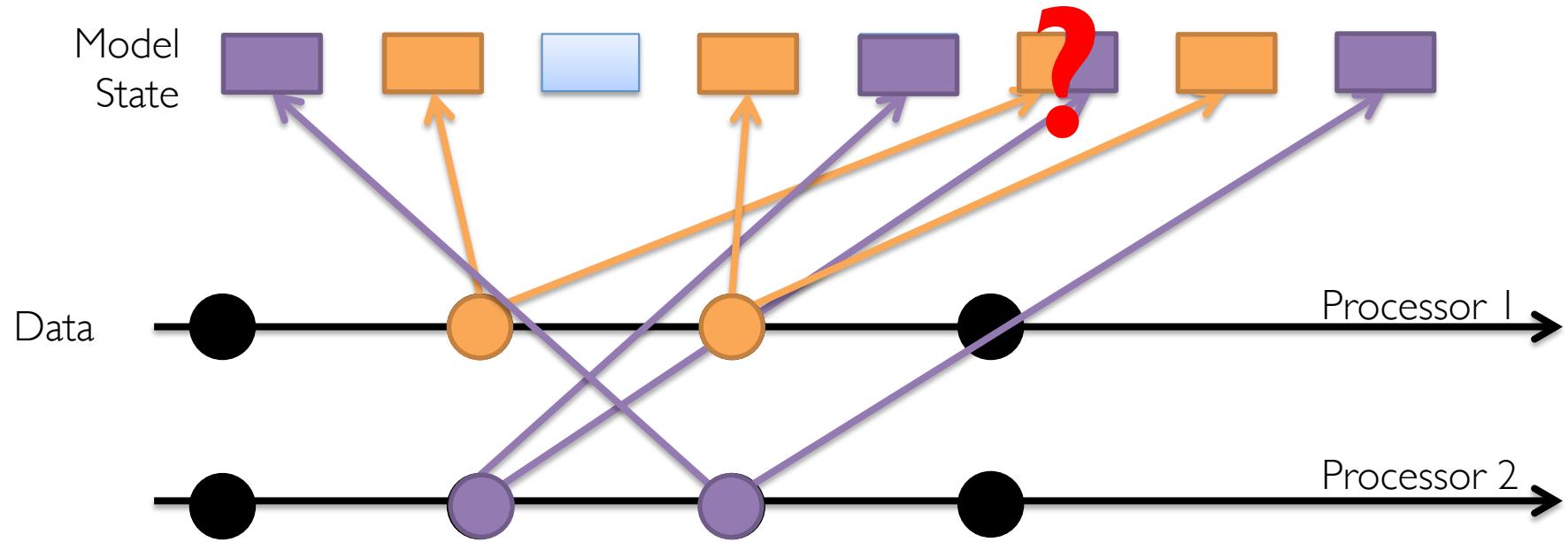
Parallel Inference



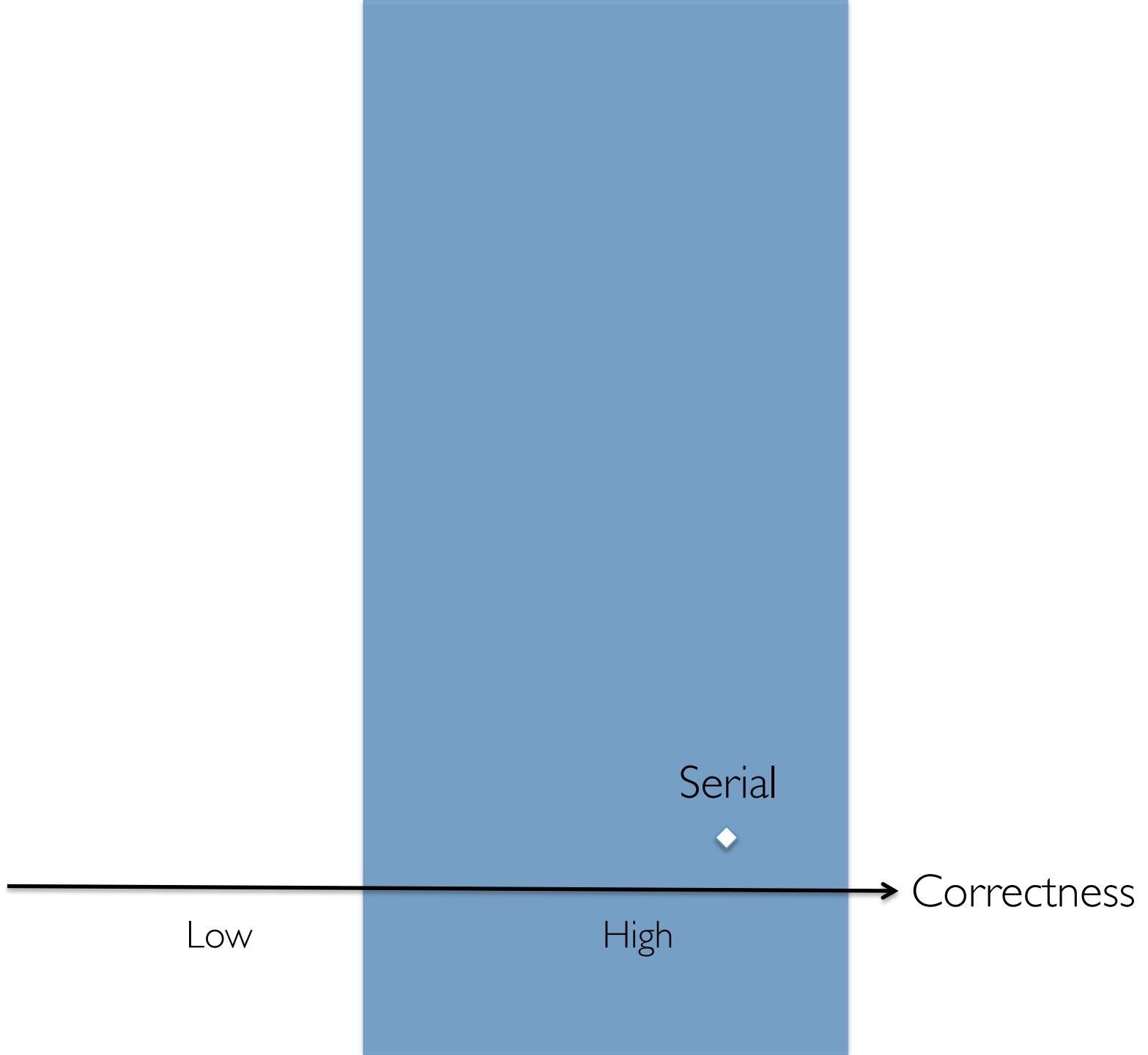
Correctness:
serial equivalence

Concurrency:
more machines = less time

Coordination Free Parallel Inference



Correctness and Consistency:
Depends on Assumptions
(almost) free



Concurrency

High

Coordination-free

Low

Serial

Low

High

Correctness

Concurrency

High

Coordination-free

Low

Low

Concurrency
Control

Database mechanisms

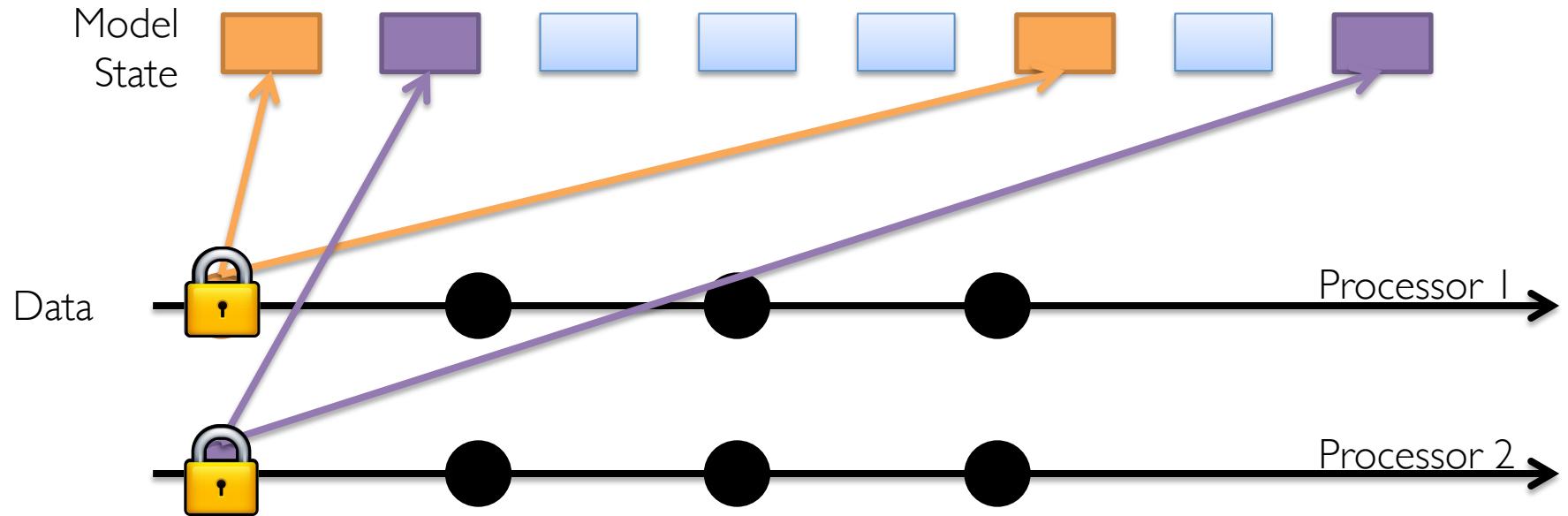
- Guarantee correctness
- Maximize concurrency
- Mutual exclusion
- Optimistic CC

Serial

High

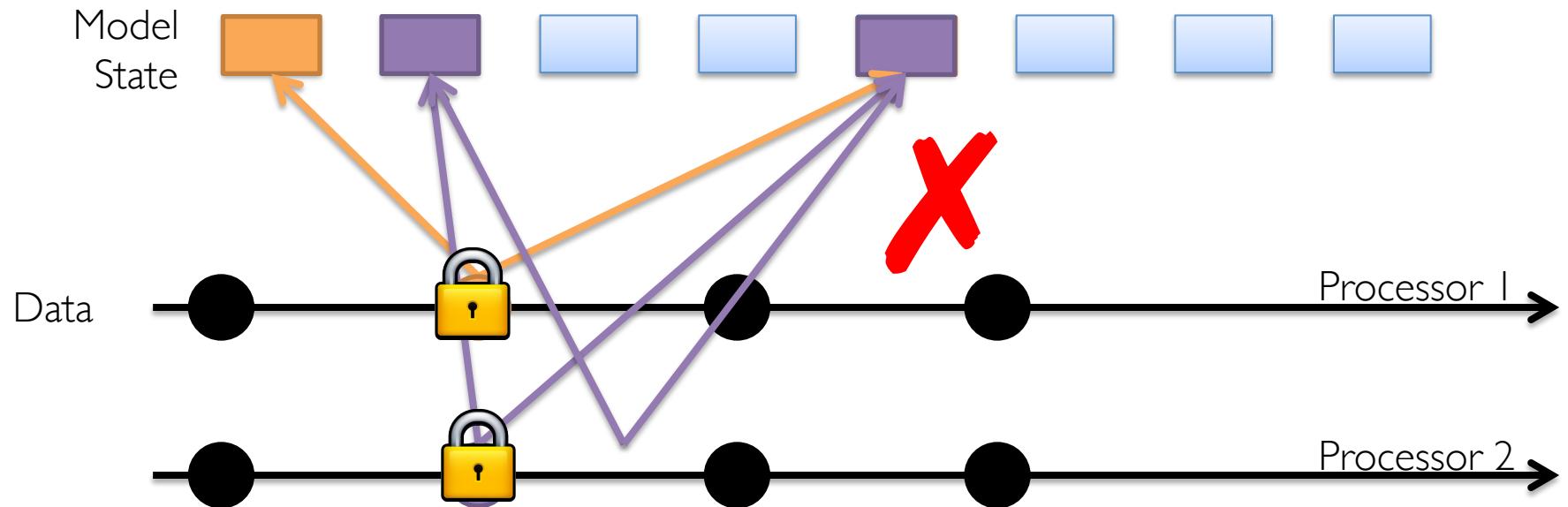
Correctness

Mutual Exclusion Through Locking



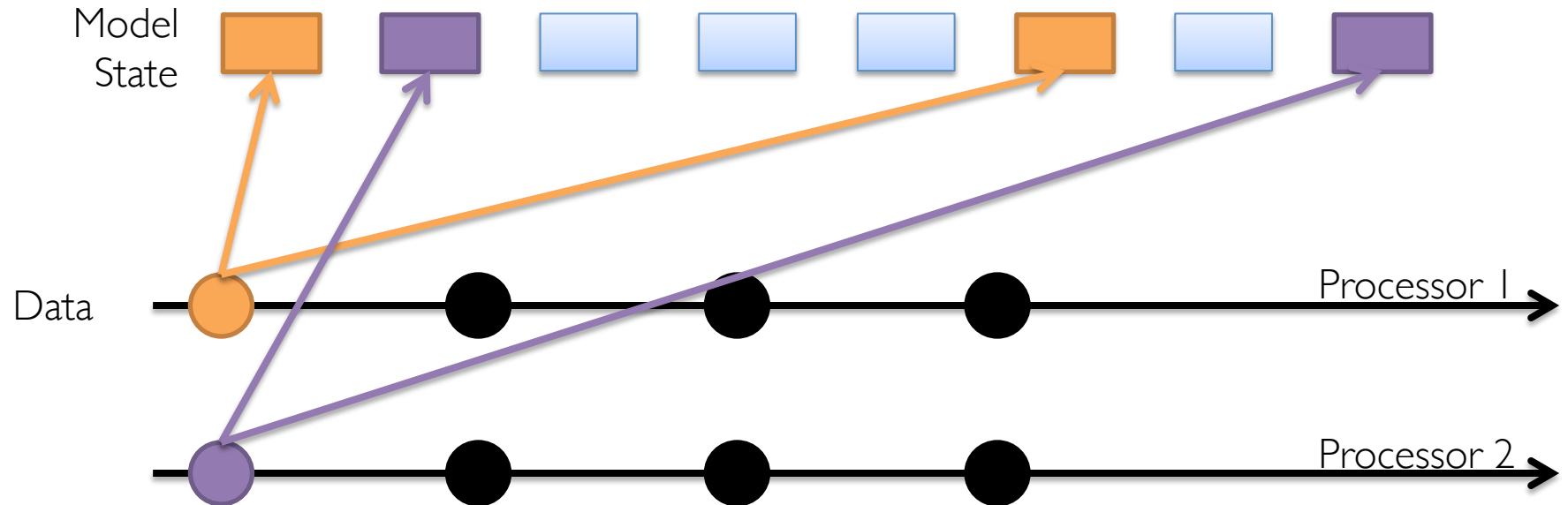
Introducing locking (scheduling) protocols to prevent potential conflicts.

Mutual Exclusion Through Locking



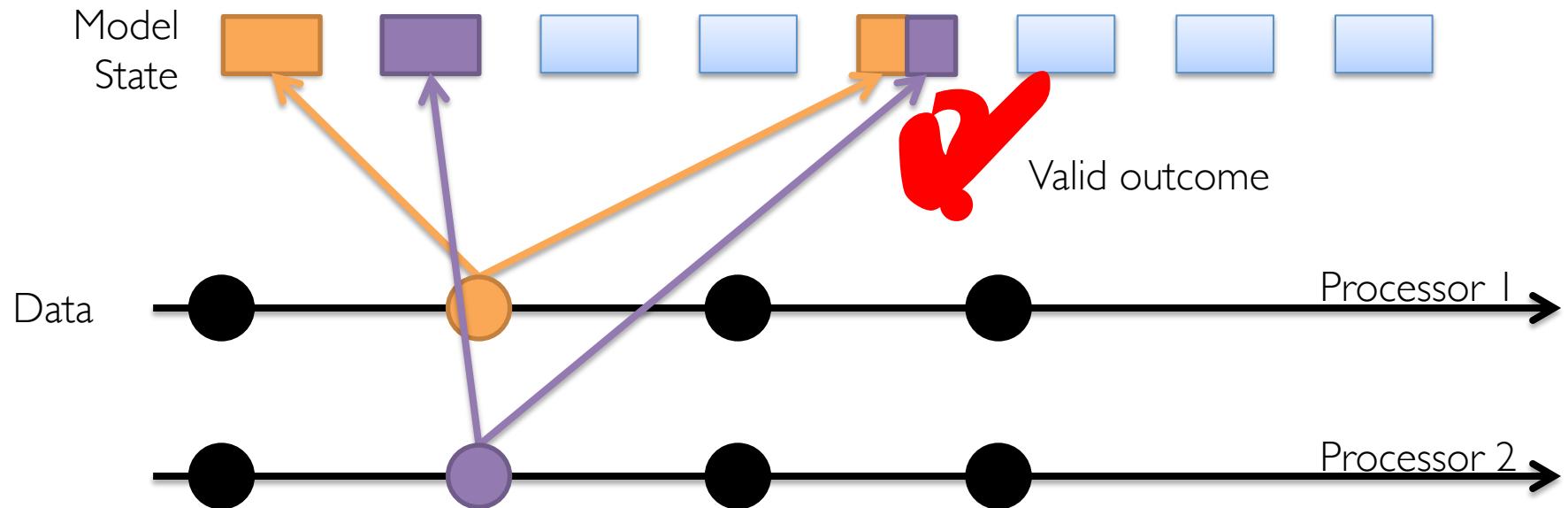
Enforce serialization of computation that could conflict.

Optimistic Concurrency Control



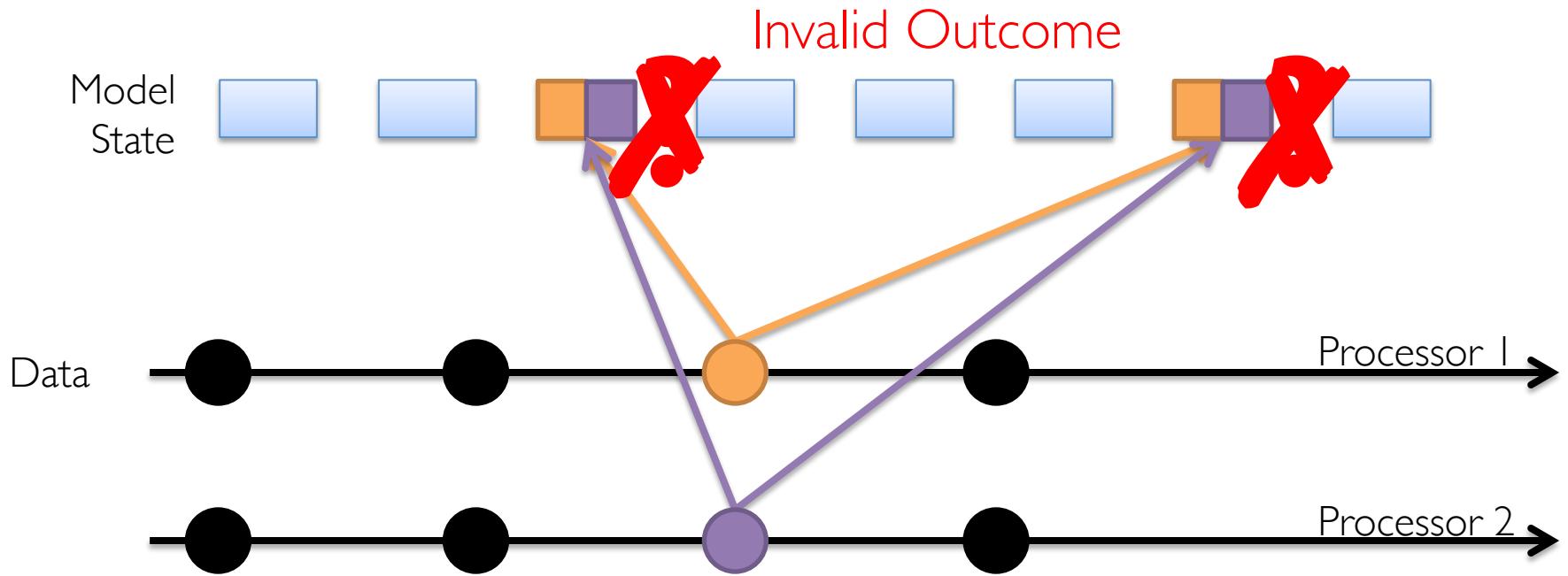
Allow computation to proceed without blocking.

Optimistic Concurrency Control



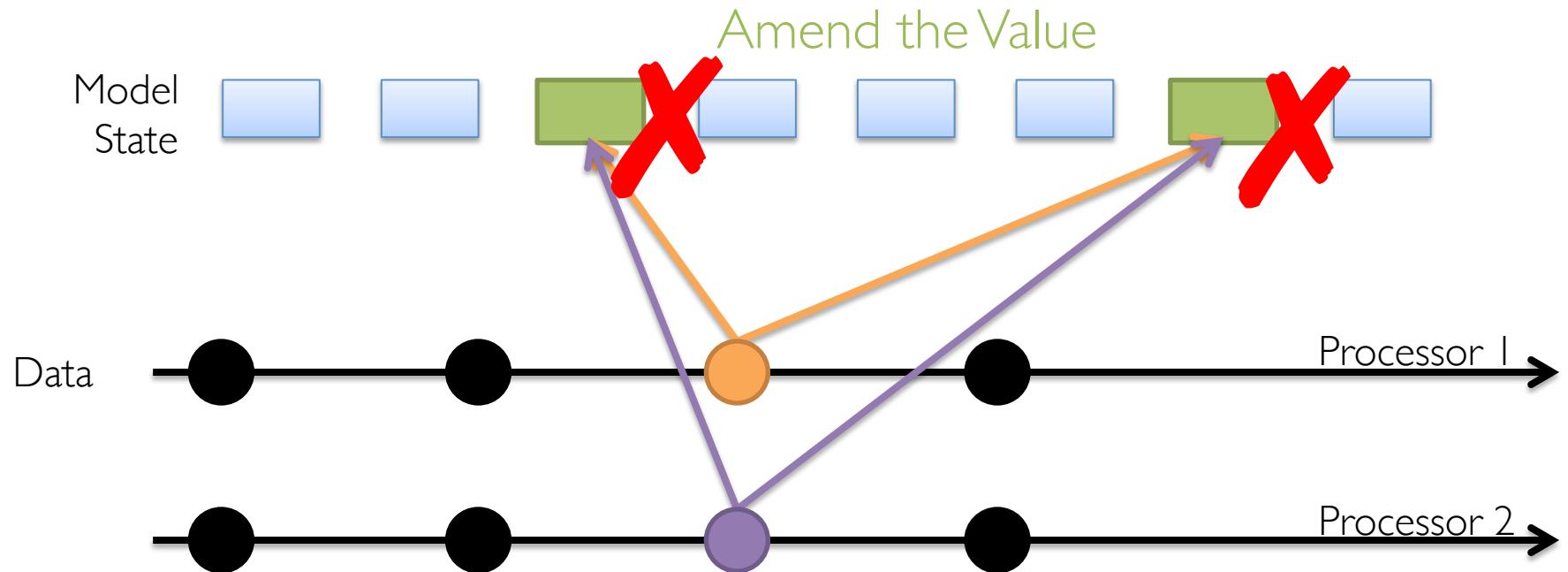
Validate potential conflicts.

Optimistic Concurrency Control



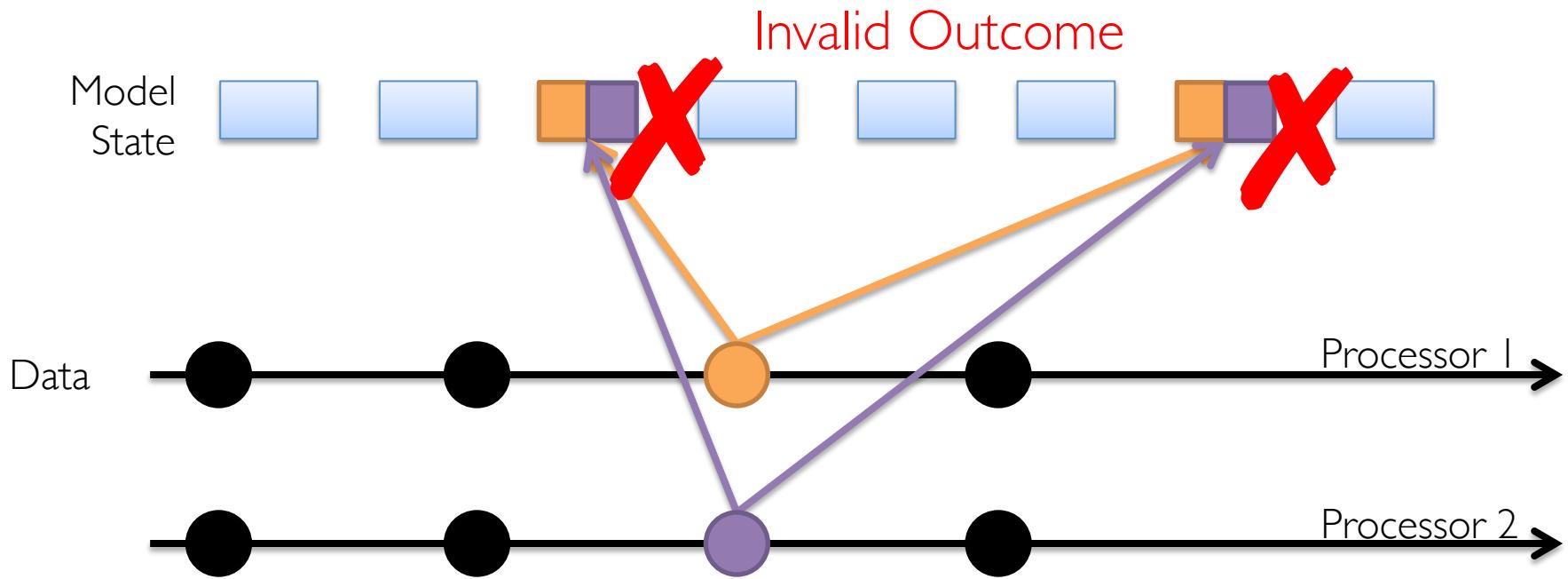
Validate potential conflicts.

Optimistic Concurrency Control



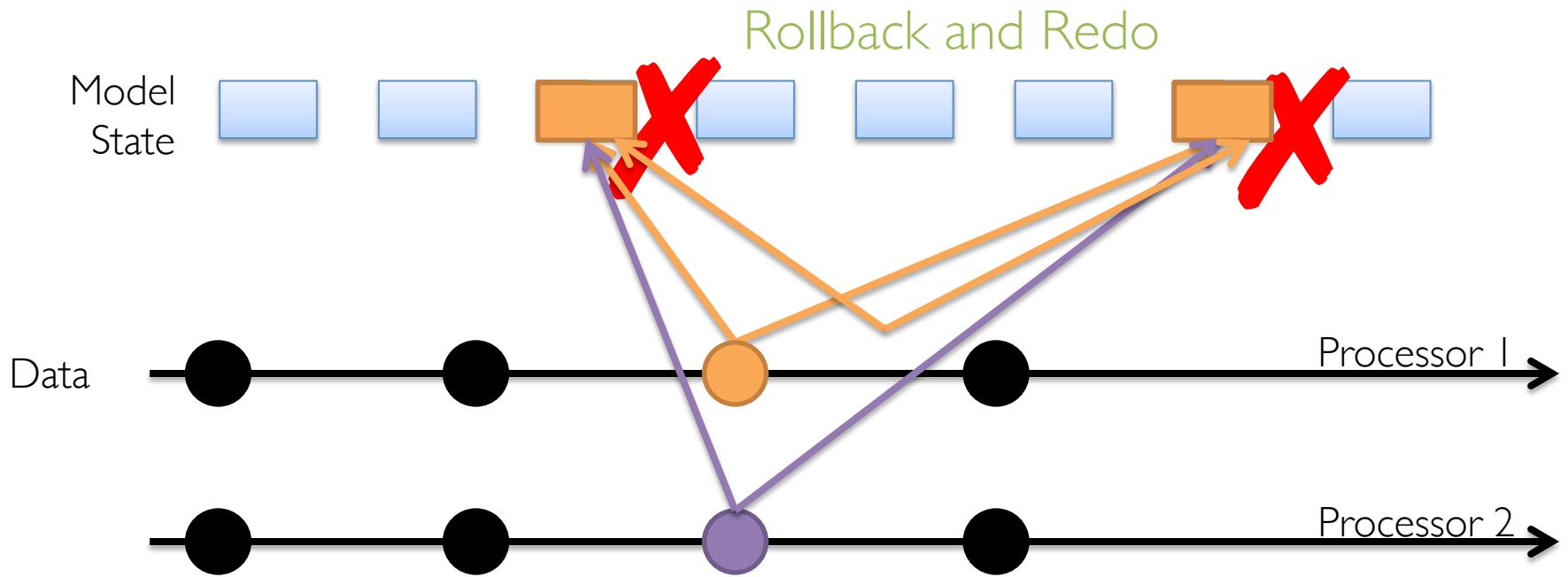
Take a compensating action.

Optimistic Concurrency Control



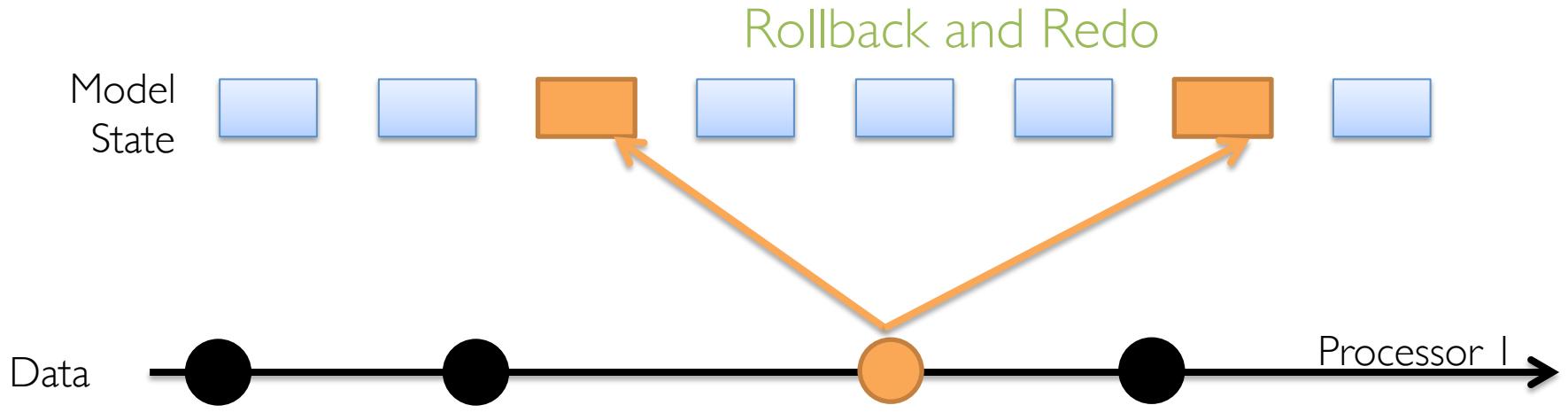
Validate potential conflicts.

Optimistic Concurrency Control



Take a compensating action.

Optimistic Concurrency Control



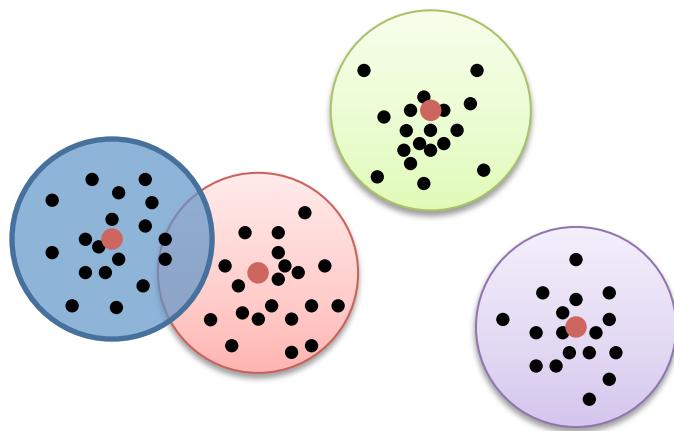
Non-Blocking Computation → Concurrency

Validation: Identify Errors
Resolution: Correct Errors

→ Correctness

Optimistic Concurrency Control for Machine Learning

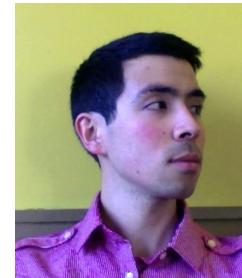
Non-parametric Clustering
Distributed DP-Means
[NIPS'13]



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



$F(S)$ = area covered by S

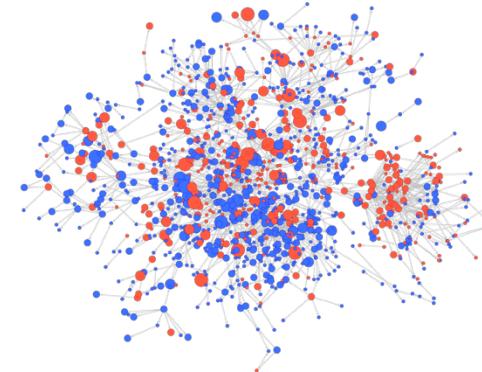
$F(S) = I(Y; X_S)$
= reduction in uncertainty
(Krause & Guestrin 2005)

Document Summarization



(Lin & Bilmes 2011)

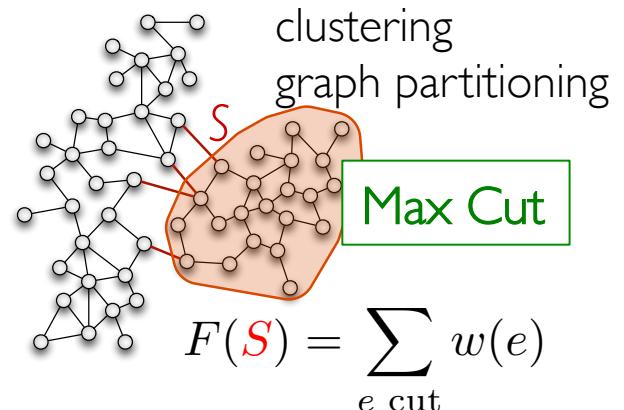
Network Analysis



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

(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

Greedy (Nemhauser et al, 1978)
 $(1 - 1/e)$ - approximation
Optimal polytime

Double Greedy (Buchbinder et al, 2012)
 $\frac{1}{2}$ - approximation
Optimal polytime

Parallel /
Distributed

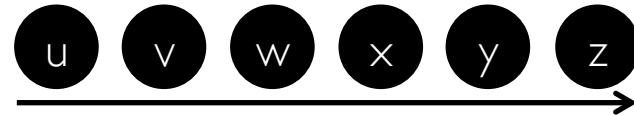
GREEDi (Mirzasoleiman et al, 2013)
 $(1 - 1/e)^2 / p$ – approximation
1 MapReduce round

(Kumar et al, 2013)
1 / $(2 + \epsilon)$ – approximation
 $O(1/\epsilon)$ MapReduce rounds

Concurrency Control Double Greedy
Optimal $\frac{1}{2}$ - approximation
Bounded overhead

Coordination Free Double Greedy
Bounded error
Minimal overhead

Double Greedy Algorithm

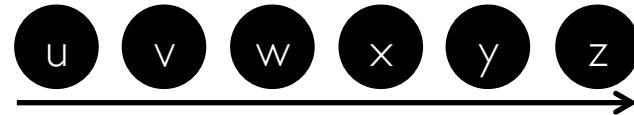


Set A

Set B



Double Greedy Algorithm

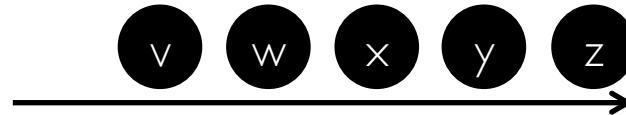
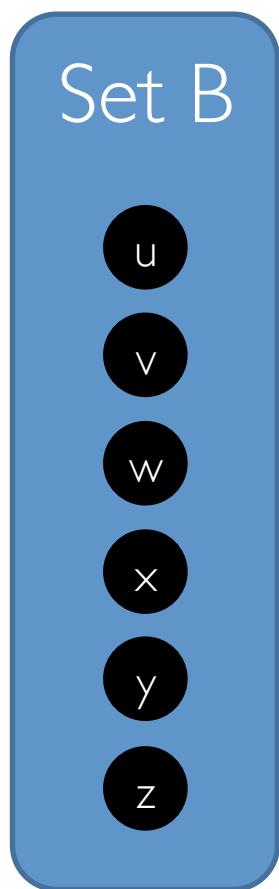


Set A

Set B

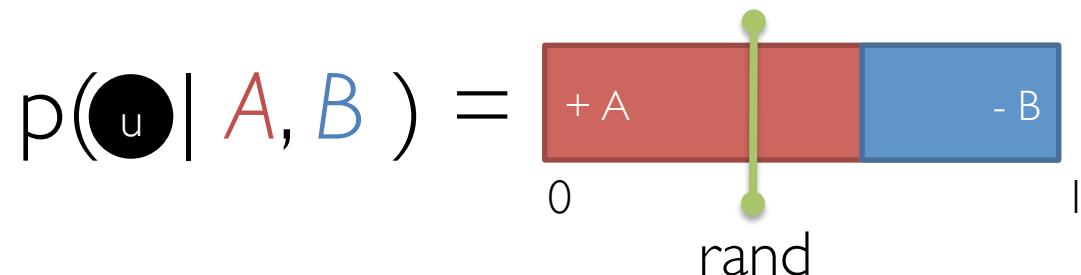


Double Greedy Algorithm

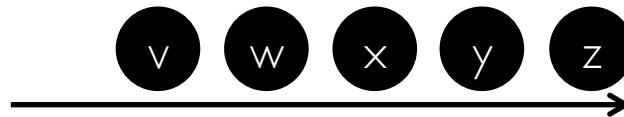
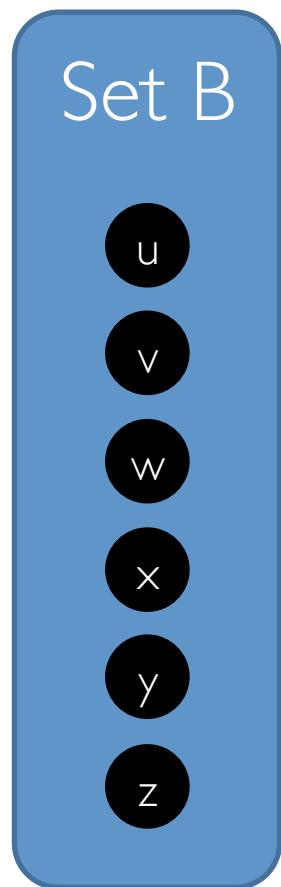


Marginal gains

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

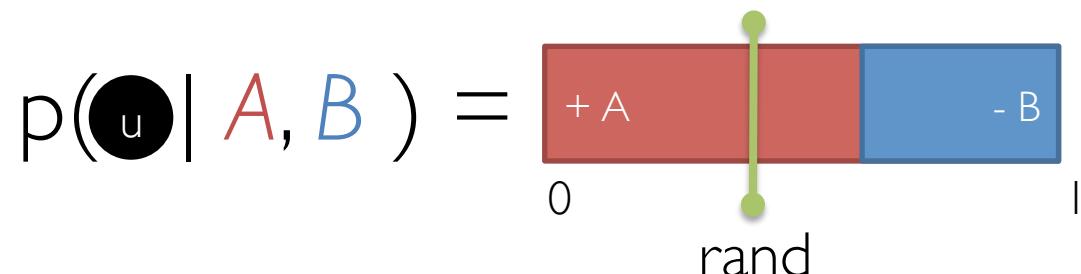


Double Greedy Algorithm

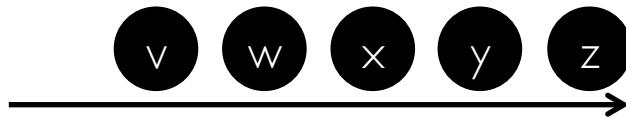
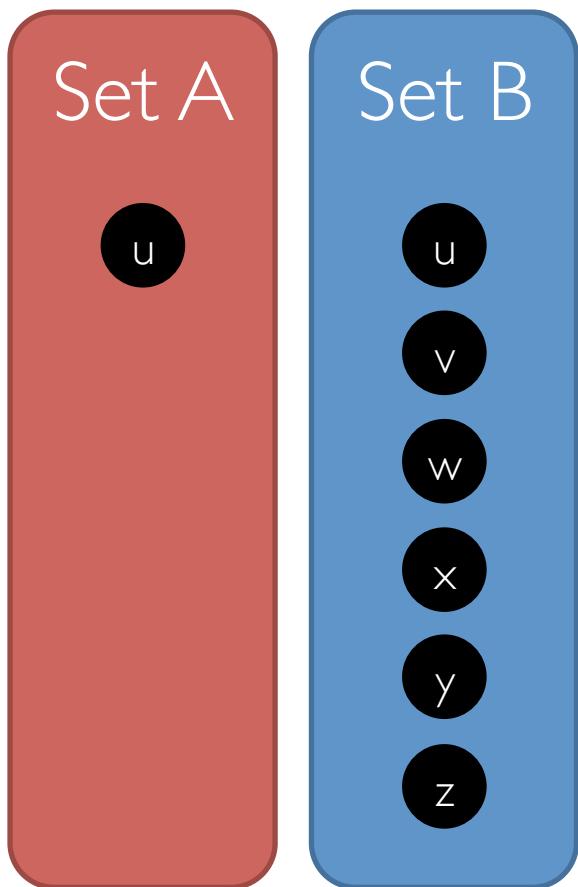


Marginal gains

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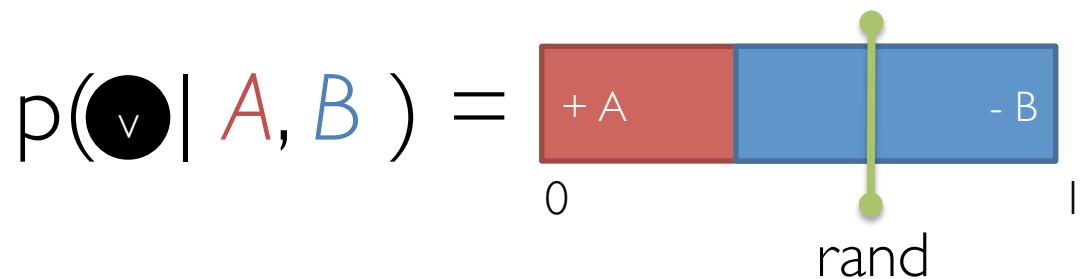


Double Greedy Algorithm

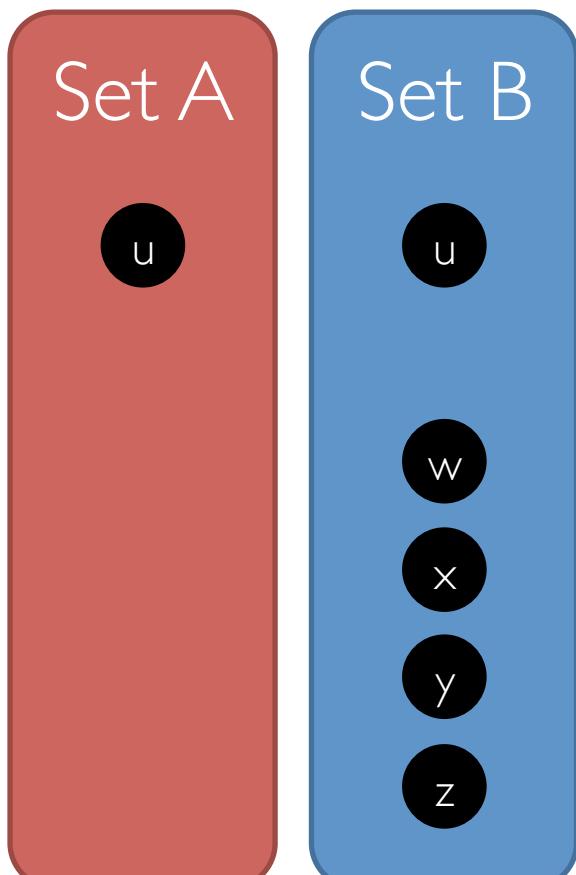


Marginal gains

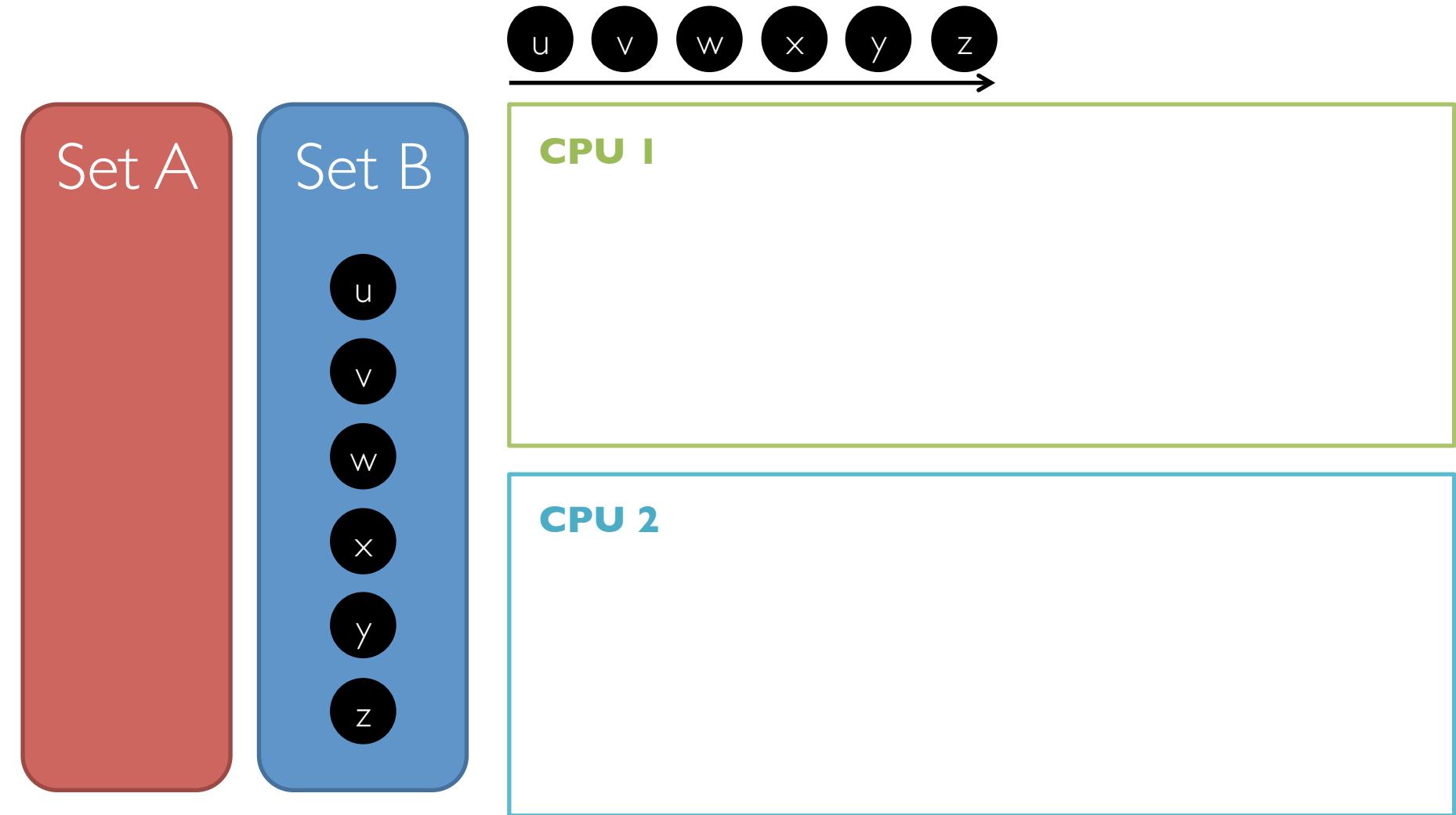
$$\Delta_+(v|A) = F(A \cup v) - F(A),$$
$$\Delta_-(v|B) = F(B \setminus v) - F(B).$$



Double Greedy Algorithm



Parallel Double Greedy Algorithm



Parallel Double Greedy Algorithm

Set A



Set B



CPU 1

u w y

$$\Delta_+(u \mid ?) = ?$$
$$\Delta_-(u \mid ?) = ?$$

CPU 2

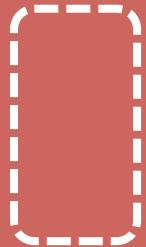
v x z

$$\Delta_+(v \mid ?) = ?$$
$$\Delta_-(v \mid ?) = ?$$

Concurrency Control Double Greedy

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

Set A



Set B



CPU 1

u

u w y

CPU 2

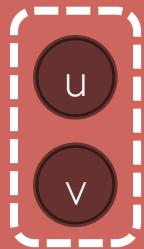
v

v x z

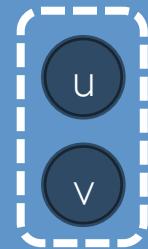
Concurrency Control Double Greedy

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

Set A



Set B



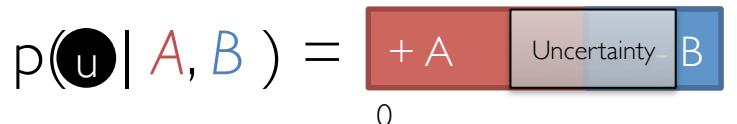
CPU 1

u

w

y

$$\Delta_+(u|A) \in [\Delta_+^{\min}(u|A), \Delta_+^{\max}(u|A)]$$
$$\Delta_-(u|B) \in [\Delta_-^{\min}(u|B), \Delta_-^{\max}(u|B)]$$



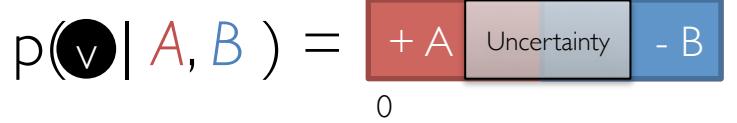
CPU 2

v

x

z

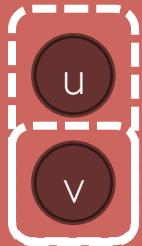
$$\Delta_+(v|A) \in [\Delta_+^{\min}(v|A), \Delta_+^{\max}(v|A)]$$
$$\Delta_-(v|B) \in [\Delta_-^{\min}(v|B), \Delta_-^{\max}(v|B)]$$



Concurrency Control Double Greedy

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

Set A



Set B



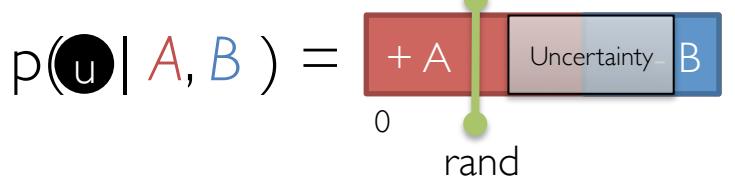
CPU 1

u

w

y

$$\Delta_+(u|A) \in [\Delta_+^{\min}(u|A), \Delta_+^{\max}(u|A)]$$
$$\Delta_-(u|B) \in [\Delta_-^{\min}(u|B), \Delta_-^{\max}(u|B)]$$



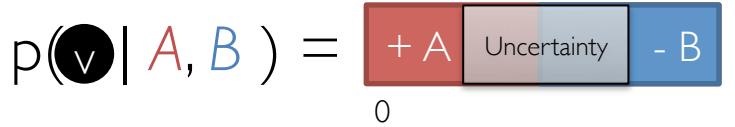
CPU 2

v

x

z

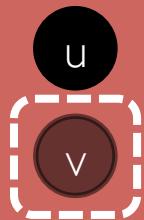
$$\Delta_+(v|A) \in [\Delta_+^{\min}(v|A), \Delta_+^{\max}(v|A)]$$
$$\Delta_-(v|B) \in [\Delta_-^{\min}(v|B), \Delta_-^{\max}(v|B)]$$



Concurrency Control Double Greedy

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

Set A



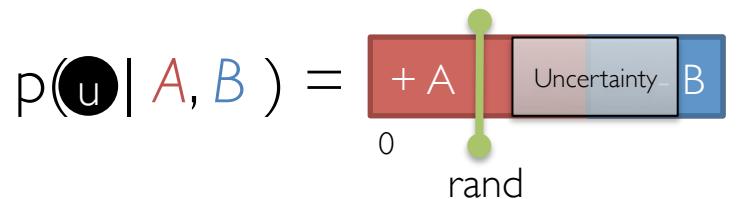
Set B



CPU 1



$$\Delta_+(u|A) \in [\Delta_+^{\min}(u|A), \Delta_+^{\max}(u|A)]$$
$$\Delta_-(u|B) \in [\Delta_-^{\min}(u|B), \Delta_-^{\max}(u|B)]$$

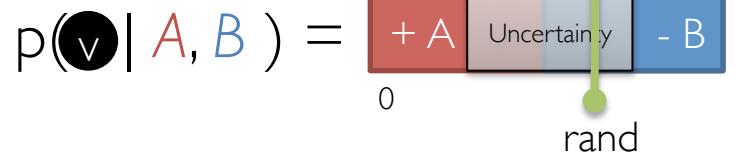


Common,
Fast

CPU 2



$$\Delta_+(v|A) \in [\Delta_+^{\min}(v|A), \Delta_+^{\max}(v|A)]$$
$$\Delta_-(v|B) \in [\Delta_-^{\min}(v|B), \Delta_-^{\max}(v|B)]$$



Rare,
Slow

Properties of CC Double Greedy

Correctness

Theorem: CC double greedy is serializable.

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

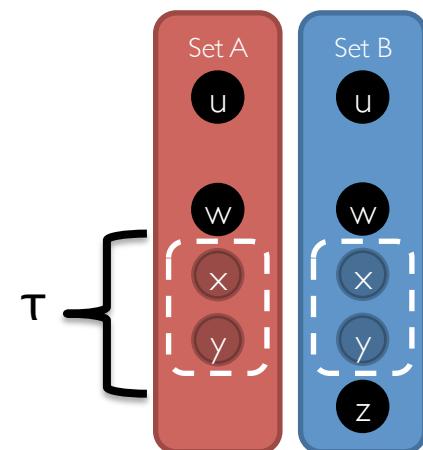
Concurrency

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

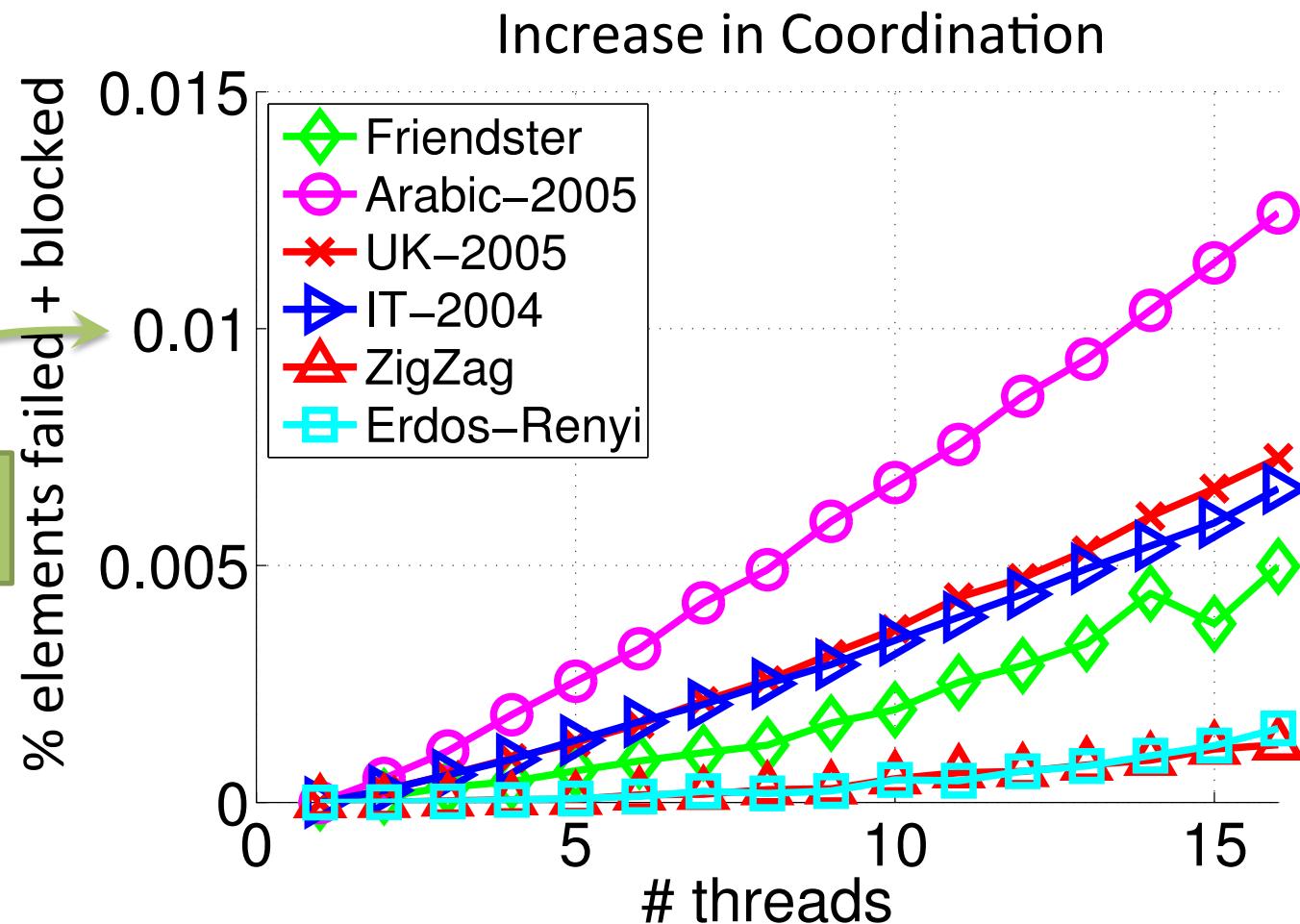
Multicore up to 16 threads

Set cover, Max graph cut

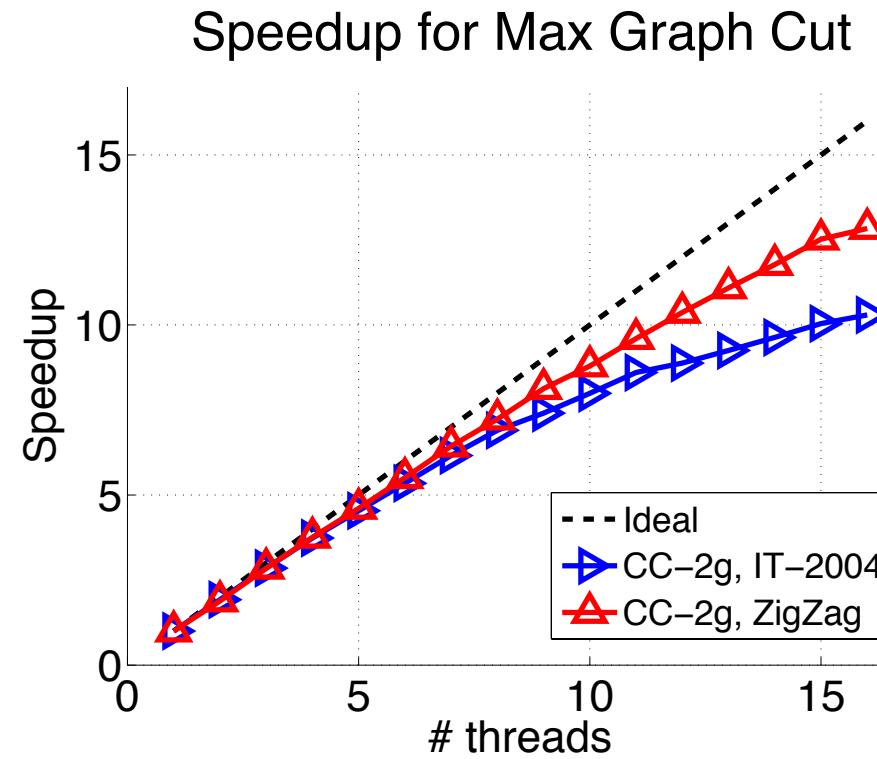
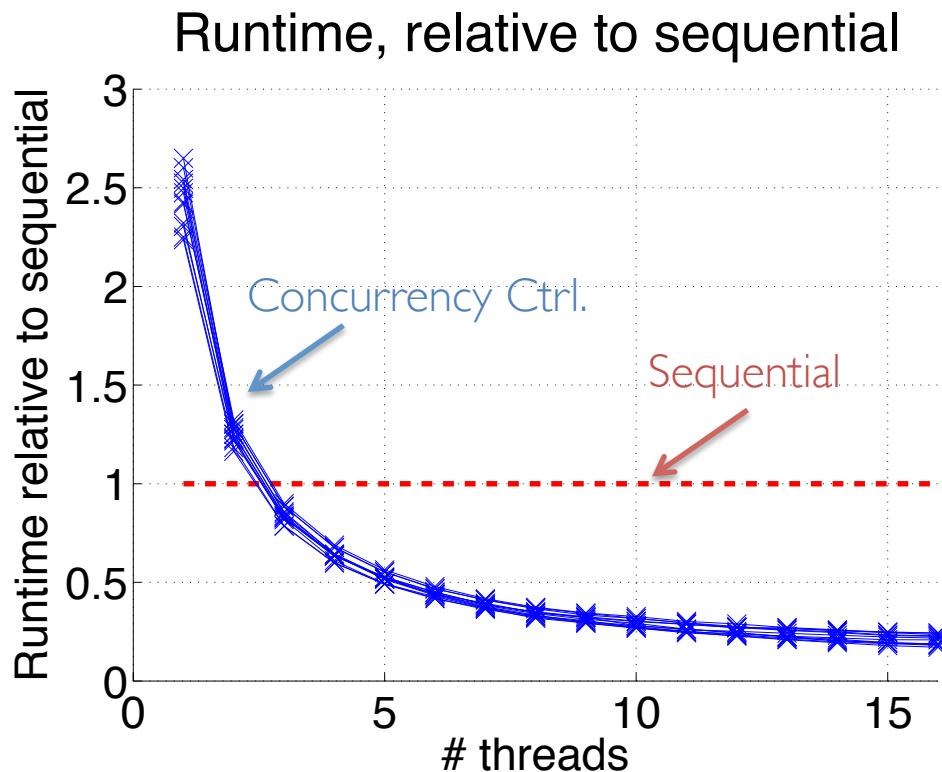
Real and synthetic graphs

Graph	Vertices	Edges
IT-2004	41 Million	1.1 Billion
UK-2005	39 Million	0.9 Billion
Arabic-2005	22 Million	0.6 Billion
Friendster	10 Million	0.6 Billion
Erdos-Renyi	20 Million	2.0 Billion
ZigZag	25 Million	2.0 Billion

CC Double Greedy Coordination



Runtime and Strong-Scaling



Conclusion

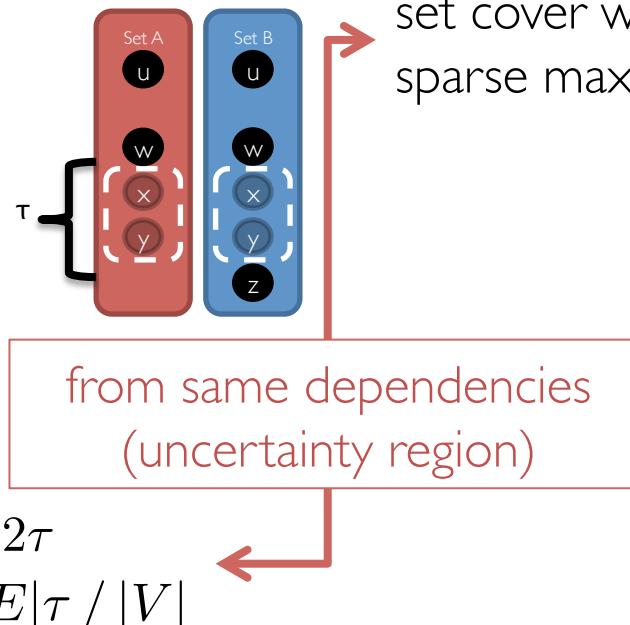
	Scalability	Approximation
Sequential Double Greedy	Always slow	Always optimal
Concurrency Control Double Greedy	Usually fast	Always optimal
Coordination Free Double Greedy	Always fast	Near optimal

Paper @ NIPS 2014:
Parallel Double Greedy Submodular Maximization.

BACKUP SLIDES

Concurrency Control

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



Lemma:
bounded overhead.

$$\begin{aligned} \text{set cover with costs: } & 2\tau \\ \text{sparse maxcut: } & 2|E|\tau / |V| \end{aligned}$$

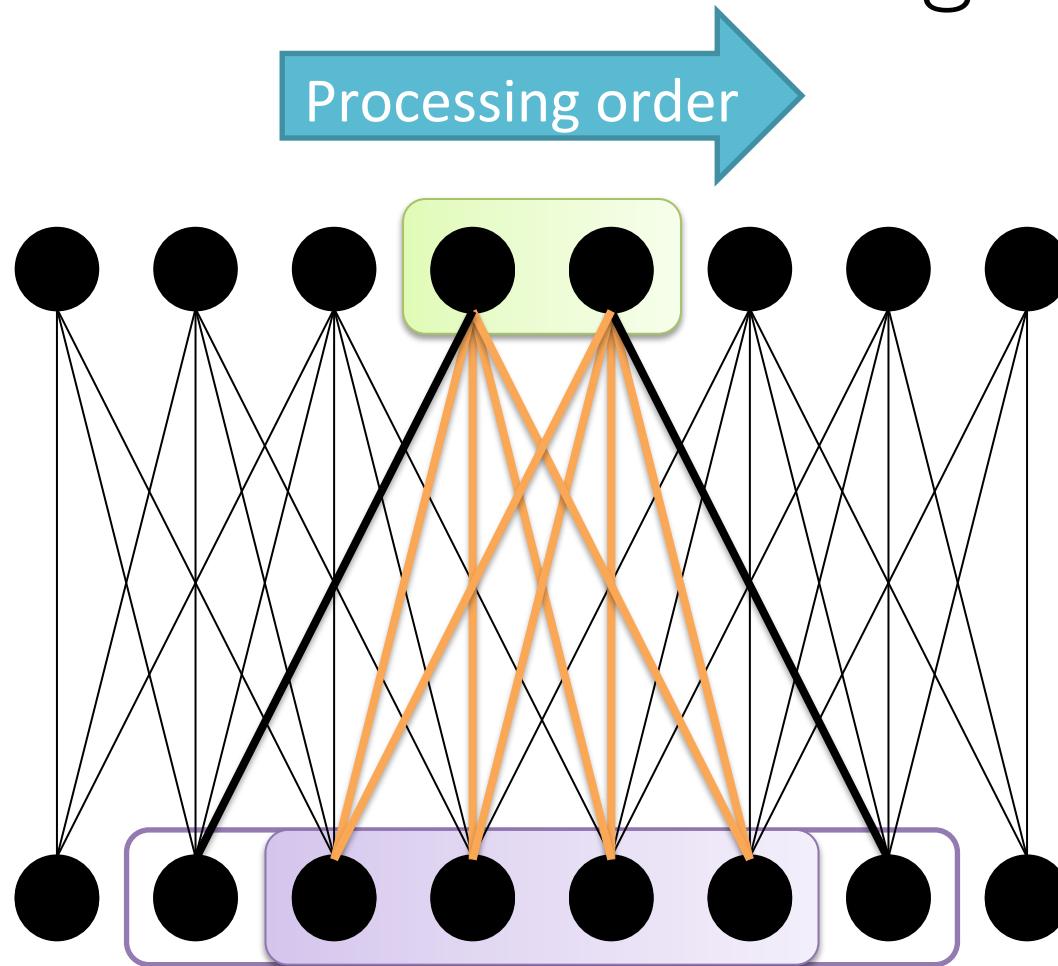
Coordination Free

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

$$\begin{aligned} \text{set cover with costs: } & \geq \tau \\ \text{sparse maxcut: } & |E|\tau / 2|V| \end{aligned}$$

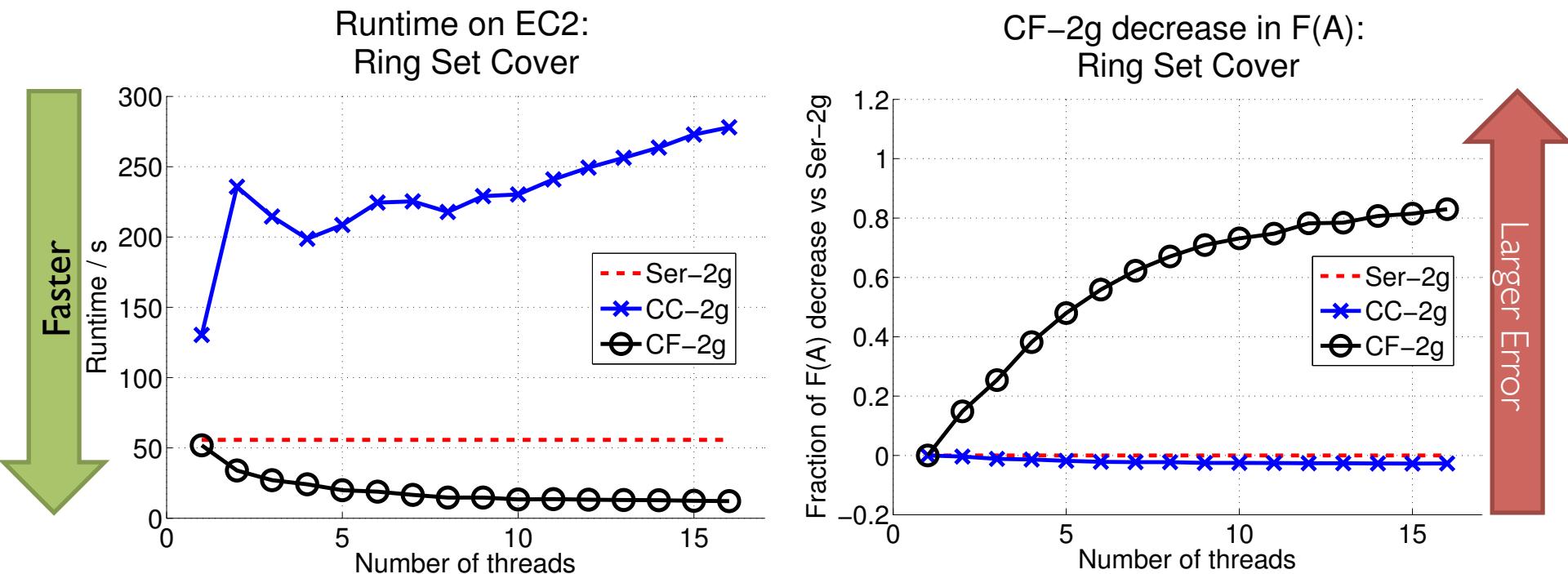
no
overhead

Adversarial Setting



Overlapping covers
→ Increased coordination

Adversarial Setting – Ring Set Cover



Coord Free	Always fast	Possibly wrong
Conc Ctrl	Possibly slow	Always optimal