

ROBUST & SPECULATIVE BYZANTINE RANDOMIZED CONSENSUS WITH CONSTANT TIME COMPLEXITY IN NORMAL CONDITIONS

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CONSENSUS

- Fundamental problem in distributed computing
 - Examples: SM Replication, Leader Election, Coordination, Group Membership, etc.
- Impossible to attain deterministically with crash-faults (partial correctness)
- Termination achievable with:
 - weaker models (ev. synchrony assumption)
 - randomization (almost-surely)

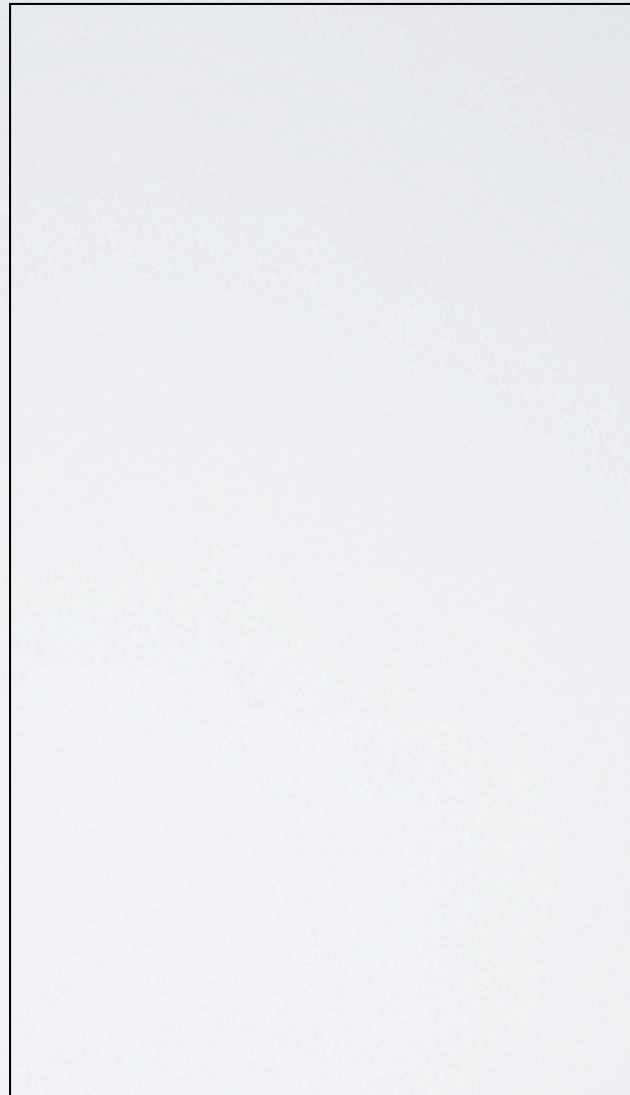
RANDOMIZED CONSENSUS

- Properties
 - *Validity: if all correct processes propose v , then v is the only possible decision*
 - *Agreement: no two correct processes decide differently*
 - *Probabilistic Termination: all correct processes eventually decide with probability **1***
- Assumptions
 - Reliable channels
 - Source-authenticated channels

BRACHA'S ALGORITHM

(PODC 1984)

- Seminal algorithm
- Asynchronous
- Byzantine resistant
- Resilient-optimal ($3f+1$)
- Correct under the Strong Adversary model



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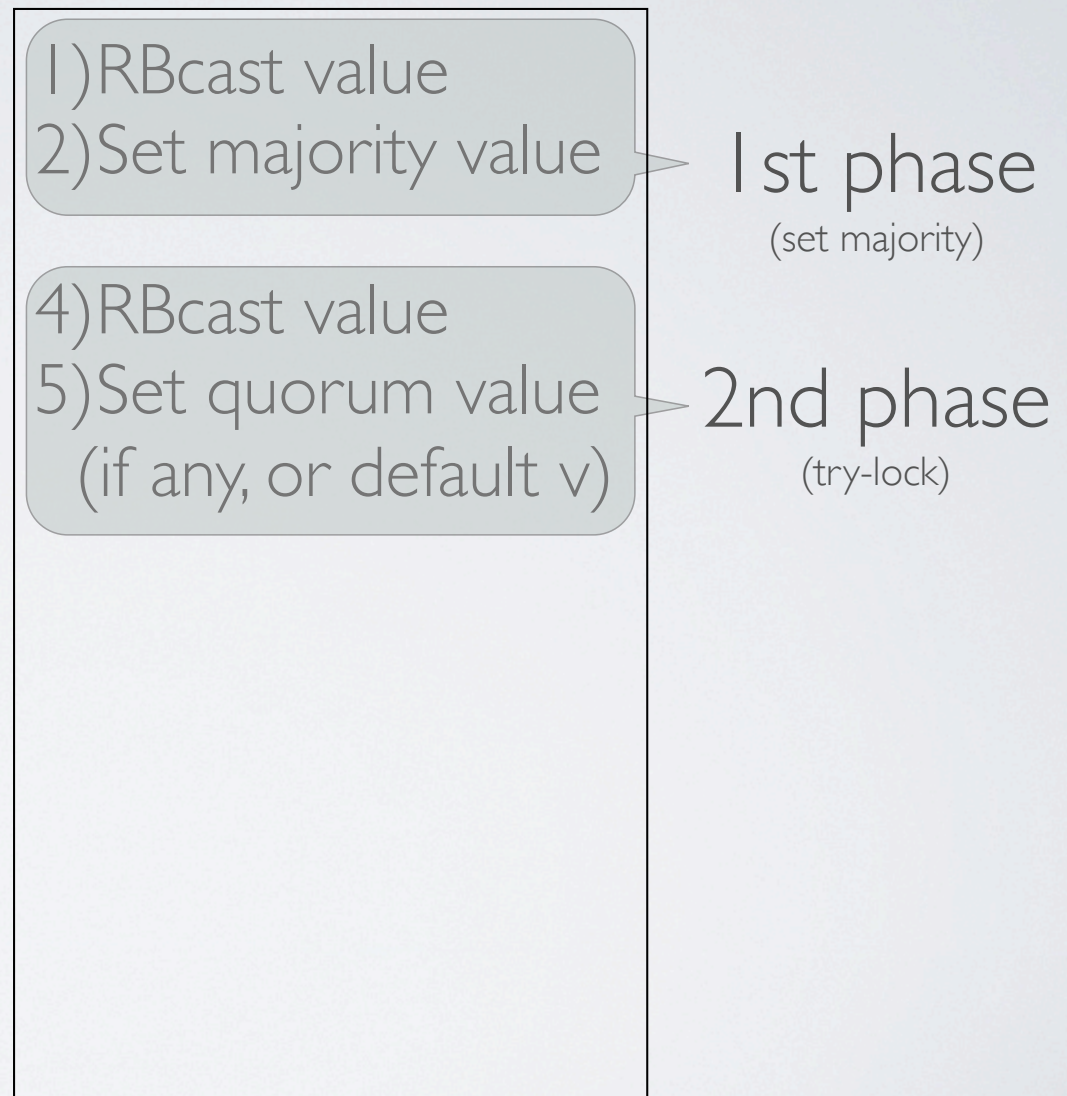
1) RBcast value
2) Set majority value

1st phase
(set majority)

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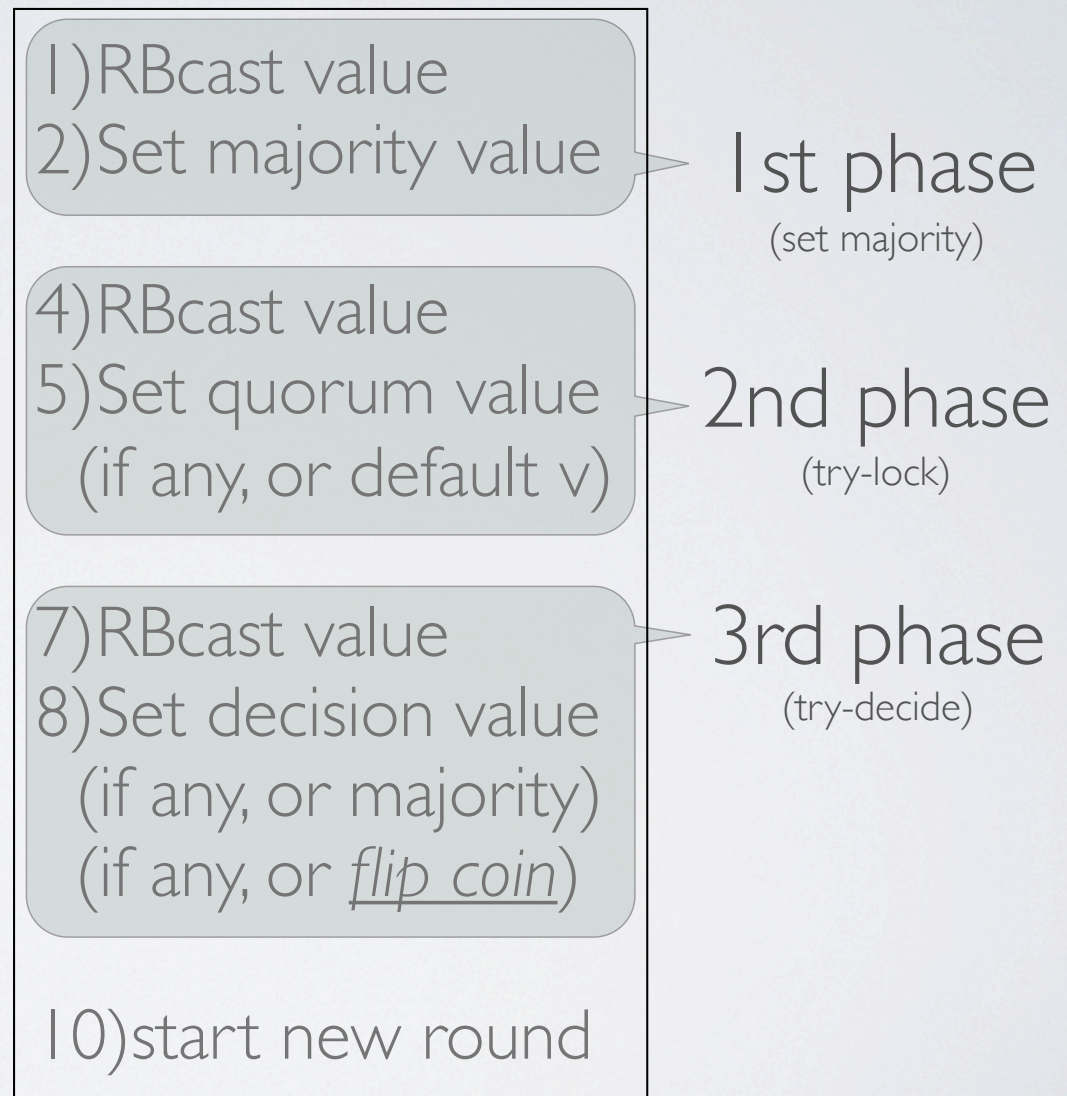
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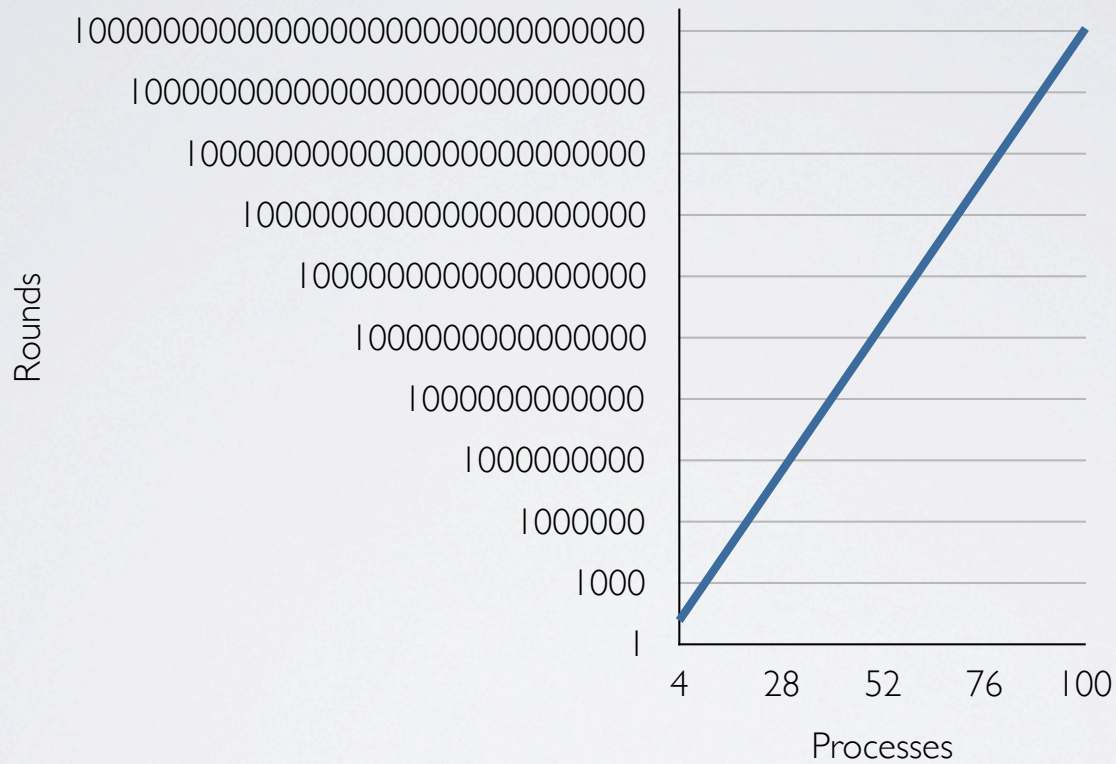
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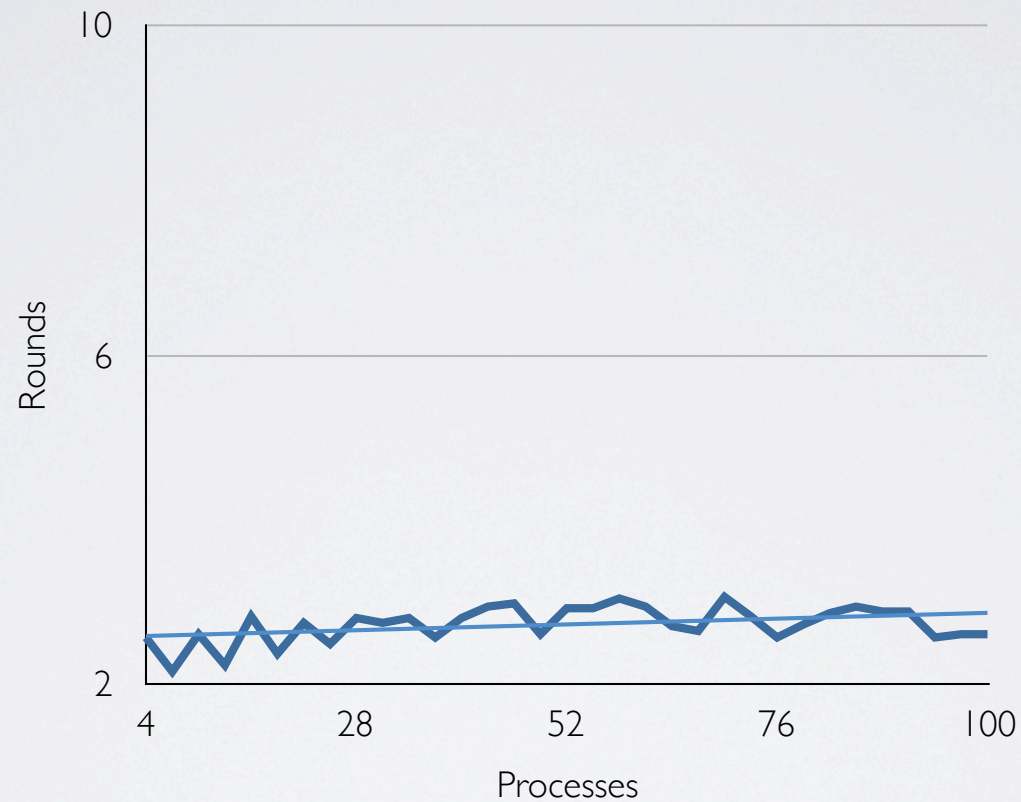


IN THEORY



Potential problem: expected exponential time execution under adverse conditions

IN PRACTICE



In reality: it terminates in a constant number of rounds under normal conditions

RELIABILITY VS. PERFORMANCE WHAT'S THE MODEL?

WHAT IS NORMAL?

- Asynchrony?
- Crash failures?
- Byzantine failures?
- Content-independent message scheduler?
- Full information adversary?
- Adversary message scheduler?



AN EXPERIMENT

PROBABILISTIC ANALYSIS

INGREDIENTS

- Hypergeometric distribution

$$\mathcal{H}(n, k, n - f)$$

- Binomial distribution

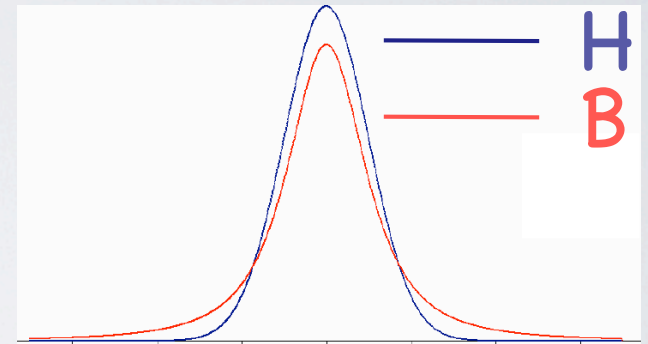
$$\mathcal{B}(n, p)$$

- Normal distribution

$$\mathcal{N}(np, np(1 - p))$$

- Some approximations

$$P(\mathcal{B}(n, p) \leq i) \approx \Phi\left(\frac{i - np}{\sqrt{np(1 - p)}}\right)$$



GOING BACKWARDS

- decision on v

message exchange

3rd Phase

- Linear bias of constant $1/4$ of procs proposing v

message exchange

2nd Phase

- Procs have constant probability of setting v
- Linear bias of (just) **positive** constant beyond the average

message exchange

1st Phase

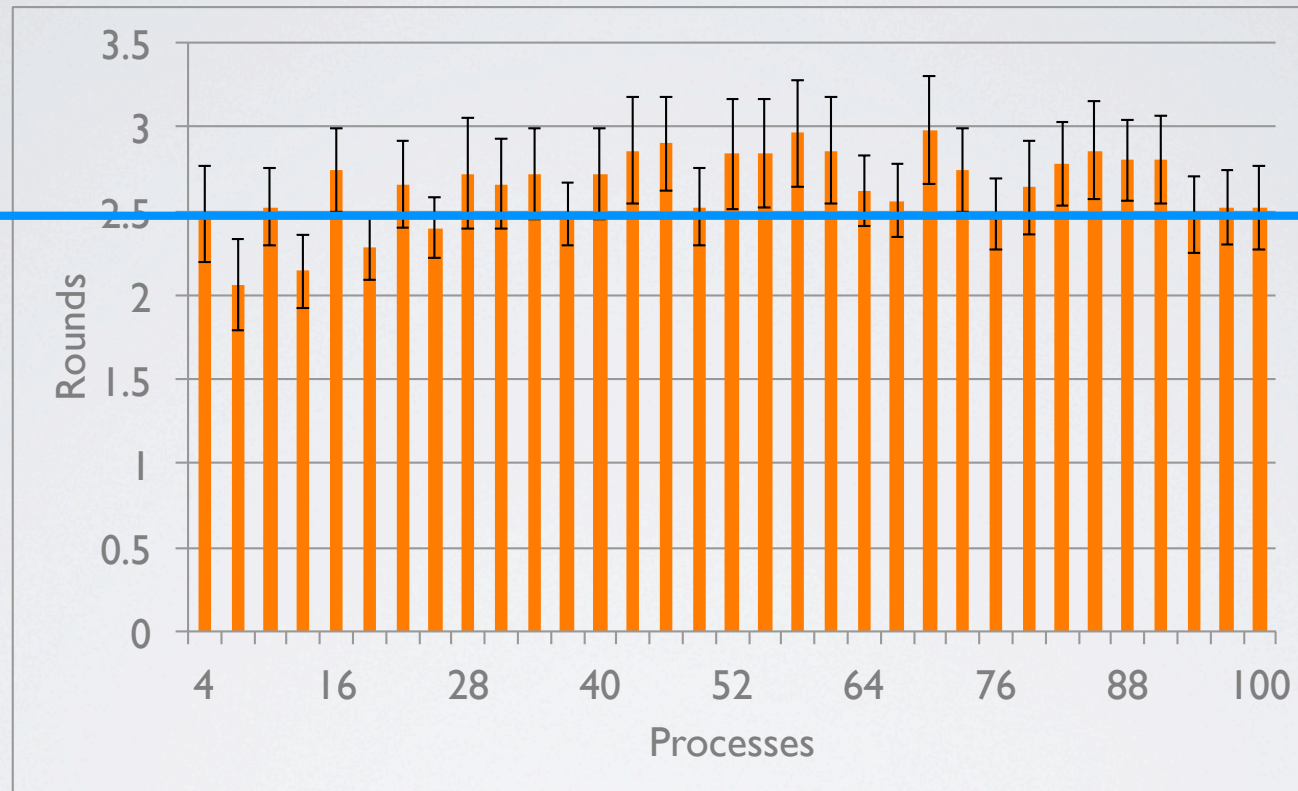
- Square root bias beyond the average of procs proposing v
- Back to coin tossing, this is a ...

Basic property of the Normal Distribution: $p=2/5$ (or **2.5 rounds**)

EVALUATION

PERFORMANCE

2.5 rounds

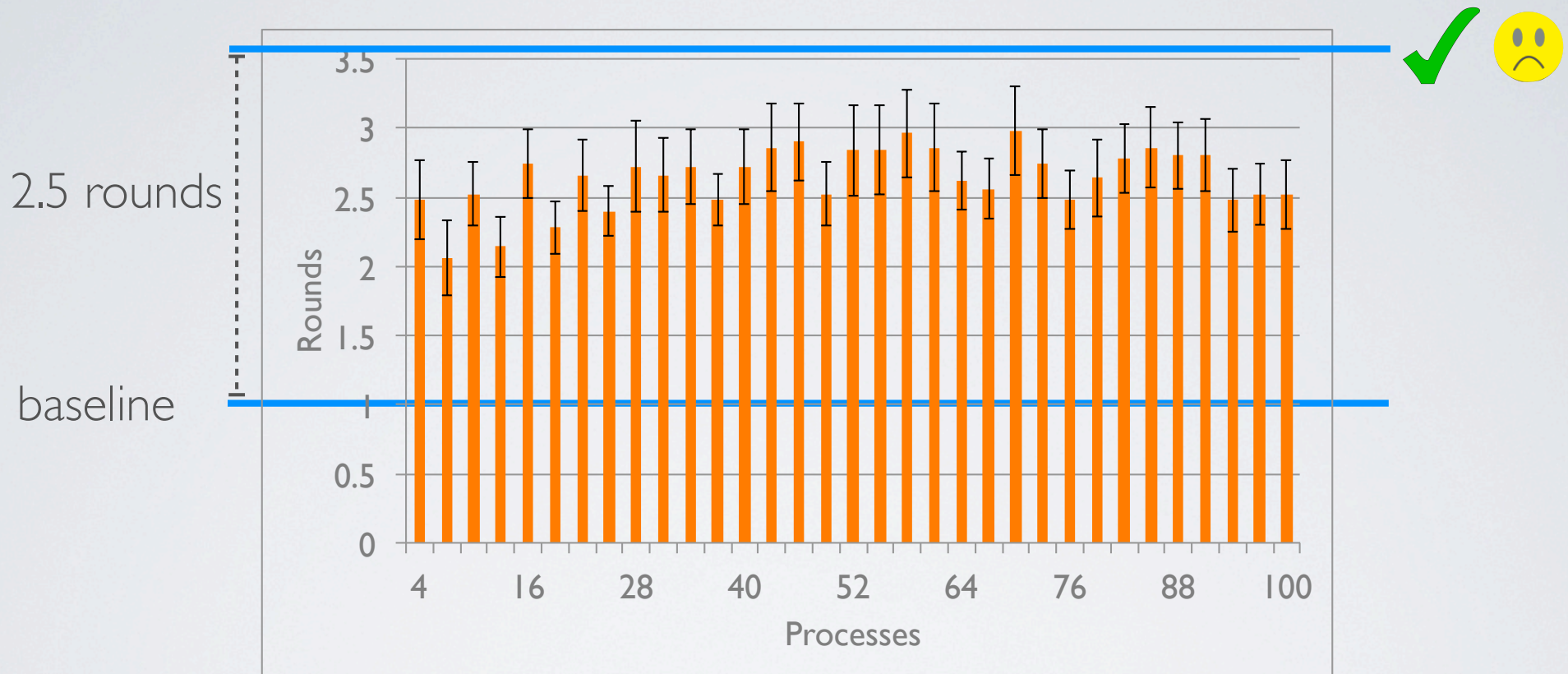


- Cluster of 6 nodes
- Up to 100 processes

- $n = 3f + 1$

- Divergent initial configuration

PERFORMANCE



- Analysis says 2.5 rounds after coin flipping
- Baseline at 1 round
- Theoretically satisfactory, but practically not precise, constant complexity

LOOK AT THE CONSTANTS

- Approximations are theoretically good
- Loss of precision when computing constant values

$$P(\mathcal{H}(n, k, n-f) \leq i) \stackrel{\text{ours}}{\leq} \Phi\left(\frac{i-\mu_{\mathcal{B}}}{\sigma_{\mathcal{B}}}\right)$$

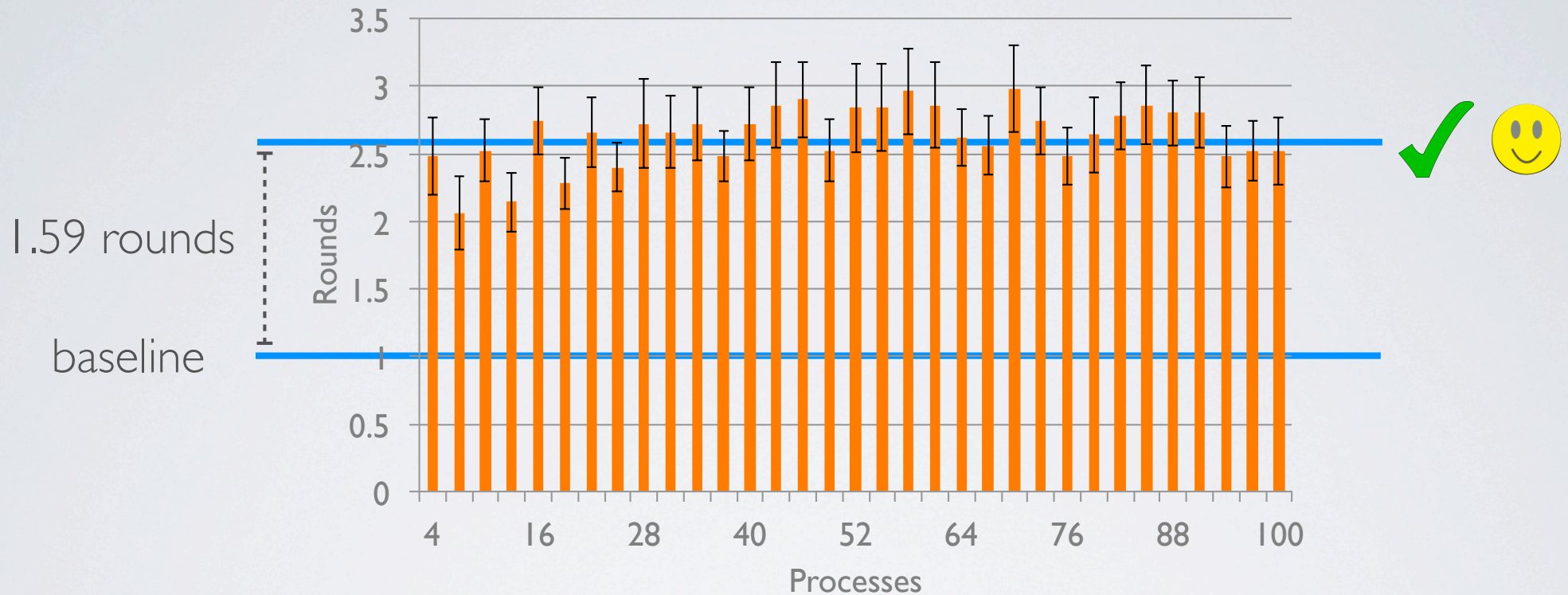
- A better approximation is available

$$P(\mathcal{H}(n, k, n-f) \leq i) \stackrel{\text{Feller}}{\approx} \Phi\left(\frac{i-\mu_{\mathcal{H}}}{\sigma_{\mathcal{H}}}\right)$$

- A multiplicative constant impacts noticeably just on constants

$$\Phi\left(\frac{i-\mu_{\mathcal{H}}}{\sigma_{\mathcal{H}}}\right) = \Phi\left(\frac{i-\mu_{\mathcal{B}}}{\sigma_{\mathcal{B}}}\sqrt{3}\right)$$

PERFORMANCE



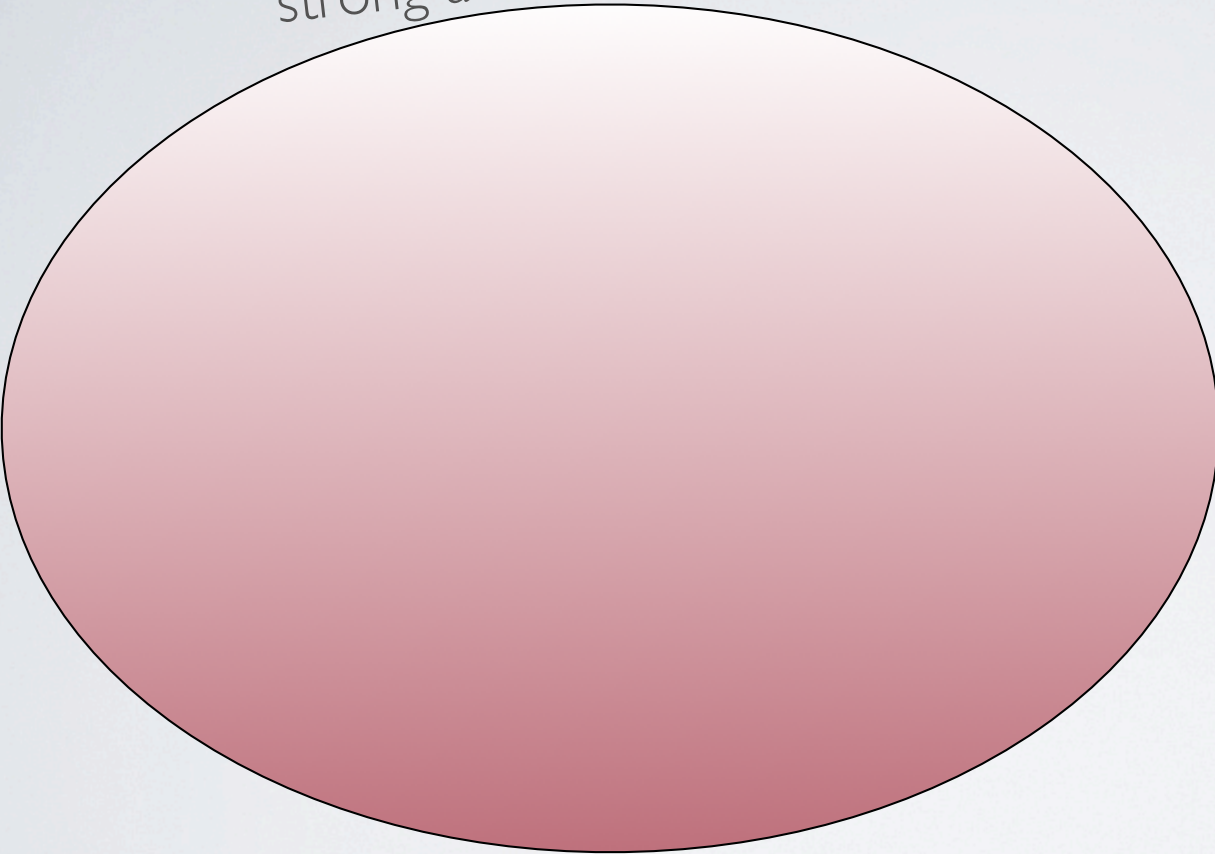
- Analysis says 1.59 rounds after coin flipping
- Baseline at 1 round
- Theoretically satisfactory and practically rather precise constant complexity

HIGH LEVEL VIEW

Model	Complexity
SA	$O(2^n)$
WA	$O(1)$
SMO	1 round

HIGH LEVEL VIEW

strong adversary



Model	Complexity
SA	$O(2^n)$
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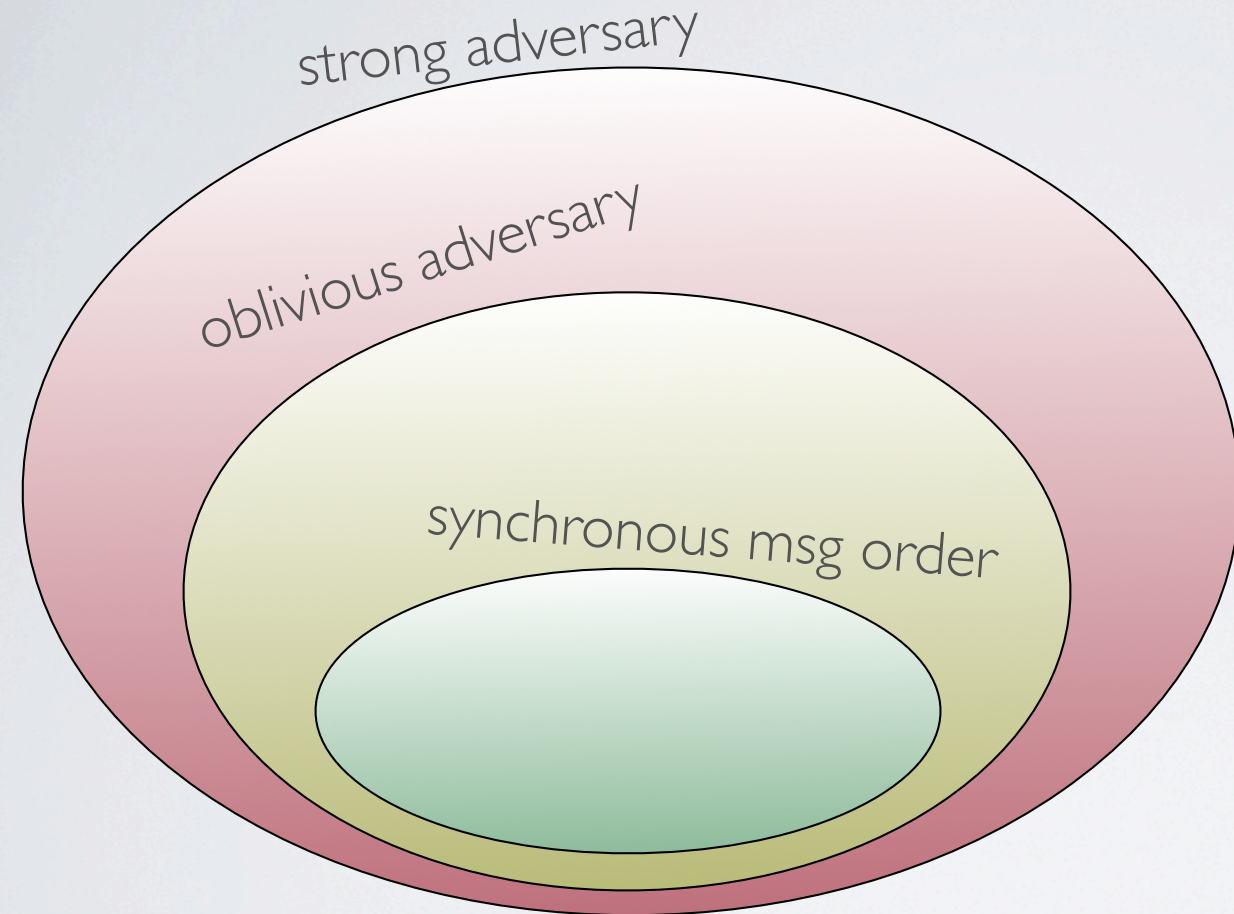
HIGH LEVEL VIEW

strong adversary

synchronous msg order

Model	Complexity
SA	$O(2^n)$
WA	$O(1)$
SMO	1 round

HIGH LEVEL VIEW



Model	Complexity
SA	$O(2^n)$
OA	$O(1)$
SMO	1 round

Complexity values are all relative to the Bracha's algorithm

LET'S GO BEYOND

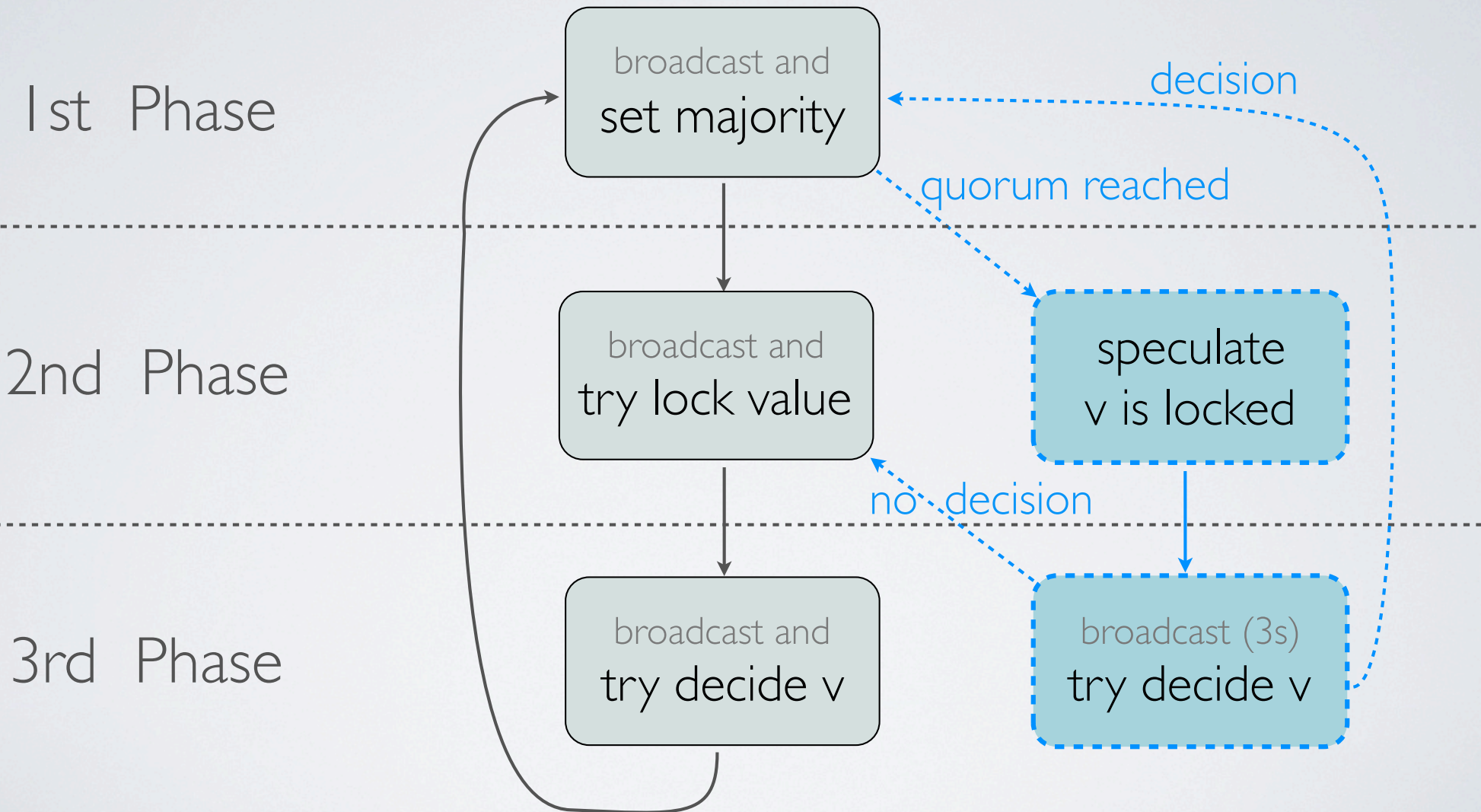
OVERVIEW

Termination in $\begin{cases} 1-2 \text{ rounds} & \checkmark \text{ (good)} \\ 3-6 \text{ phases} & \times \text{ (bad)} \end{cases}$

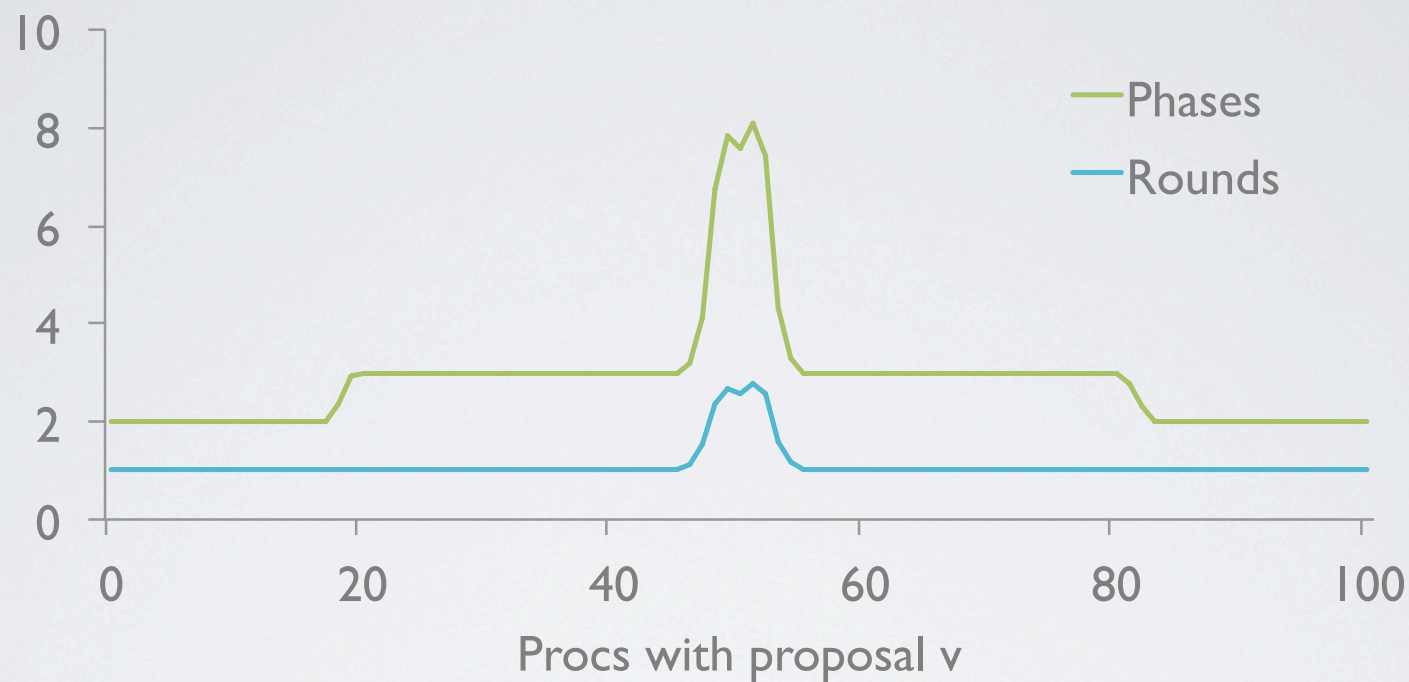
Objective: can we improve phase complexity in normal conditions while maintaining reliability?

- (oblivious) crash-failures may happen
 - Decision in 1 phase possible in a weaker model
- Focus on the set of $(n-f)$ received messages

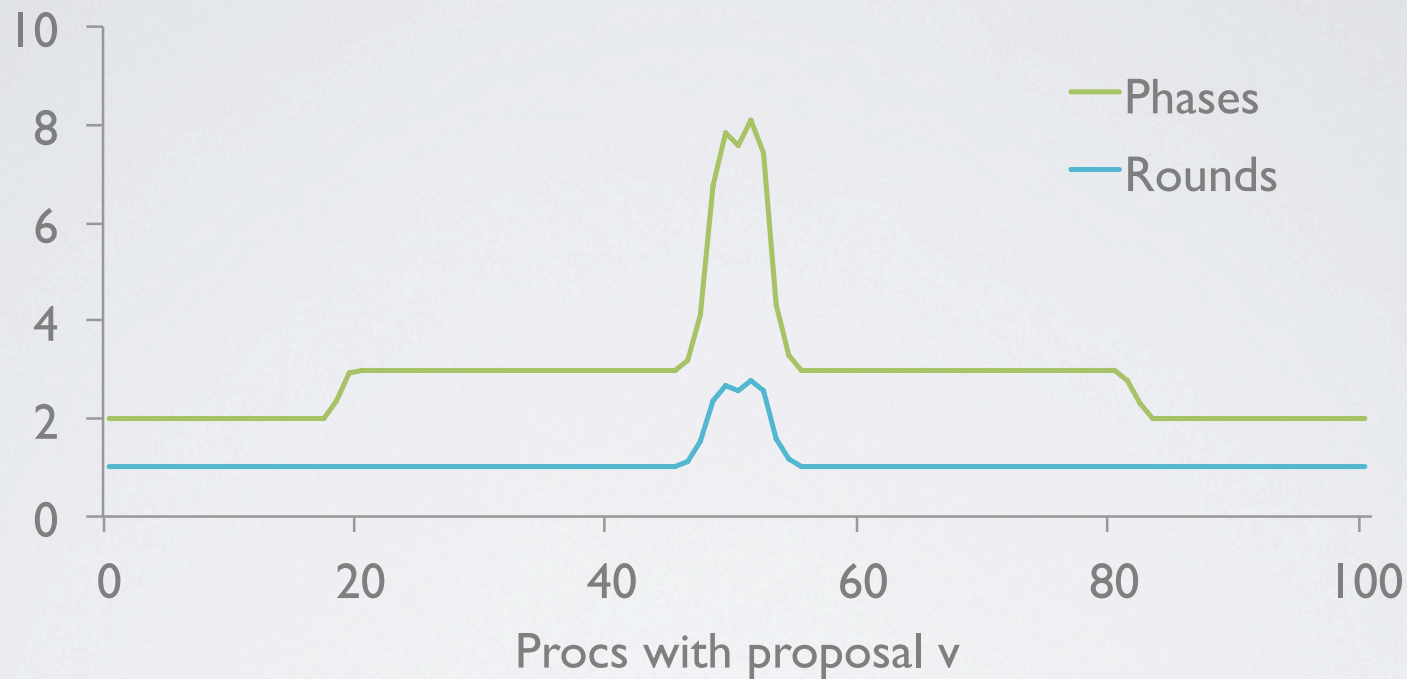
SPECULATION



PERFORMANCE



PERFORMANCE



- 2-phase termination more frequent with more msgs

BENEFITS AND DRAWBACKS

PROs	CONs
2 phases/round in the best case	Algorithm complexity increased due to speculation
3 phases/round if speculation fails	Fragile for near divergent proposals
Does not compromise original algorithm's properties	

SUMMARIZING

- Bracha's algorithm (PODC 1984) terminates in constant time (1.59 expected rounds) in normal conditions
 - First cross-model (non-trivial) analysis
 - Enhanced detection of anomalous/malicious behavior
- (Almost) matching upper-bound with respect to Attiya-Censor's lower bound (PODC 2008)
- Improved algorithm through inexpensive Speculation

THANK YOU!