

# HYBRID FAULT MODELS

- The material in the following slides is courtesy of Dr. Azad Azadmanesh (UNO), i.e., they were prepared using many slides he had given to me.
  - Thanks Azad!
- This discussion of fault models uses the “dependability” fault assumptions and does not focus on malicious act. We will discuss what that really means in class.

# REFERENCES

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- A. Azadmanesh, R.M. Kieckhafer, “New Hybrid Fault Models for Asynchronous Approximate Agreement”, IEEE Transactions on Computers, Vol. 45, No. 4, pp. 439-449, April 1996.
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# OVERVIEW

- The Byzantine Fault Model (BYZ-1)
- Hybrid Fault Models
  - MPH-2
  - TPH-3
  - OTH-5
- Taxonomy of Fault Models
- Advantages of Hybrid Fault Models
- Summary of Fault-Tolerance

# BYZANTINE FAULT MODEL

- Assumes that every fault is Byzantine (asymmetric)
- Oral  $t$ -fault-tolerance requires
  - $N \geq 3t + 1$  -- independent processes
  - $r \geq t$  -- rounds of re-broadcast
- Observations (in the classic dependability sense):
  - Asymmetry is hard to achieve (especially with busses).
  - Asymmetric faults are very rare.
  - The majority of faults are very likely to be less severe.
  - Designs based on this model are likely to be strongly over-conservative.
  - Designs based on this model are normally used for applications which can not afford to tolerate any form of faults.

# HYBRID FAULT MODELS

- Objective of Hybrid Fault Modeling is:
  - A fault model with fault “modes” of different severities.
  - Fault modes must be:
    - Mutually Exclusive
    - Collectively Exhaustive
- Motivation:
  - Allow a set of coincident faults to be of mixed modes (severities).
  - Design systems which are not so over-conservative.
  - Analyze agreement algorithms in the presence of mixed-mode faults.

# MPH-2 FAULT MODEL (MEYER/PRADHAN)

- Partitions all faults into:
  - Benign faults
  - Malicious faults
- Benign fault: Immediately self-evident to all receivers.
  - Can not send an erroneous value that can not be detected.
  - Can not cause the system to undergo incorrect state transition.
  - Also called self-incriminating, fail-notify, dormant, manifest.
  - Must be pre-agreed upon by all non-faulty nodes:
    - Not immediately evident in all authors' definitions
    - Embedded in assumptions of proofs.
    - All nodes simply ignore benign faults.

# MPH-2 FAULT MODEL

- Examples of Benign faults:
  - Out of bound data
  - Timing-fault
  - Crash fault
- Malicious fault: Does not identify itself as faulty:
  - Any faulty behavior that is not considered benign.
  - Not self-evident to all nodes.
  - Can behave symmetrically or asymmetrically.
  - Must be handled as Byzantine.

# MPH-2 FAULT MODEL

- Total Number of Faults
  - $m$  = number of malicious faults
  - $b$  = number of benign faults
  - Thus total number of faults  $t = m + b$
- Oral Message Fault-Tolerance:  $OM(m)$
- $N \geq 3m + b + 1$  -- independent processes
  - $r \geq m$  -- rounds of re-broadcast



# TPH-3 FAULT MODEL

- Thambidurai & Park model
- BenignFault: same as MPH-2
- Symmetric Malicious:
  - Can send an erroneous value.
  - Must broadcast the same error to all receivers.
- Asymmetric Malicious:
  - Can send different errors to different receivers.

# TPH-3 FAULT MODEL

- Total Number of Faults:
  - $a$  = Number of asymmetric faults
  - $s$  = Number of symmetric faults
  - $b$  = Number of benign faults
  - Thus: Total number of faults  $t = a + s + b$
- Oral Message Fault-Tolerance:  $OM(a)$ 
  - $N \geq 2a + 2s + b + r + 1$
  - $r \geq a$

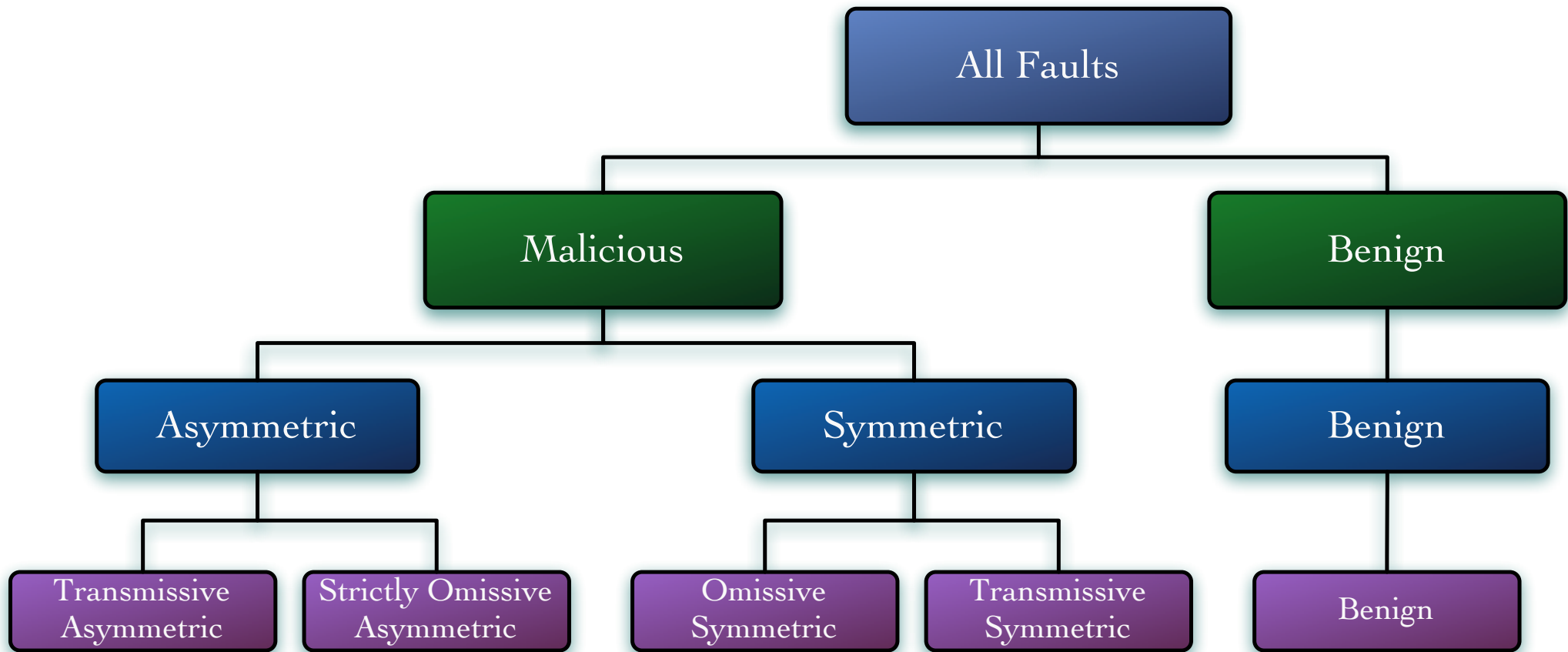
# OTH-5 FAULT MODEL

- OTH stands for Omissive & Transmissive Hybrid model
- Benign: Same as TPH-3 and MPH-2 benign
- Omissive Symmetric:
  - Does not send a value to any receiver.
  - Unlike a benign fault, it is not previously diagnosed and agreed to by all nodes.
- Transmissive Symmetric: same as TPH-3 symmetric.
- Strictly Omissive Asymmetric: Can transmit only:
  - Correct value to some receivers.
  - And No value to other receivers.
- Transmissive Asymmetric: Same as TPH-3 asymmetric.

# OTH-5 FAULT MODEL

- Total Number of Faults
  - $a'$  = Number of transmissive asymmetric faults
  - $\omega_a$  = Number of strictly omissive asymmetric faults
  - $s'$  = Number of transmissive symmetric faults
  - $\omega_s$  = Number of omissive symmetric faults
  - $b$  = Number of benign faults
  - Thus total number of faults:  $t = (a' + \omega_a) + (s' + \omega_s) + b$
- Oral Message Fault-Tolerance
  - OTH has not been applied to Byzantine Agreement
  - Has been applied to “Approximate” Agreement:
    - $N \geq 3a' + 2s' + \omega_a + \omega_s + b + 1$

# TAXONOMY OF FAULT MODELS



# FT ADVANTAGES OF HYBRID FAULT MODELS

- 1) More accurate model of the system: Less “overly” conservative
- 2) Resulting reliability estimates are better.

Consider  $N = 6$ :

OM is capable of tolerating 1 fault.

Whereas TPH-3 can tolerate: 1 asymmetric and 1 symmetric fault, or  
1 asymmetric and 2 benign faults.

Also, with  $N=4$ , OM is capable of still tolerating 1 fault.

Therefore the system reliability when  $N=6$  is less, and

Thus, it is better to turn off the additional nodes.

# FT ADVANTAGES OF HYBRID FAULT MODELS

## 3) Smarter degradation

Consider the incompatible fault scenarios:

Scenario 1:  $a=1, s=0, b=0$

Scenario 2:  $a=0, s=2, b=0$

Consider a 5-node system:

$r=0$  can handle scenario 2, but not scenario 1

(simple majority vote), (see N for TPH-3).

$r=1$  can handle scenario 1, but not scenario 2.

Thus: By specifying the number of rounds, the algorithm can be “tuned” for the most likely scenarios.

# FT ADVANTAGES OF HYBRID FAULT MODELS

- Requirements for success:
  - To have a good estimate of fault rates  $\lambda_a$  ,  $\lambda_s$  ,  $\lambda_b$ 
    - Typically  $\lambda_a \ll \lambda_s \ll \lambda_b$
  - To have a good estimate of recovery rates  $\rho_a$  ,  $\rho_s$ 
    - Typically  $\rho_a < \rho_s$



# SUMMARY

- BYZ-1
  - $N \geq 3t + 1$
  - $r \geq t$
- MPH-2
  - $N \geq 3m + b + 1$
  - $r \geq m$
- TPH-3
  - $N \geq 2a + 2s + b + r + 1$
  - $r \geq a$
- OTH-5
  - Byzantine Agreement: Unknown at this time. Approximate Agreement:  
 $N \geq 3a' + 2s' + \omega a + \omega s + b + 1$

# DATA AGGREGATION

- There are different yet similar definitions:
  - Ability to provide global information for purposes of network management and user services.
  - A set of functions that provide components of a distributed system access to global information.
- Reasons to do Data Aggregation (DA)
  - To coordinate tasks.
  - The need for node/component duplication for higher performance.
  - The need for redundancy for higher fault tolerance.

# DATA AGGREGATION

- Data Aggregation has gone by other names:
  - Data Fusion
  - Approximate Agreement
  - Consensus
  - Distributed Agreement

# DATA AGGREGATION

- General Scenario
  - Each node in the network holds a value.
  - Nodes exchange values to decide on values.
- Agreement conditions:
  - Agreement: The decided values, one value by each node, are within a predefined tolerance of each other.
  - Validity: The decided values are within the range of initial values held by non-faulty nodes.

# A ROUND OF COMMUNICATION

- Agreement is reached executing multiple rounds.
- Each round consists of:
  - 1. Broadcast: Each node broadcasts its value to others.
  - 2. Collect: Each node forms a multiset of values.
  - 3. Filter: Select values to vote with.
  - 4. Average: Average the selected values to vote.

# MAJOR EVOLUTIONS

- Static networks
  - Multiprocessor/multiprocessing systems
    - Mostly concentrated on Fully Connected Networks (FCN)
    - Small scale critical systems: power plants, aircrafts, automobiles
  - Distributed Networks
    - Limited focus on Partially Connected Networks (PCN)
    - Large scale distributed systems (Internet)

# MAJOR EVOLUTIONS

- Mobile/Adhoc networks
  - Include both FCN and PCN
  - Growing PCN applications: tactical military operations, tracking endangered species, Unmanned Autonomous Vehicles (UAV), etc.

# FULLY CONNECTED NETWORKS

- Most agreement algorithms are devised for FCNs.
- Full exchange of values in a round is immediate.
- Diameter of values shrinks in each round (single-step).
- Agreement is reached gradually by shrinking diameter in each round.



# EXAMPLE: MEDIAN SELECT ALGORITHM

- 5 nodes with real values in  $[0, 1]$
- 4 (of 5) correct nodes possess the values:  $\langle 0, 0, 1, 1 \rangle$
- The faulty node behaves asymmetrically.
- After broadcasting, collecting, filtering, and selecting:
  - Node i:  $\langle 0, 0, 0, 1, 1 \rangle$  thus median = 0
  - Node j:  $\langle 0, 0, 1, 1, 1 \rangle$  thus median = 1
- Result:
  - Validity: voted values are within range of correct values.
  - Not convergent: values are still within  $[0, 1]$ .

# EXAMPLE: MIDPOINT SELECT ALGORITHM

- 4 nodes with real values in  $[0, 1]$
- 3 (of 4) correct nodes possess the values:  $\langle 0, 1, 1 \rangle$
- The faulty node behaves asymmetrically.
- After broadcasting, collecting, filtering, and selecting:
  - Node i:  $\langle 0, 0, 1, 1 \rangle$  thus midpoint = 0.5
  - Node j:  $\langle 0, 1, 1, 1 \rangle$  thus midpoint = 1
- Result:
  - Validity: voted values are within range of correct values.
  - Convergent: voted values are  $<$  diameter in  $[0, 1]$ .

# HOW TO ACHIEVE AA CONDITIONS

- Three general approaches:
  - Mean Select Reduce (MSR)
  - MSR algorithms: Remove the extreme-end values.
  - Egocentric algorithms: Replace the extreme-end values with your own value.
  - Egophobic algorithms: Replace the extreme-end values with your own plus a predefined threshold.

# MSR: REMOVE THE EXTREME-END VALUES

- Example:
  - Assume 5 nodes with values in the range:  $0 \dots 1$
  - 4 of the nodes possess the values:  $\langle 0, 0, 1, 1, \rangle$
  - The one faulty node ( $t=1$ ) behaves asymmetrically.
  - Use “fault tolerant mean” function.
  - Case 1: Assume, the faulty node transmits a value outside of the range.

$$\begin{array}{l} \left. \begin{array}{l} 0 \\ 0 \\ 1 \\ 1 \end{array} \right\} 0 \ 1 \ 1 \left[ \begin{array}{l} 1.5 \\ 1 \\ 1 \\ 1 \end{array} \right. \\ 0 \ 1 \ 1 \rightarrow (0+1+1)/3 = 0.66 \end{array} \quad \left| \quad \begin{array}{l} \left. \begin{array}{l} -1 \\ 0 \\ 0 \\ 1 \end{array} \right\} 0 \ 0 \ 1 \left[ \begin{array}{l} 1 \\ 1 \\ 1 \\ 1 \end{array} \right. \\ 0 \ 0 \ 1 \rightarrow (0+0+1)/3 = 0.33 \end{array}$$

- Average is convergent  $[(0.66-0.33) < (1-0)]$  & valid.

# MSR: EXAMPLE CONT.

- Example, Cont.:
  - Case 2: Assume, the faulty node transmits a value within the range.

$$\left. \begin{array}{l} 0 \\ 0 \\ 0 \end{array} \right\} \begin{array}{l} 0 \\ 0.5 \\ 1 \end{array} \left[ \begin{array}{l} 1 \\ 1 \\ 1 \end{array} \right]$$

$$0 \quad 0.5 \quad 1 \quad \rightarrow \quad (0+0.5+1)/3=0.5$$

$$\left. \begin{array}{l} 0 \\ 0 \\ 0 \end{array} \right\} \begin{array}{l} 0 \\ 0.9 \\ 1 \end{array} \left[ \begin{array}{l} 1 \\ 1 \\ 1 \end{array} \right]$$

$$0 \quad 0.9 \quad 1 \quad \rightarrow \quad (0+0.9+1)/3=0.63$$

- Average is convergent  $[(0.63-0.5) < (1-0)]$  & valid.

# EGOCENTRIC: REPLACE THE EXTREME--END VALUES WITH YOUR OWN

- Example:
  - Assume 5 nodes with values in the range:  $0 \dots 1$
  - 4 of the nodes possess the values:  $\langle 0, 0, 1, 1 \rangle$
  - The one faulty node ( $t=1$ ) behaves asymmetrically.
  - Assume two correct nodes  $i$  &  $j$  performing the voting hold the values  $1$  &  $0$  respectively.
  - Use Fault Tolerant Mean function.

# EGOCENTRIC: REPLACE THE EXTREME--END VALUES WITH YOUR OWN, CONT.

- Example, Cont.:
  - Case 1: Assume, the faulty node transmits the values (-0.5 & 1.5) outside of the range.
    - node i receives: 0 0 1 1 1.5
      - $1\ 0\ 1\ 1\ 1 \Rightarrow (1+0+1+1+1)/5 = 0.8$
    - node j receives: -0.5 0 0 1 1
      - $0\ 0\ 0\ 1\ 0 \Rightarrow (0+0+0+1+0)/5 = 0.2$
  - The result is both valid and convergent.

# EGOCENTRIC: REPLACE THE EXTREME--END VALUES WITH YOUR OWN, CONT.

- Example, Cont.:
  - Case 2: Assume, the faulty node transmits values (0.2 & 0.7) within the range.
    - node i receives: 0 0 0.2 | 1 |
      - $1\ 0\ 0.2\ | \ 1\ | \Rightarrow (1+0+0.2+1+1)/5 = 0.64$
    - node j receives: 0 0 0.7 | 1 |
      - $0\ 0\ 0.7\ | \ 0 \Rightarrow (0+0+0.7+1+0)/5 = 0.34$
  - The result is both valid and convergent.



# EGOCENTRIC: REPLACE THE EXTREME--END VALUES WITH YOUR OWN, CONT.

- Example, Cont.:
  - Case 3: Assume, the faulty node transmits value 0.2 (within range) to i and 1.5 (outside of range) to j.
    - node i receives: 0 0 0.2 | |
      - | 0 0.2 | |  $\Rightarrow (1+0+0.2+1+1)/5 = 0.64$
    - node j receives: 0 0 | | 1.5
      - 0 0 | | 0  $\Rightarrow (0+0+1+1+0)/5 = 0.4$
  - The result is both valid and convergent.

# SINGLE STEP CONVERGENCE

$\mathbf{V}$  =  $\langle v_1, \dots, v_V \rangle$ , where  $V = |\mathbf{V}|$ , is the multiset of values received and sorted in ascending order.

$\rho(\mathbf{V})$  =  $[\min(\mathbf{V}), \max(\mathbf{V})]$ , is the range of values spanned by  $\mathbf{V}$ .

$\delta(\mathbf{V})$  =  $[\max(\mathbf{V}) - \min(\mathbf{V})]$ , is the diameter of  $\mathbf{V}$ .

$\mathbf{U}_{all}$  = The multiset of all correct values received by non-faulty nodes.

During each round of voting, each non-faulty node  $i$  broadcasts its initial value to all nodes including itself.

Then it collects all the values received into the voting multiset  $V_i$ .

Node  $i$  then applies a function  $F$  to  $V_i$  to attain its latest estimate (voted value) for the round, which it uses as its initial value in the next round.

# SINGLE STEP CONVERGENCE

- An algorithm is single-step-convergent if both of the following conditions are true following every round of voting:

**C1: VALIDITY**—For each nonfaulty process  $i$ , the voted value is within the range of correct values, i.e.,  $F(\mathbf{V}_i) \in \rho(\mathbf{U}_{all})$ .

**C2: CONVERGENCE**—For each pair of nonfaulty processes,  $i$  and  $j$ , the difference between their voted values is strictly less than the diameter of the correct values received, i.e.,  $|F(\mathbf{V}_i) - F(\mathbf{V}_j)| \leq C\delta(\mathbf{U}_{all})$ , where  $0 \leq C < 1$ .

# SINGLE STEP CONVERGENCE

In convergence condition C2 above, i.e.,

$$|F(\mathbf{V}_i) - F(\mathbf{V}_j)| \leq C\delta(\mathbf{U}_{all})$$

parameter  $C$ , called the Convergence Rate, is the primary performance measure of a voting algorithm. The constraint  $0 \leq C < 1$  ensures that the algorithm is indeed single-step convergent. The smaller the values of  $C$ , the faster the voted values converge.

# MSR ALGORITHMS

- Mean Select Reduce (MSR)
- MSR algorithms can handle the following fault modes:
  - Benign (b)
  - Symmetric (s)
  - Asymmetric (a)
  - Thus:  $t = a + s + b$

# MSR ALGORITHMS

$N$  = The total number of processes in the system.

$\tau$  = The maximum number of malicious errors that could be received by a process. This number is known a priori and is identical for all nonfaulty processes.

$\mathbf{V}_i$  = The multiset of values received in a given round by nonfaulty process  $i$ . The number of elements in  $\mathbf{V}_i$  is  $V_i = |\mathbf{V}_i|$ .

$\mathbf{M}_i = \text{Red}^\tau(\mathbf{V}_i)$ , the *Medial Multiset* of  $\mathbf{V}_i$ . The number of elements in  $\mathbf{M}_i$  is  $M_i = |\mathbf{M}_i| = V_i - 2\tau$ .

$\mathbf{S}_i = \text{Sel}_{\sigma_i}(\mathbf{M}_i) = \text{Sel}_{\sigma_i}(\text{Red}^\tau(\mathbf{V}_i))$ , the *Selected Multiset* generated by  $F(\mathbf{V}_i)$ . The number of elements in  $\mathbf{S}_i$  is  $\sigma_i = |\mathbf{S}_i|$ .

# MSR ALGORITHMS

The voting function:

$$F(V) = \text{mean}[\text{Sel}_\sigma (\text{Red}^\tau (V))]$$

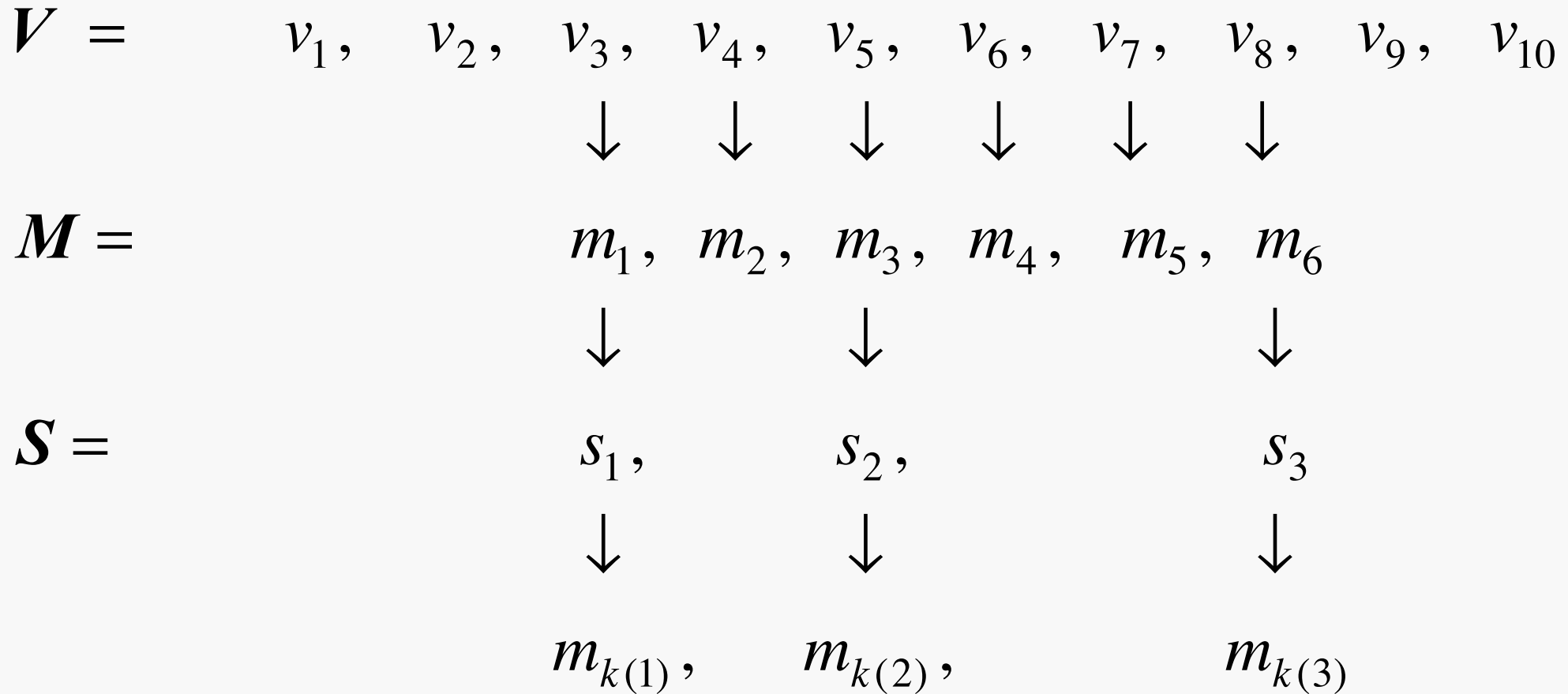
where

$\tau = (a + s)$ , maximum number of erroneous values

$\text{Red}^\tau$  = The subsequence  $M$  remaining after removing the  $\tau$  extreme values

$\text{Sel}_\sigma$  = The subsequence  $S$  of size  $\sigma$ , after selecting  $\sigma$  elements from  $M$

# EXAMPLE



$$\Rightarrow s_i = m_{k(i)}$$



## TPH-3 MODEL

- With TPH-3 we have  $N \geq 3a + 2s + b + 1$
- With Lamport we have  $N \geq 3t + 1$
- It was shown [Kieckhafer & Azadmanesh 1994] that single-step convergent MSR voting algorithms can exist if

$$N \geq 3a + 2s + b + 1 = 3t - (s + 2b) + 1$$

- Examples of algorithms that achieve this bound are
  - Fault-Tolerant Midpoint (if  $a > 0$ )
  - Simple Median Select (if  $a = 0$ )

## DEFINITION

$g, h \in \{1, \dots, \sigma\}$ , where  $g \leq h$ :

$(h - g) \equiv$  the distance between  $s_g$  and  $s_h$ ,

$k(h) - k(g) \equiv$  the distance between  $m_{k(g)}$  and  $m_{k(h)}$ ,

Define:

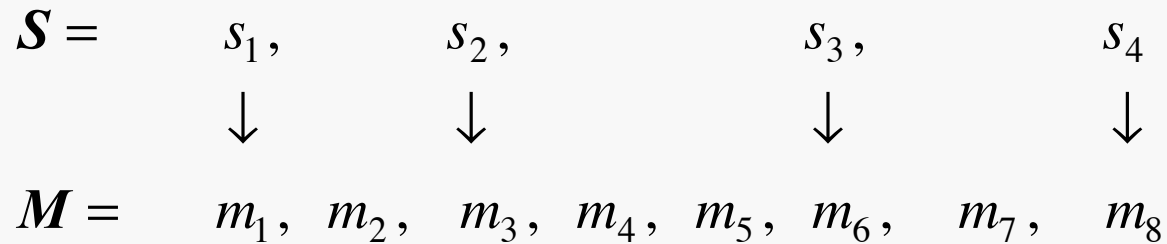
$\gamma_\alpha \equiv \min(h - g)$  such that  $k(h) - k(g) \geq \alpha$ ,

$\forall g, h \in \{1, \dots, (\sigma - \gamma_\alpha)\}$

$\gamma_\alpha \equiv$  the minimum distance between any two elements of  $\mathcal{S}$  which guarantees that the distance between the corresponding elements of  $\mathcal{M}$  is at least  $\alpha$ .

# EXAMPLE

Let  $\mathbf{S} = \langle s_1, s_2, s_3, s_4 \rangle = \langle m_1, m_3, m_6, m_8 \rangle$



If  $\alpha = 2$ :

- ◆ A distance of 1 in  $\mathbf{S}$  always yields a distance of at least 2 in  $\mathbf{M}$ .
- ◆ Since  $k(g+1) - k(g) \geq 2$ ,  $\gamma_2 = (g+1) - g = 1$ .

If  $\alpha = 3$ :

- ◆ A distance of 1 in  $\mathbf{S}$  can yield a distance of less than 3 in  $\mathbf{M}$ .
- ◆ A distance of 2 in  $\mathbf{S}$  always yields a distance of 5 in  $\mathbf{M}$ .
- ◆ Since  $k(g+2) - k(g) \geq 3$ ,  $\gamma_3 = (g+2) - g = 2$ .

# DEFINITIONS

$V_i \equiv$  Multiset of values received by node  $i$ .

$U_i \equiv$  Multiset of values received by node  $i$  from non faulty nodes.

$U_{i \cap j} \equiv U_i \cap U_j$

$\delta(U) \equiv \max(U) - \min(U)$ , for some multiset  $U$ .

$C \equiv$  Convergence Rate =  $\frac{\max |F(V_i) - F(V_j)|}{\delta(U_{i \cap j})}$ ,  $\delta(U_{i \cap j}) > 0$

If  $0 \leq C < 1$ , then the algorithm is convergent.

# MSR PERFORMANCE FOR FCN

$$C = \frac{\gamma_a}{\sigma}$$

$$N \geq 3a + 2s + b + 1$$

$$V = N - b$$

$$\tau \geq a + s$$

# SUMMARY

- Hybrid Fault Models were discussed with the 5-mode model being offering special value, as they consider the impact of omissive behavior on symmetric and asymmetric faults,
  - transmissive
  - omissive
- Approximate agreement algorithms were considered in the context of their convergence rates
  - allows to see how fast algorithms converge
  - some algorithms may work fast, but convergence is not guaranteed, e.g., simple median with  $a > 0$