



**Carnegie  
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# **Gotta have HeART**

Improving storage efficiency by exploiting  
disk-reliability heterogeneity

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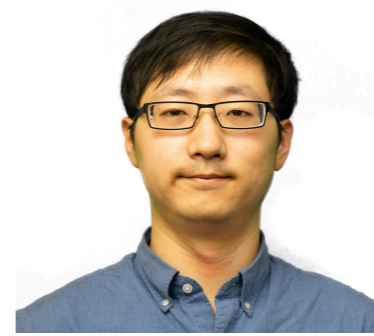
Greg  
Ganger



Francisco  
Maturana



Jason  
Yang



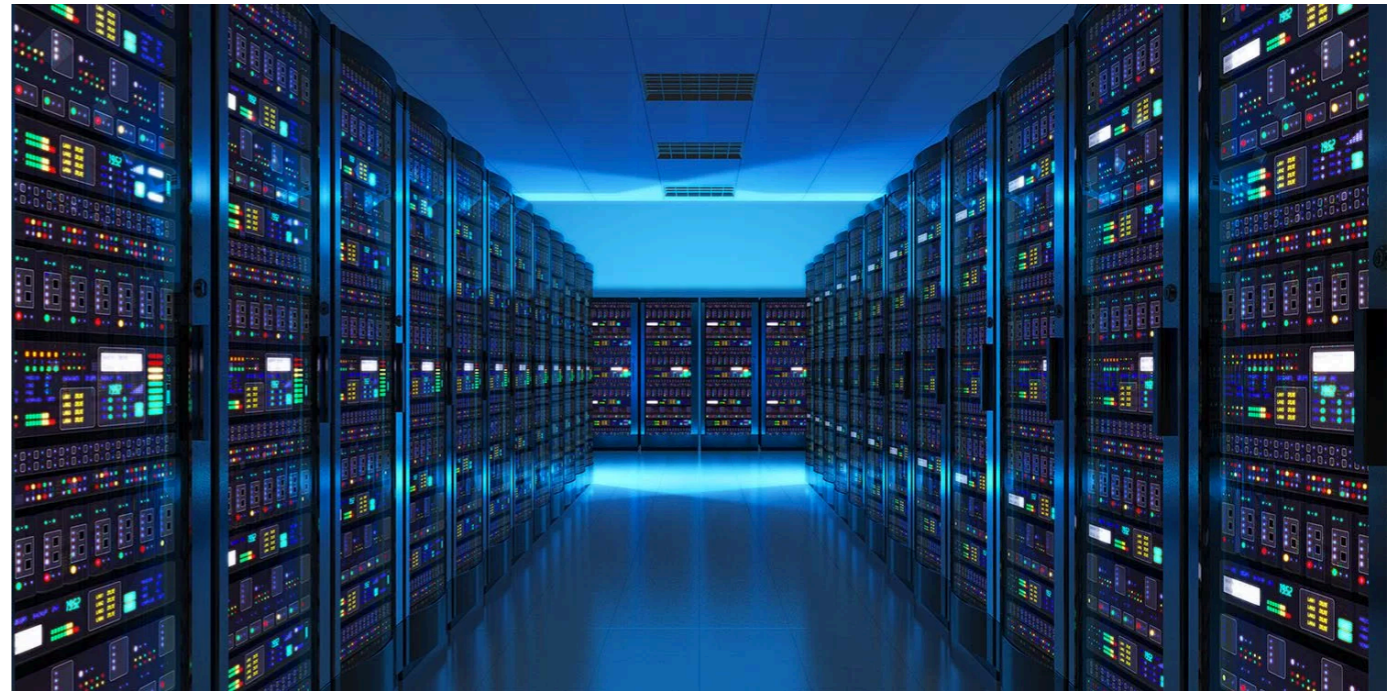
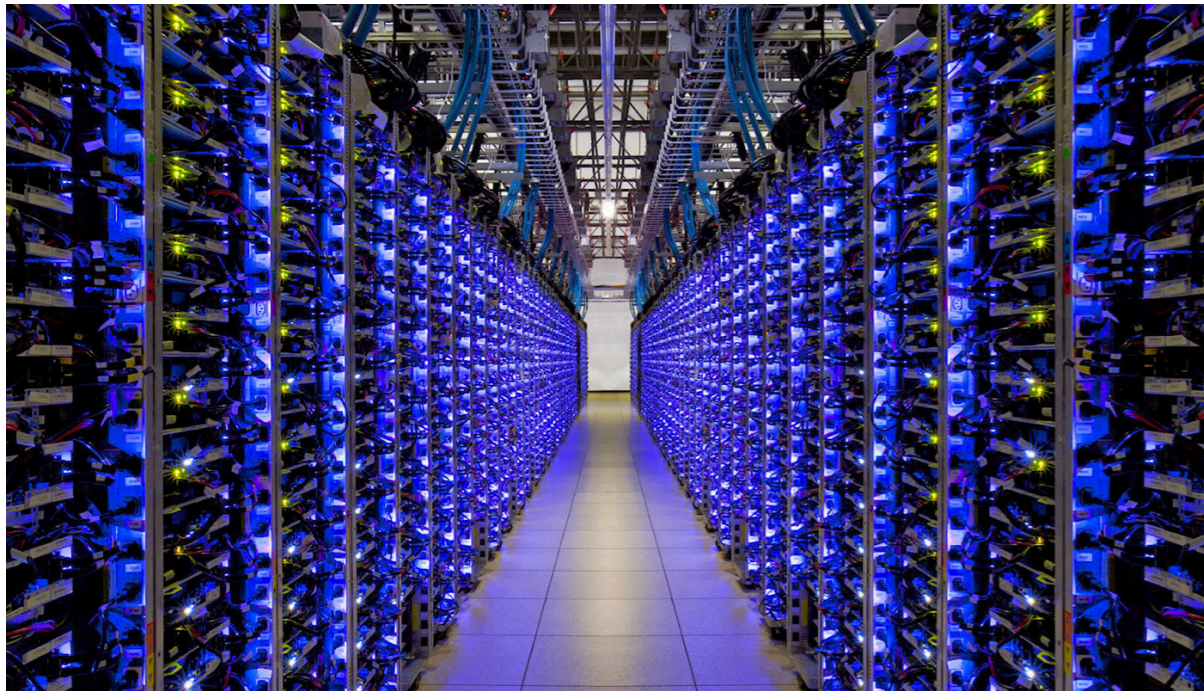
Suhas  
Jayaram



# Cluster storage systems

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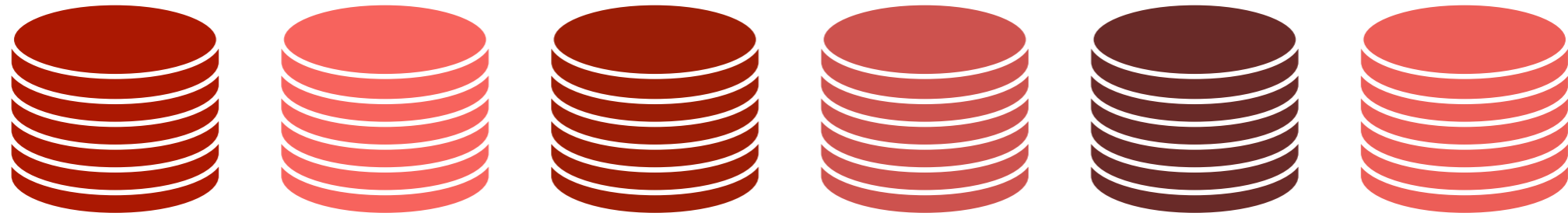
- Storage subsystem of distributed systems



- Thousands to millions of disks in primary storage tier
- Built incrementally according to demand

# Reliability heterogeneity in disks

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- Disk fleet has heterogeneous collection of disks
- **Different in reliability**
  - Across disks:
    - Manufacturing differences across makes/models
    - Experiences: different vibration / temperature / IO churn
  - For each disk:
    - 3 reliability phases throughout lifetime

# Overview of exploiting reliability heterogeneity

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- Data redundancy typically same across disk fleet
  - E.g., 3-replication: 3 copies of data on independent devices
- Disks from same storage tier vary a lot in failure rates
  - E.g., HDDs from different makes/models fail differently
- **Explicitly consider reliability heterogeneity in deciding redundancy**
- **HeART: Heterogeneity Aware Redundancy Tuner**
  - Tailors redundancy to disk failure rate heterogeneity
  - A safe, accurate and online framework
  - Reduces storage overhead, and thus cost
- **Pacemaker: regulating the HeART**
  - Manages redundancy management overheads
  - Perform cheap re-encoding
  - Converts urgent re-encoding tasks into schedulable tasks
- **HeART + Pacemaker reduces overall storage space by > 20% [m(b)illions \$]**

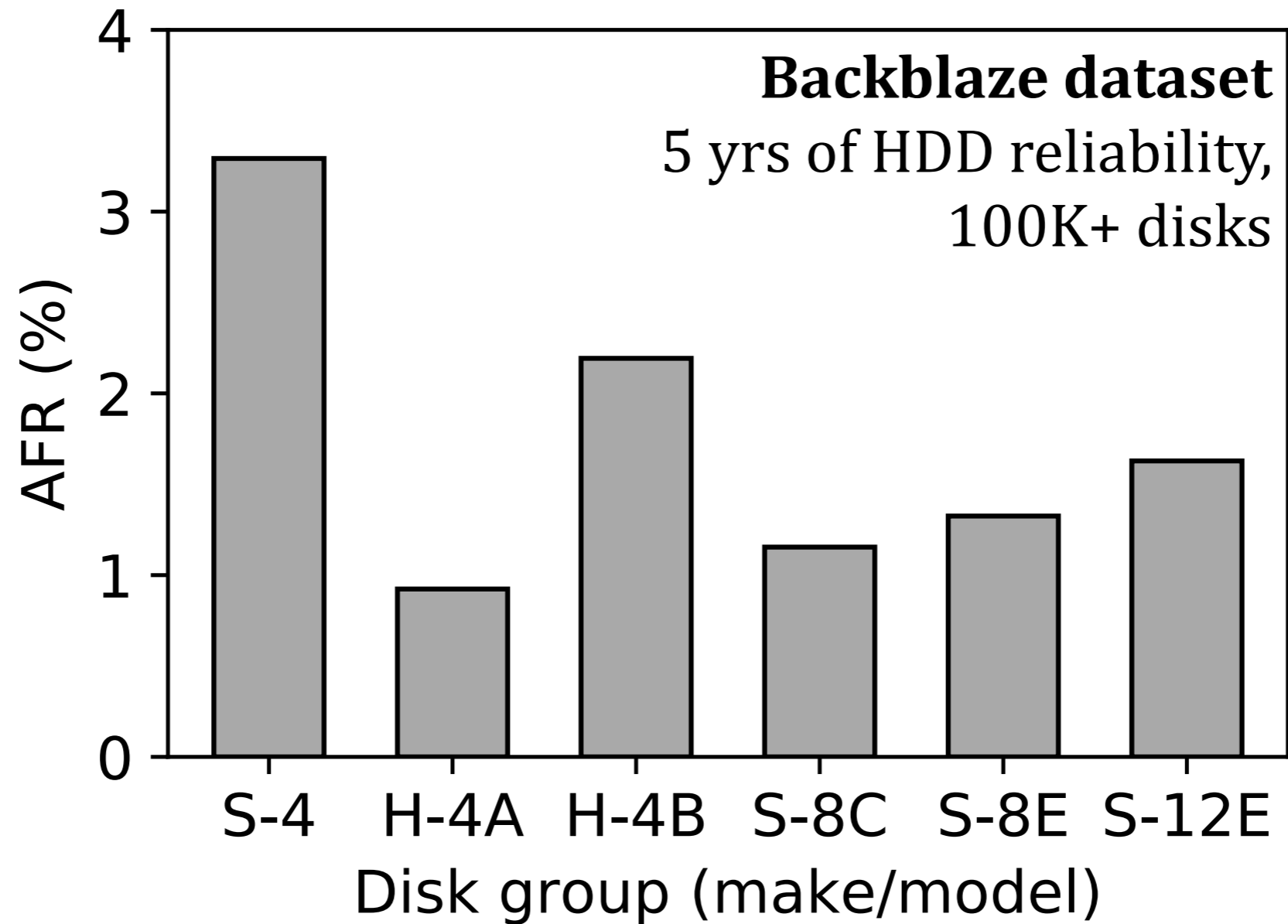
# Cluster storage system reliability

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- Failures common in today's cluster storage systems
  - Disk failures measured as **annualized failure rates (AFR)**
  - AFR  $\rightarrow$  expected % of disk failures in a year
- Popular fault tolerance mechanism  $\rightarrow$  redundancy
  - Full data replication (n-replication)
  - Erasure coding (**k-of-n**: k data chunks, n-k parity chunks)
- Reliability measured in mean-time-to-data-loss (MTTDL)
- Redundancy configurations ignore disk AFR differences

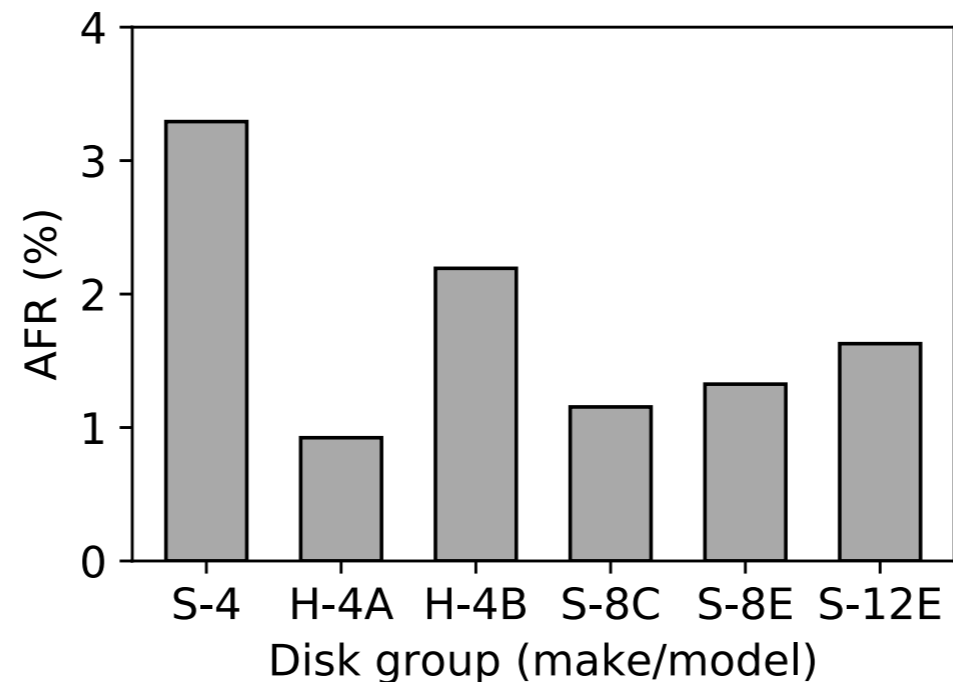
# Reliability heterogeneity

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# Reliability heterogeneity

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- HDD failure rates vary a lot in the field
- No single redundancy scheme is good enough for all disks
  - Conservative redundancy → overprotection for strong disk types
  - Lower redundancy → subset of disks risk data loss

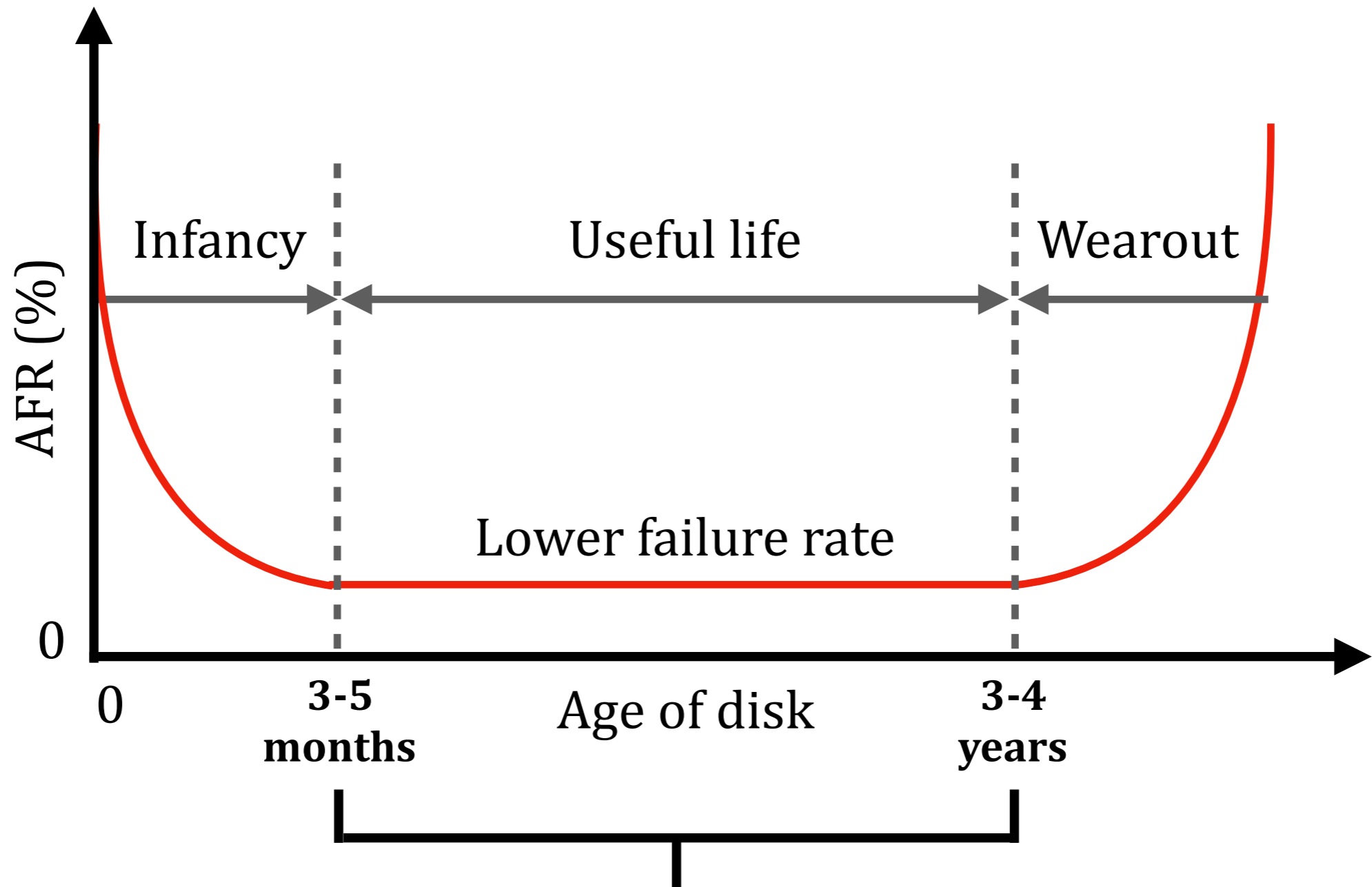
# Exploiting reliability heterogeneity

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- Redundancy decisions informed by AFR differences
- **Challenges**
  1. Has to be **monitored in the field**
  2. Disk failure rate **varies over its lifetime**
- Redundancy tailoring mechanism needs to be:
  - **Safe**: prevent under-redundancy from causing data loss
  - **Accurate**: identify different reliability phases correctly
  - **Online**: benefits only realizable during disk's low failure rate



# The bathtub curve (each disk group)

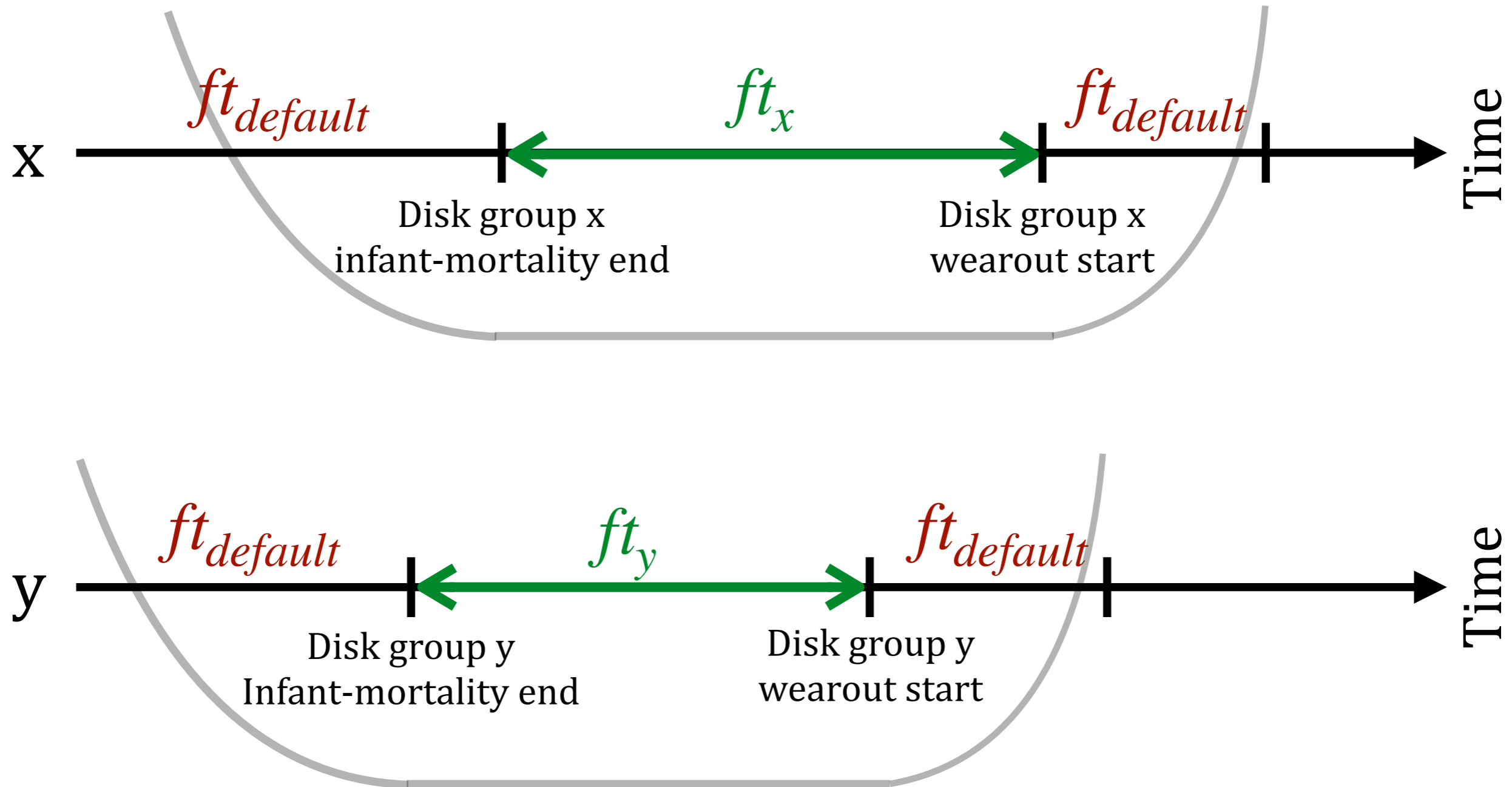


lower AFR → **lower redundancy** → **lower storage cost**

# Two disk groups over time

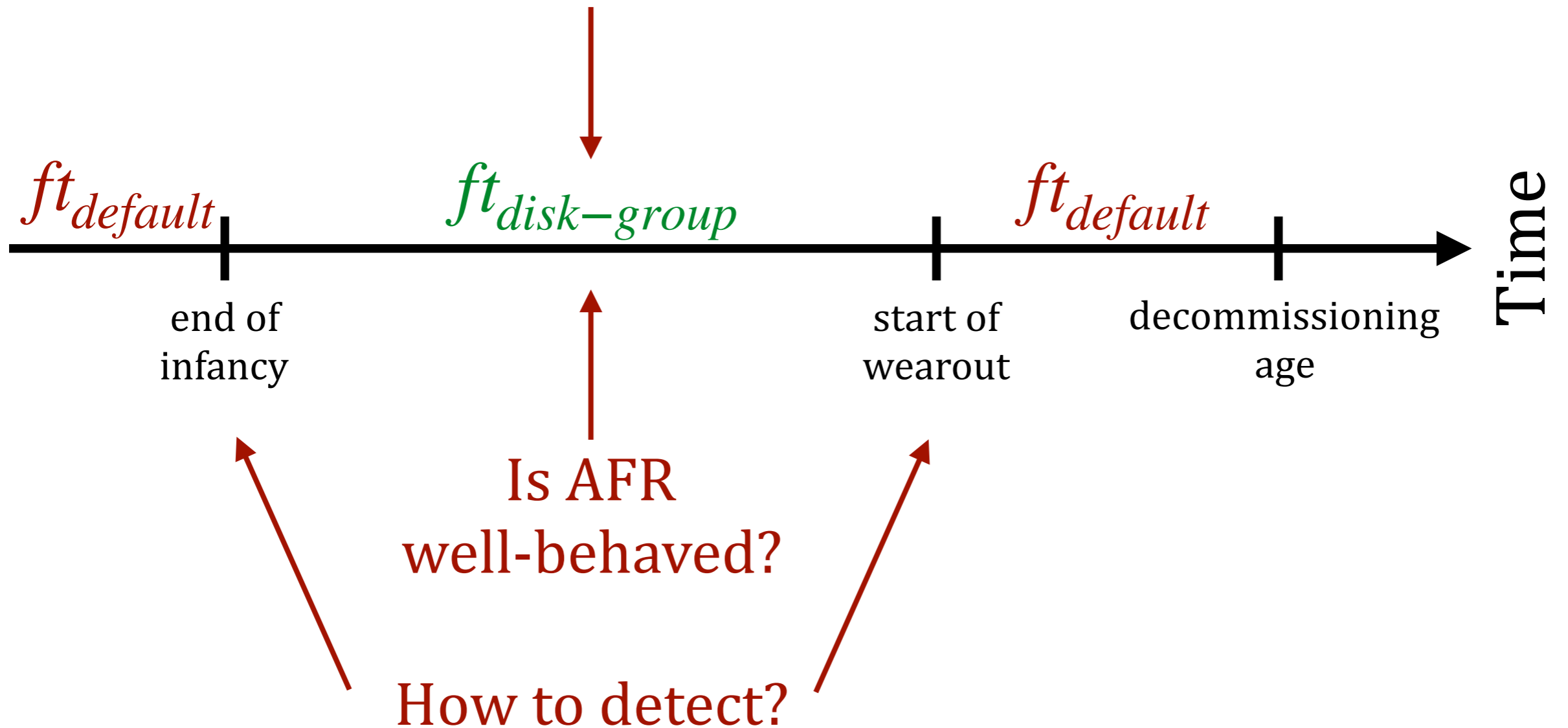
Deployment  
(start monitoring)

$ft_{default}$  = default fault tolerance scheme

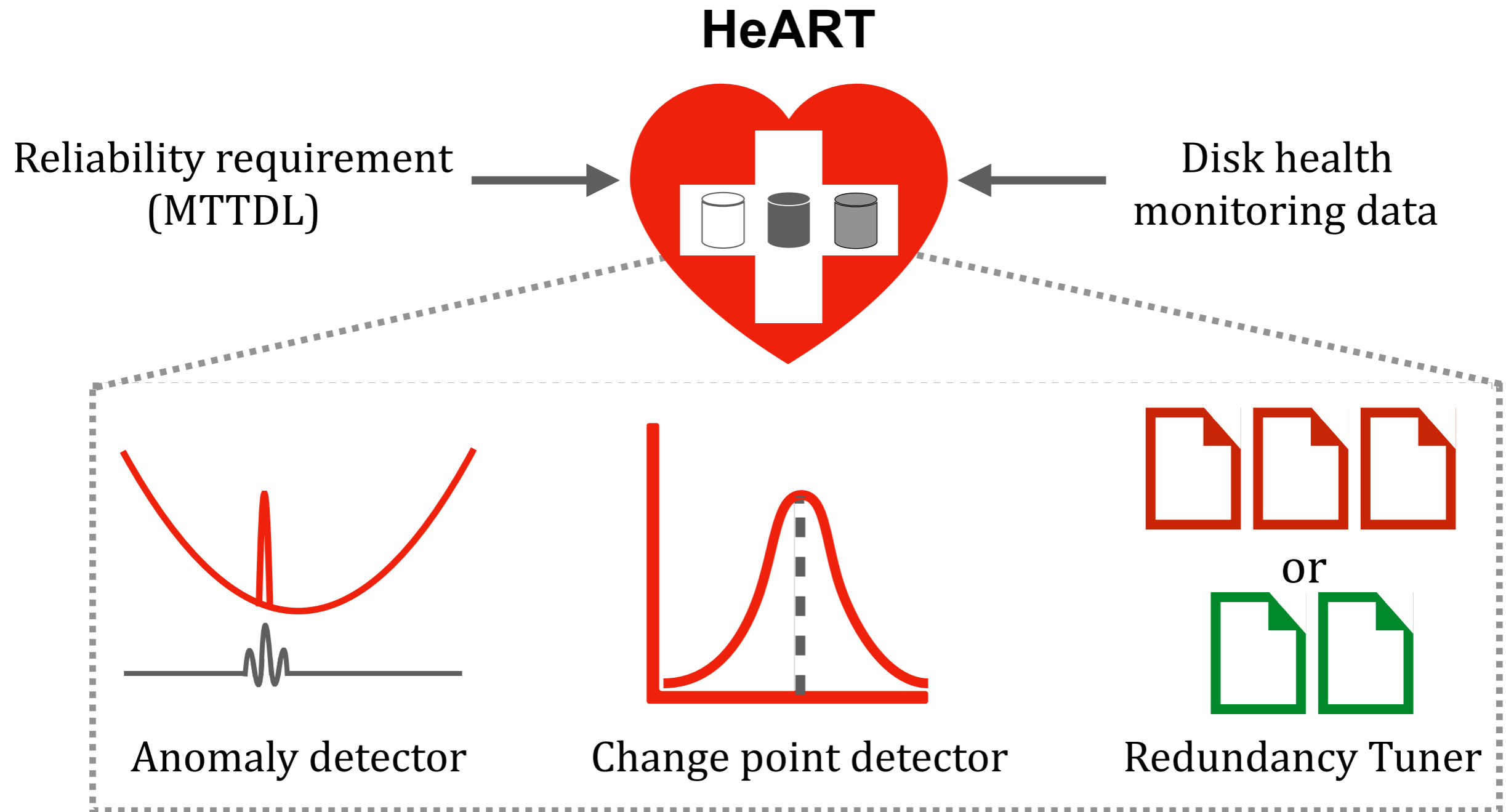


# Disk-group reliability timeline

What should the redundancy be?

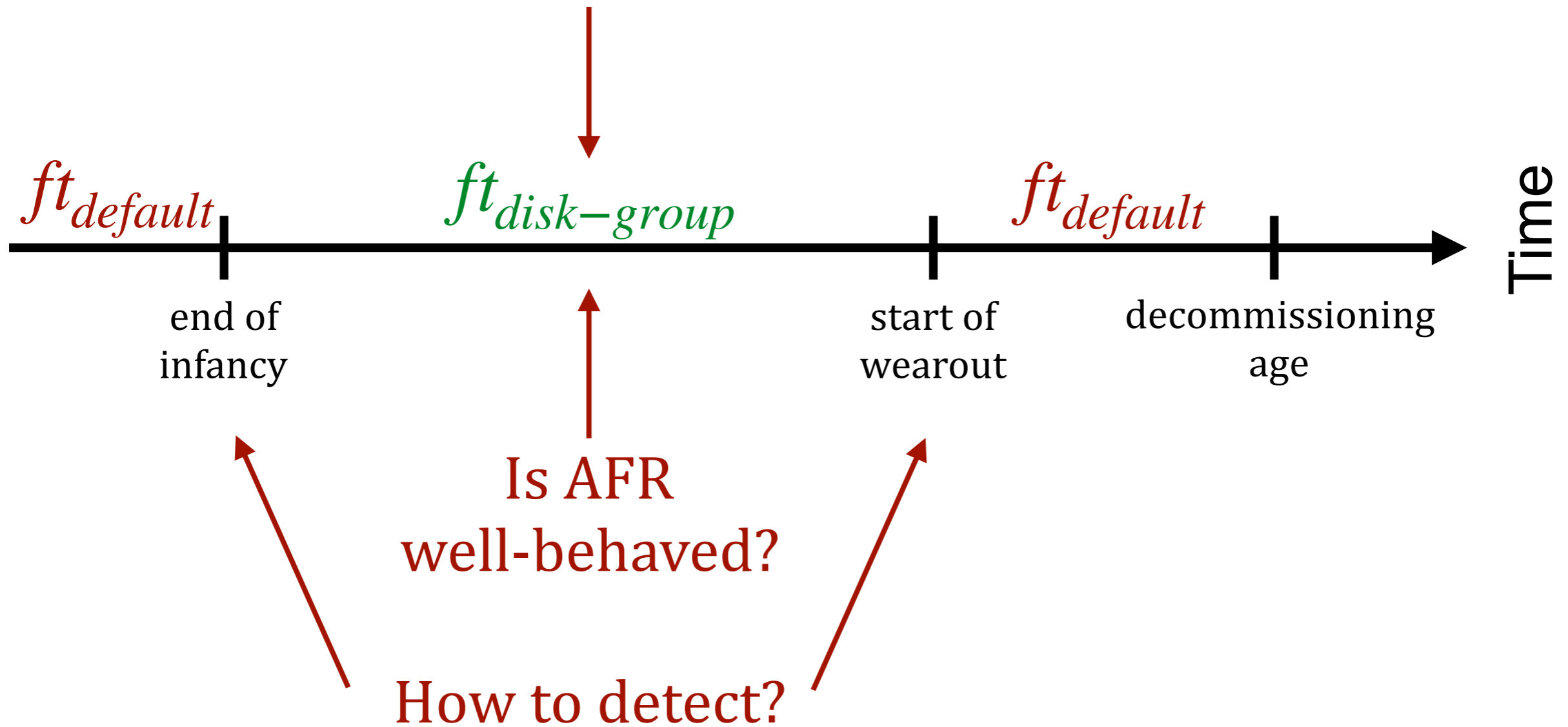


# Heterogeneity-Aware Redundancy Tuner

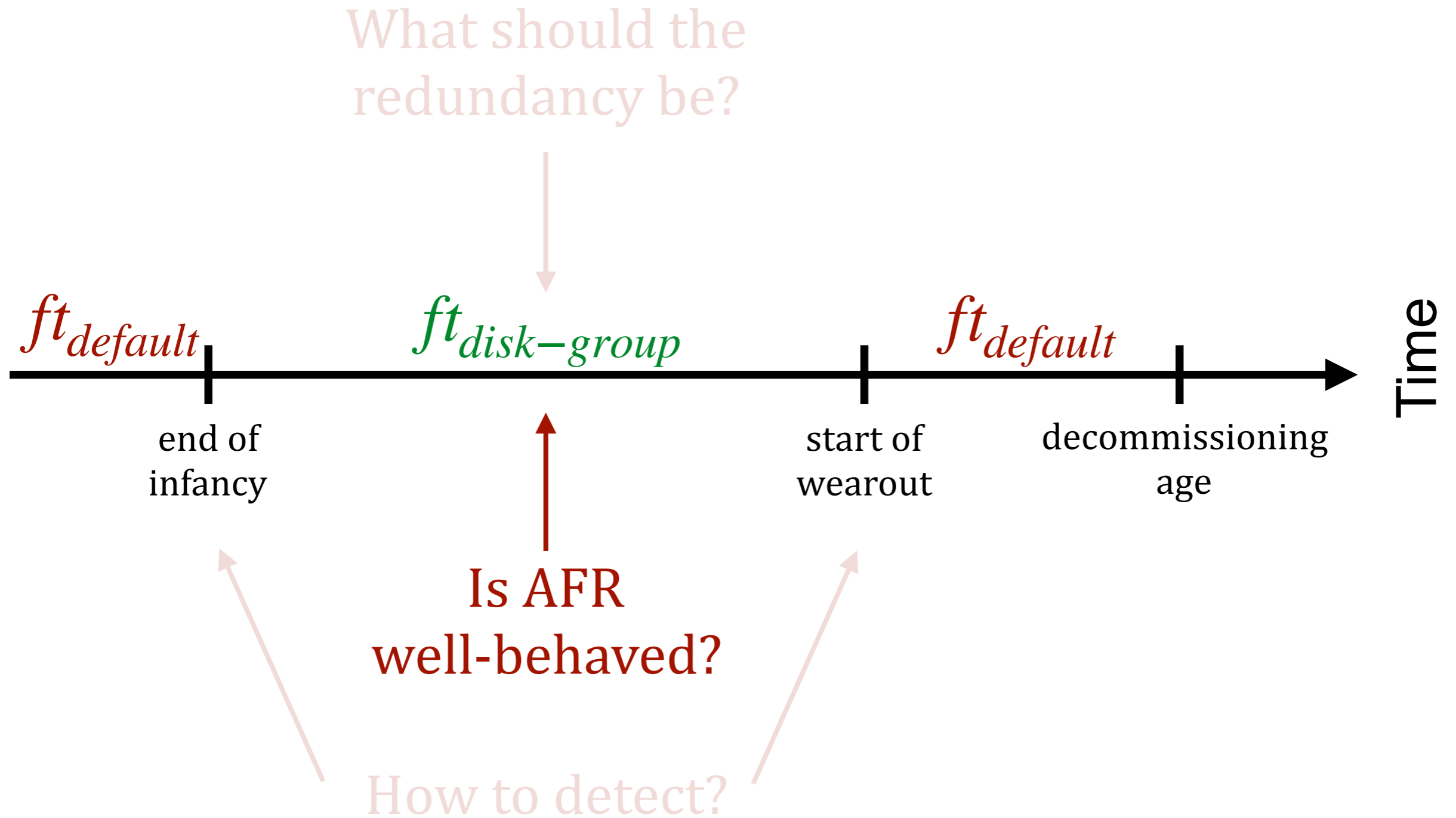


# Disk-group reliability timeline

What should the redundancy be?



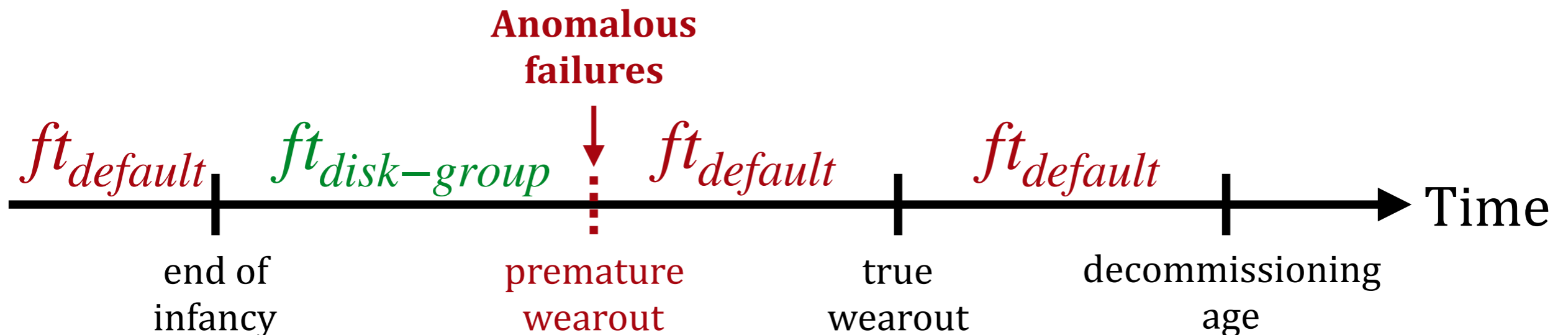
# Disk-group reliability timeline



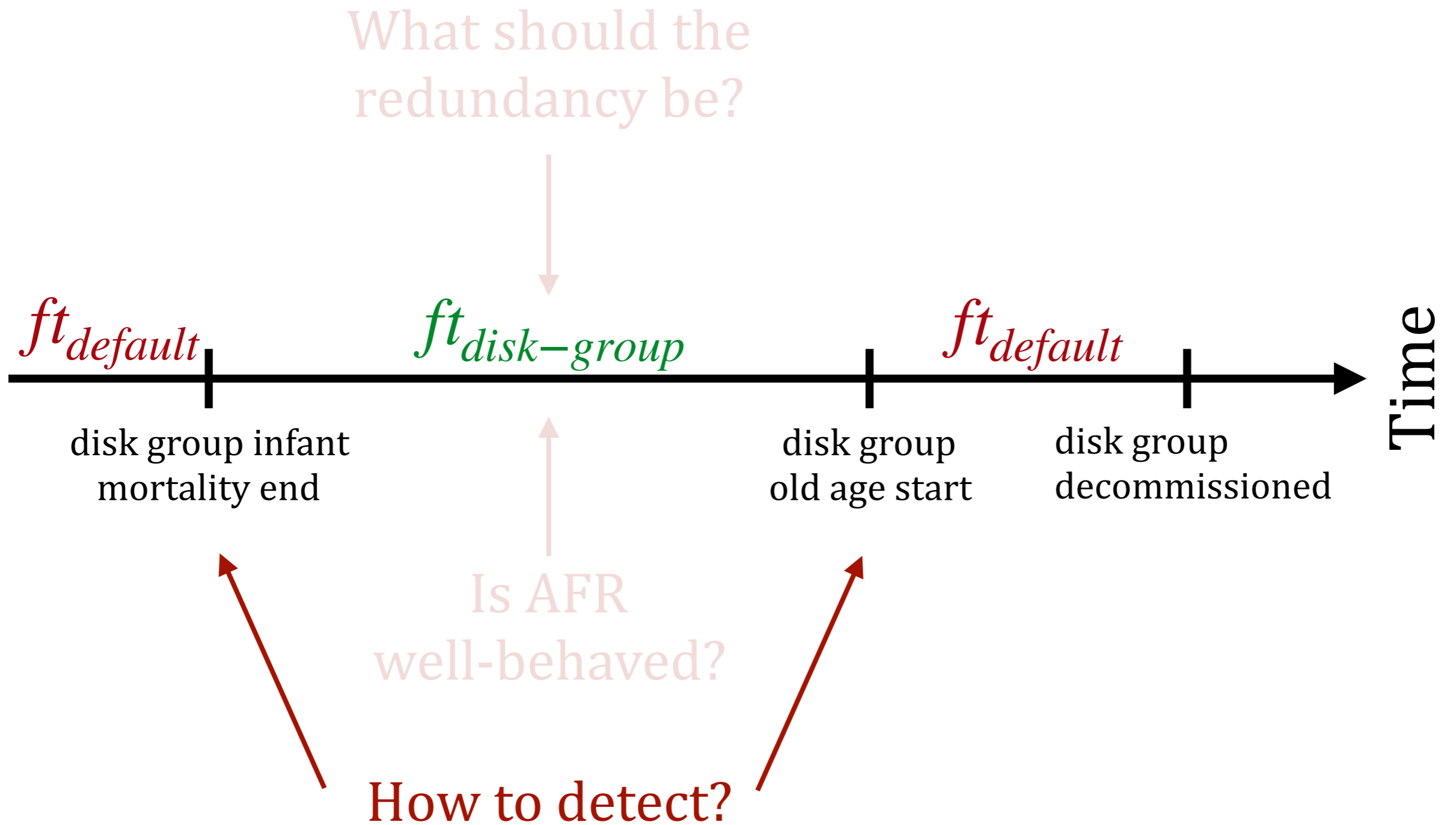
# AFR in useful life: stability & anomalies

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- Useful life AFR is typically stable, within reasonable bounds
- External factors can cause simultaneous bulk failures
  - Rack power failure, accidents, human error, etc.
- “Anomalies” appear like (premature) wearout
  - Benefits proportional to length of useful life
  - Bulk failures may not reflect true HDD failure rate



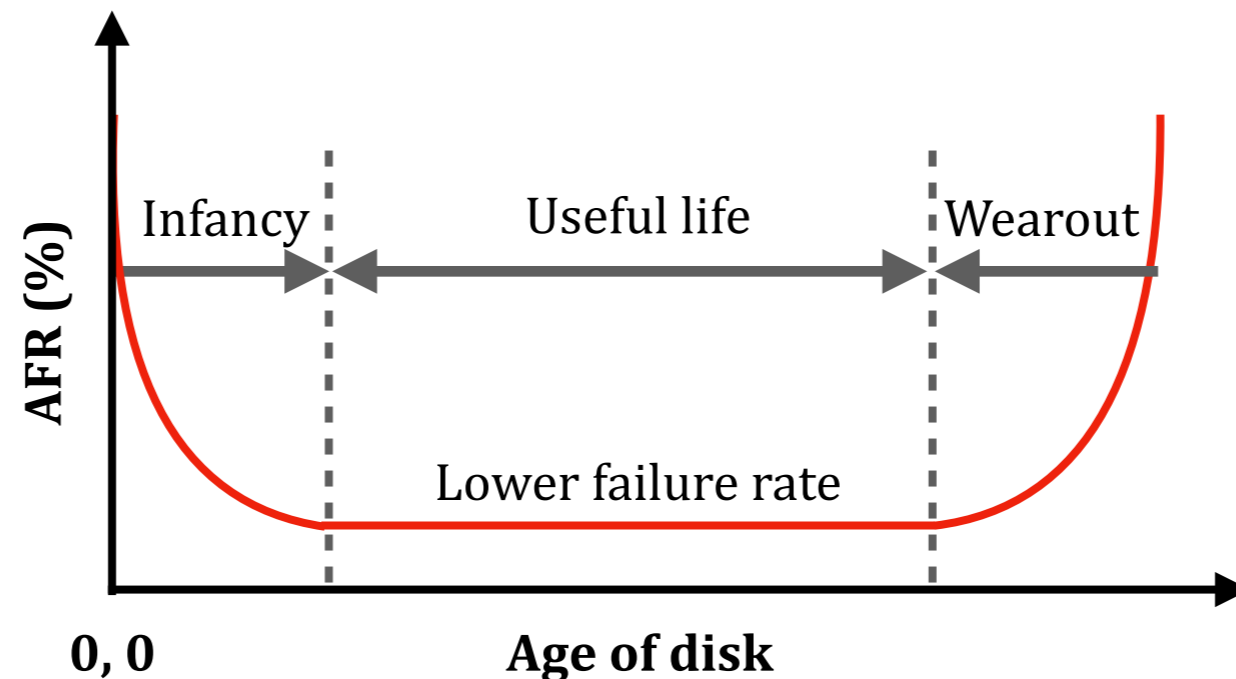
# Disk-group reliability timeline





# Change point detection

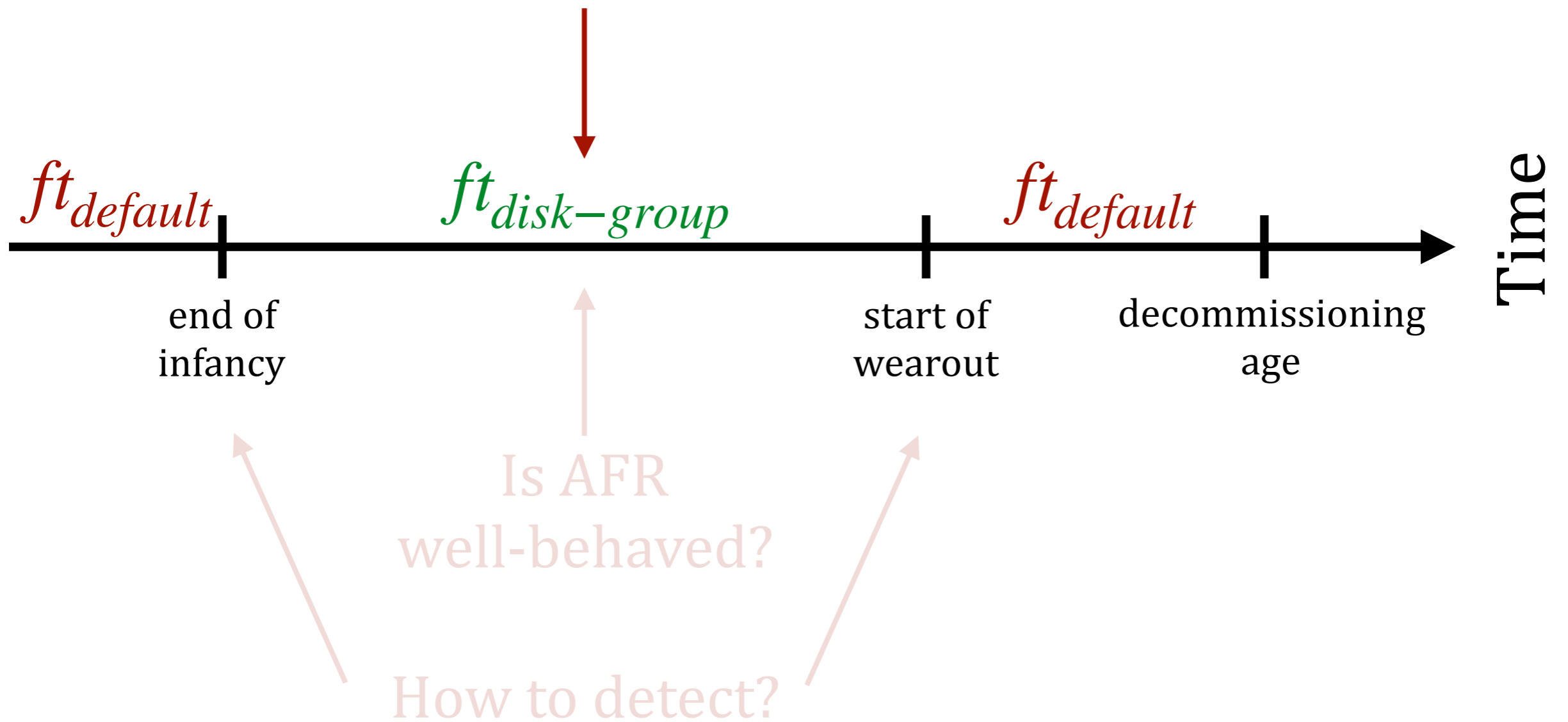
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- Reliability target can be missed if:
  - Hasty declaration of end of infancy
  - Delayed declaration of onset of wearout
- Tradeoff between extracting benefits and safety
- Use online change point detectors to identify change points

# Disk-group reliability timeline

What should the redundancy be?



# Redundancy scheme selection

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- $ft_{disk-group}$  MTTDL  $\geq$   $ft_{default}$  MTTDL (default AFR = 16%)
  - MTTDL: mean time to irrecoverable data loss

# Redundancy scheme selection

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- $f_{disk-group}$  MTTDL  $\geq f_{t}$  (Target reliability constraint = 16%)
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# Redundancy scheme selection

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- $ft_{disk-group}$  MTTDL  $\geq ft$  (Target reliability constraint = 16%)
  - MTTDL is the time to unrecoverable data loss
- Failures tolerated in  $ft_{disk-group} \geq$  failures tolerated in  $ft_{default}$

# Redundancy scheme selection

---

- $ft_{disk-group}$  MTTDL  $\geq ft_{target}$  (Target reliability constraint = 16%)
  - MTTDL is the time to unrecoverable data loss
- Failure rate  $\leq ft_{default}$  (Min num failures constraint)

# Redundancy scheme selection

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- $ft_{disk-group}$  MTTDL  $\geq ft_{target}$  (Target reliability constraint = 16%)
  - MTTDL is the mean time to unrecoverable data loss
- Failure rate  $\leq ft_{default}$  (Min num failures constraint)
- $ft_{disk-group}$  dimension  $\leq$  max dimension (max k = 30)

# Redundancy scheme selection

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- $ft_{disk-group}$  MTTDL  $\geq ft_{target}$  (Target reliability constraint = 16%)
  - MTTDL is the time to unrecoverable data loss
- Failure rate  $\leq ft_{default}$  (Min num failures constraint)
- $ft_{disk-group}$   $\leq ft_{default}$  (Max code width constraint (max K = 30))



# Redundancy scheme selection

- $ft_{disk-group} \times MTTDL \geq ft_{target}$  (Target reliability constraint = 16%)
  - MTTDL: Mean Time To Data Loss (recovery time to unrecoverable data loss)
- Failure rate  $\times ft_{disk-group} \geq$  failures tolerated in  $ft_{default}$  (Min num failures constraint)
- $ft_{disk-group} \times k \geq ft_{default}$  (Max code width constraint, max K = 30)
- Default AFR  $\times ft_{default} \geq$  Useful life AFR  $\times ft_{disk-group}$ 
  - Reconstruction IO:  $k \times \text{disk-capacity} \times \text{AFR}$

# Redundancy scheme selection

- $ft_{disk-group}$  MTTDL  $\geq ft_{target}$  (Target reliability constraint = 16%)
  - MTTDL  $\geq \frac{ft_{target}}{K}$  (where K = number of recoverable data loss)
- Failure rate  $\leq \frac{ft_{default}}{ft_{target}}$  (Min num failures constraint tolerated in  $ft_{default}$ )
  - $ft_{disk-group} \leq \frac{ft_{default}}{K}$  (where K = number of failures tolerated)
- $ft_{disk-group}$   $\leq K \times AFR$  (Max code width constraint, max K = 30)
  - $ft_{disk-group} \leq K \times AFR$
- Default AFR  $\times ft_{target} \leq H \times C$  (Max reconstruction work constraint)
  - Rec  $\leq K \times disk-capacity \times AFR$

# Redundancy scheme selection

- $ft_{disk-group}$  MTTDL  $\geq ft_{target}$  (Target reliability constraint = 16%)

  - MTTDL  $\geq \frac{ft_{target}}{K}$  (where  $K$  is the number of recoverable data loss)
- Failure rate  $\leq \frac{ft_{default}}{K}$  (Min num failures constraint)
- $ft_{disk-group}$   $\leq K \times ft_{disk}$  (Max code width constraint, max  $K = 30$ )
- Default AFR  $\times ft_{disk} \times K \leq \frac{ft_{target}}{disk-capacity}$  (Max reconstruction work constraint)

  - Rec  $\leq K \times disk-capacity \times AFR$
- $ft_{disk-group}$  reconstr. time  $\leq$  max reconstr. time (1.5 hrs)

# Redundancy scheme selection

- $ft_{disk-group}$  MTTDL  $\geq ft_{target}$  (Target reliability constraint = 16%)

  - MTTDL  $\geq \frac{ft_{target}}{K}$  (where  $K$  is the number of recoverable data loss)
- Failure rate  $\leq \frac{ft_{default}}{K}$  (Min num failures constraint tolerated in  $ft_{default}$ )
- $ft_{disk-group}$   $\leq K \times ft_{disk}$  (Max code width constraint, max  $K = 30$ )
- Default AFR  $\times ft_{disk} \times K \times disk-capacity \leq H_{max}$  (Max reconstruction work constraint)

  - Rec  $\leq \frac{H_{max}}{K \times disk-capacity \times AFR}$
- $ft_{disk-group}$   $\leq \frac{H_{max}}{K \times disk-capacity \times AFR}$  (Max reconstruction time constraint, 1.5 hrs)

# HeART is possible, but is it feasible?

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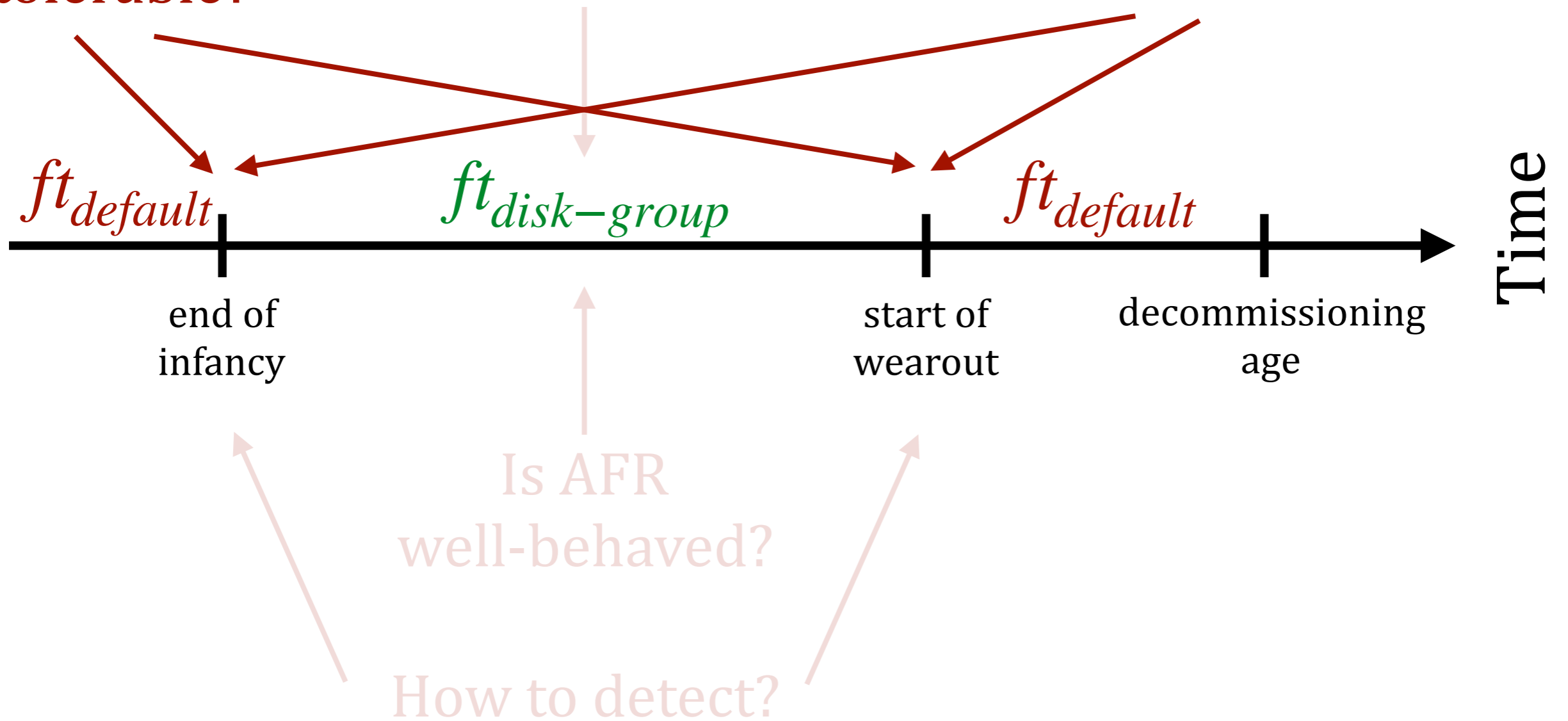
- Data gets re-encoded twice for each disk
  - Infancy —> useful life
  - Useful life —> wearout
- Read—re-encode—write cycle can be very expensive
  - Re-encoding 1TB disk from 30-of-33 to 6-of-9 is at least 75TB IO
- Re-encoding IO can hurt because of two main reasons:
  - Wide redundancy schemes used
  - Too many disks requiring re-encoding at the same time

# Disk-group reliability timeline

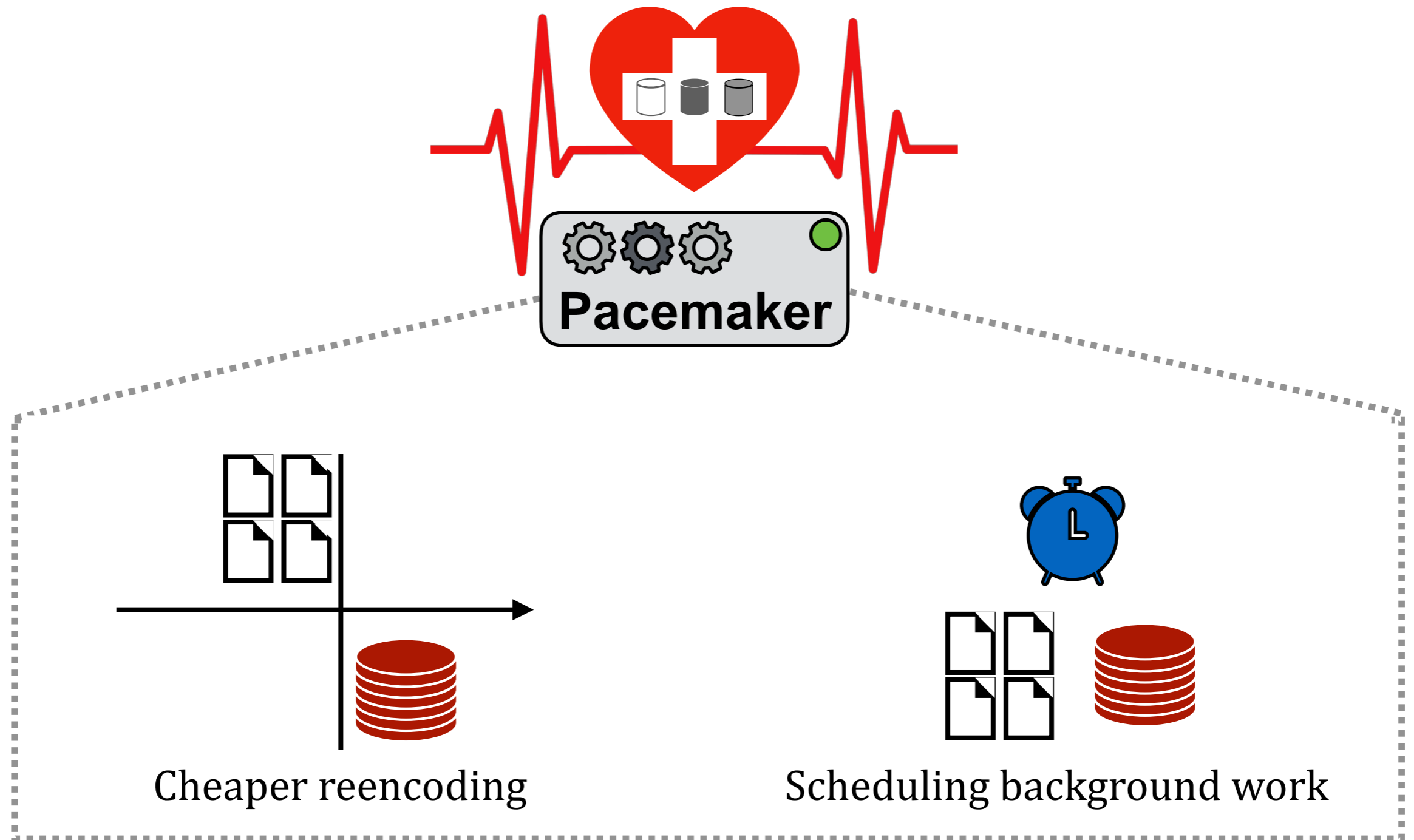
Are re-encoding overheads tolerable?

What should the redundancy be?

Is all re-encoding possible together?



# Pacemaker: regulating the HeART

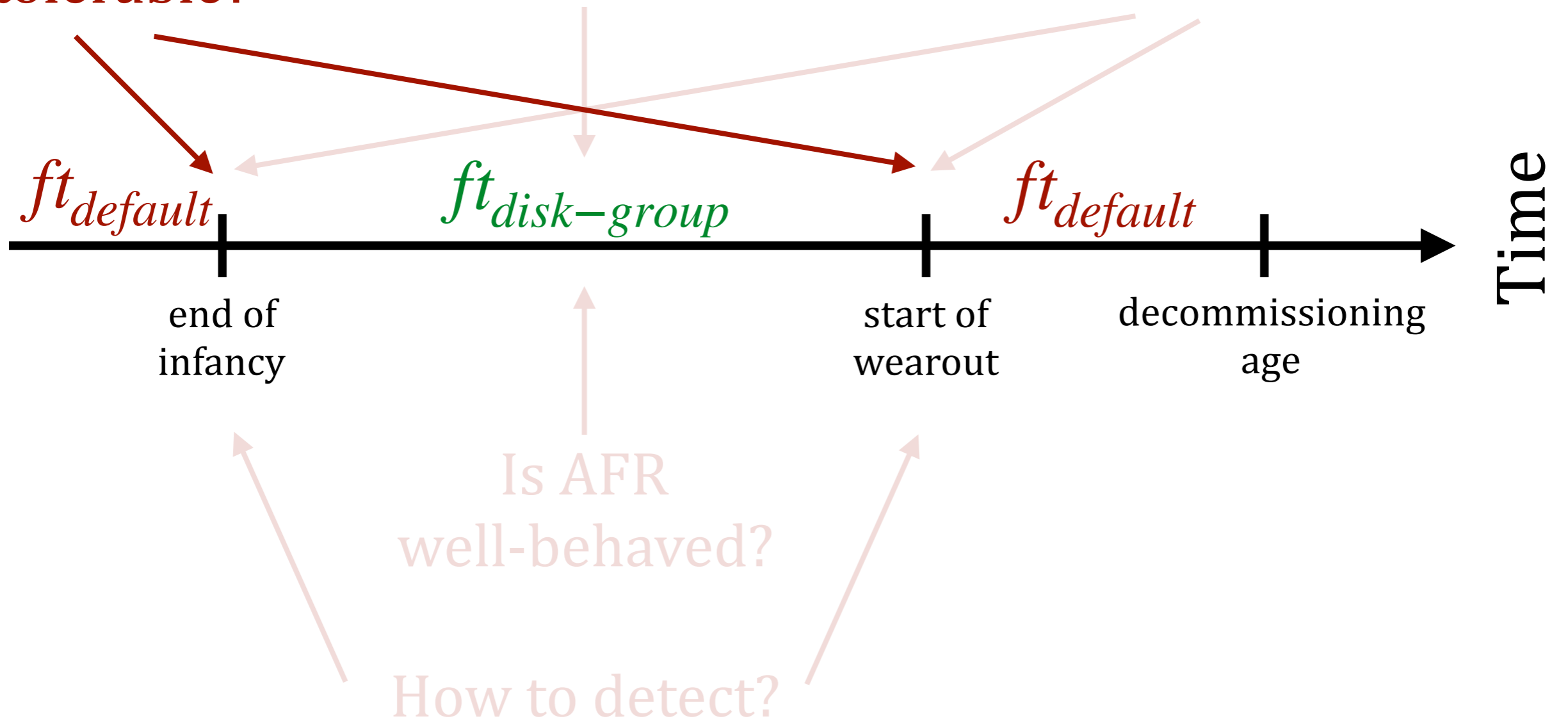


# Disk-group reliability timeline

Are re-encoding overheads tolerable?

What should the redundancy be?

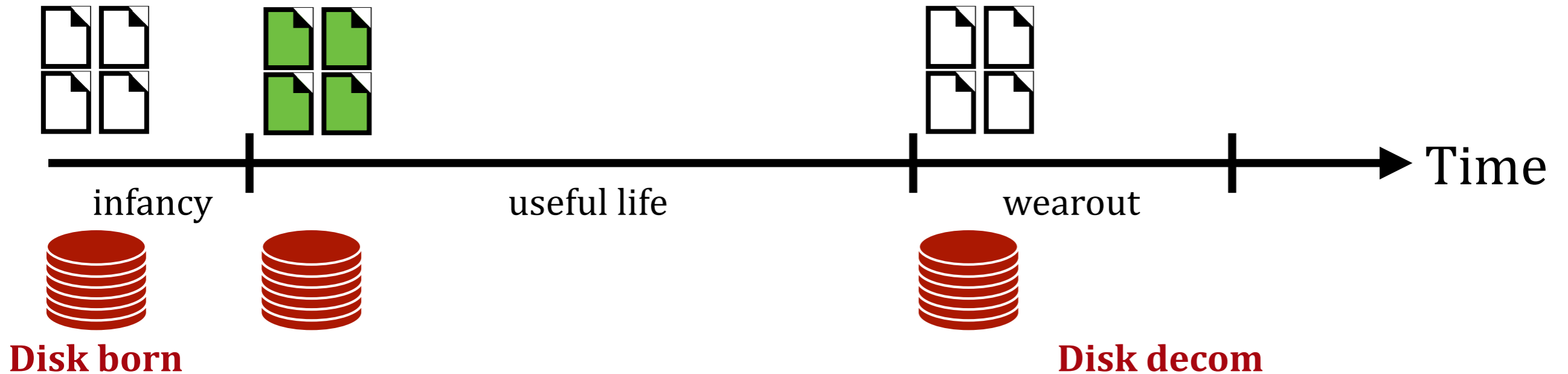
Is the re-encoding possible together?





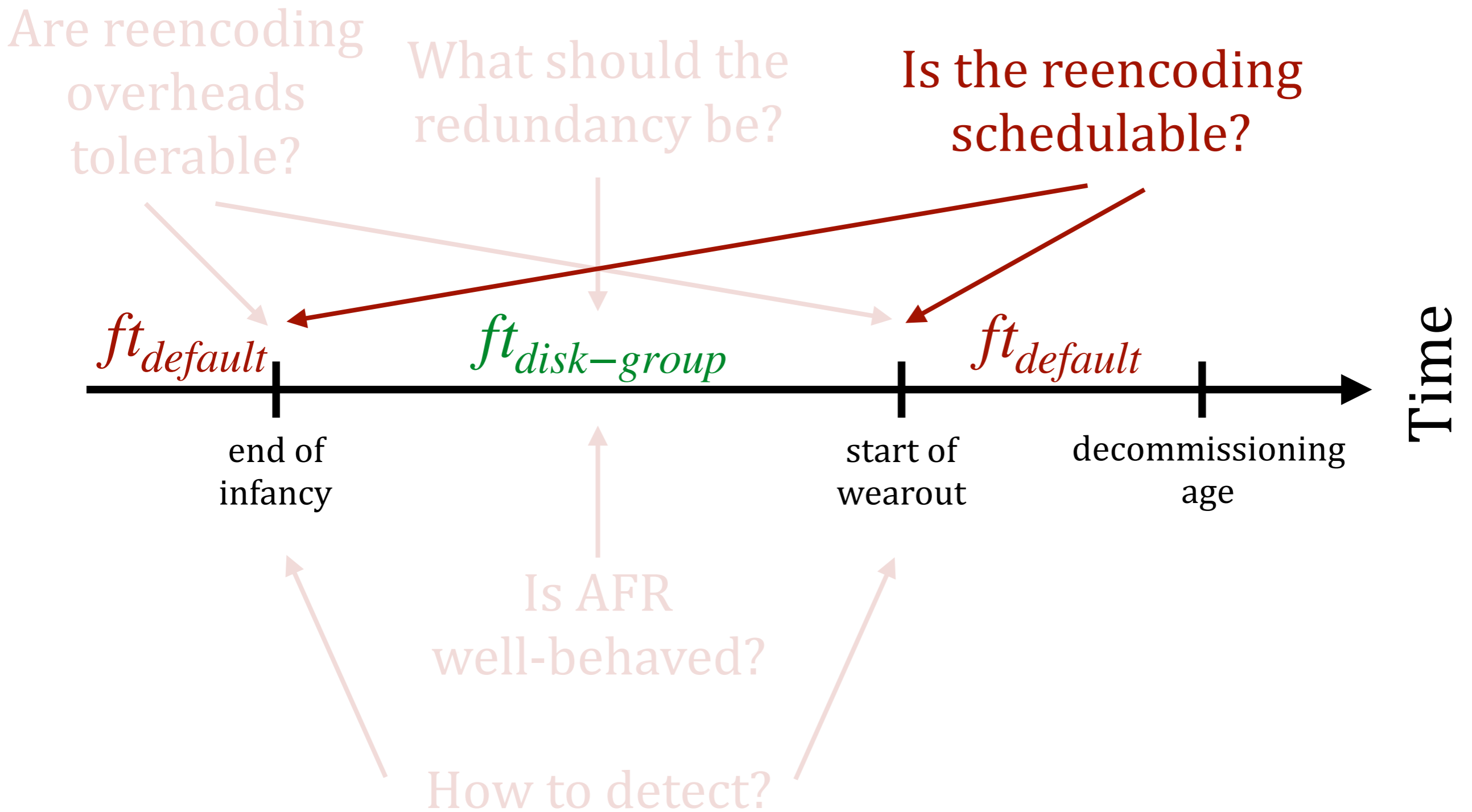
# Data reencoding = data redistribution

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- Recall that naive read—re-encode—write is very expensive
  - $k \times$  disk capacity needs to be read and written
- Key idea: **disks change failure families, it's data need not**
- Moving one stripe unit cheaper than reencoding entire stripe
  - Decouples reencoding IO from redundancy scheme used
  - Moving eliminates the computation overhead, only generates I/O

# Disk-group reliability timeline



# Schedulable background work

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- Infancy to useful life transition is completely schedulable
  - Only impacts savings because of reduced useful life
- Useful life to wearout is urgent
  - But not all of it...
- Key observation: **not all disks enter wearout together**
- Incremental disk deployments help schedule urgent work
  - **Only the first disk batch used to detect wearout is urgent**
  - Subsequent disks wearout transitions can be scheduled

# Other optimizations

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- **Canary disks**
  - Canaries can be encoded in conservative redundancy schemes
  - 2000 for detecting end of infancy & 1000 for detecting wearout
- **Useful life AFR buffer**
  - Buffer helps protect against jitter in AFR during useful life
  - Buffer also helps in exercising caution when tuning redundancy
- **Deciding wearout based on what  $ft_{disk-group}$  can tolerate**
  - Useful life redundancy scheme chosen on basis of detected AFR
  - Transition to wearout based on what the scheme can tolerate
- **Iterative change point detection**
  - One-shot change point detection too conservative
  - More data => lower useful life AFR => greater savings

# The Backblaze dataset

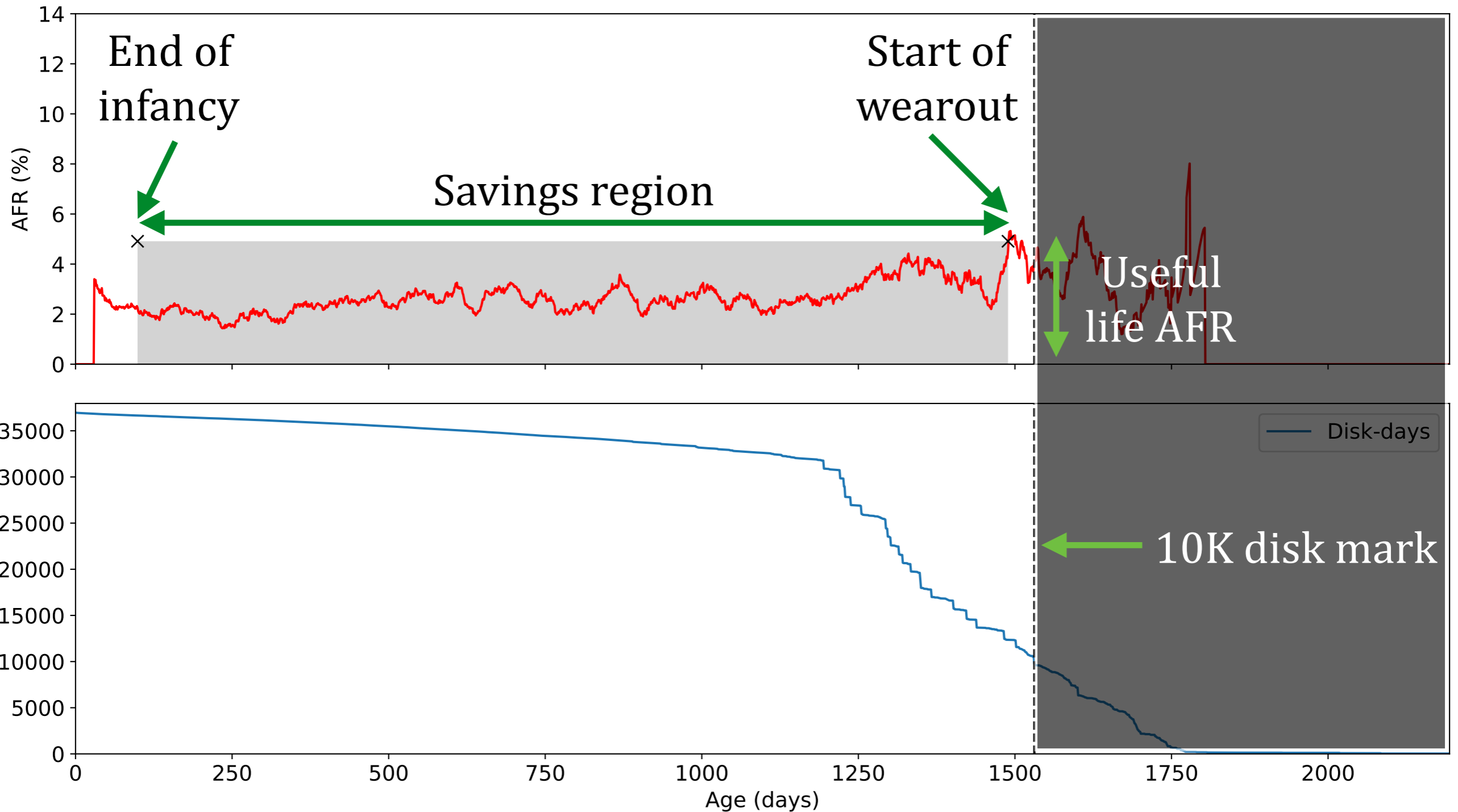
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- 100K+ HDDs belonging to Backblaze: a backup company
  - Daily reliability statistics from mid 2013 - mid 2019
  - Open sourced
  - 7 drive makes/models with significant number of disks to test:

Disk Grp	Num Drives	Num Failed	Age so far (yrs)
S-4	36962	3535	6
H-4A	8708	137	6
H-4B	16316	207	5
S-8C	10150	275	3
S-8E	14716	331	2.5
S-12E	35435	735	1.5
H-12E	9680	10	0.5

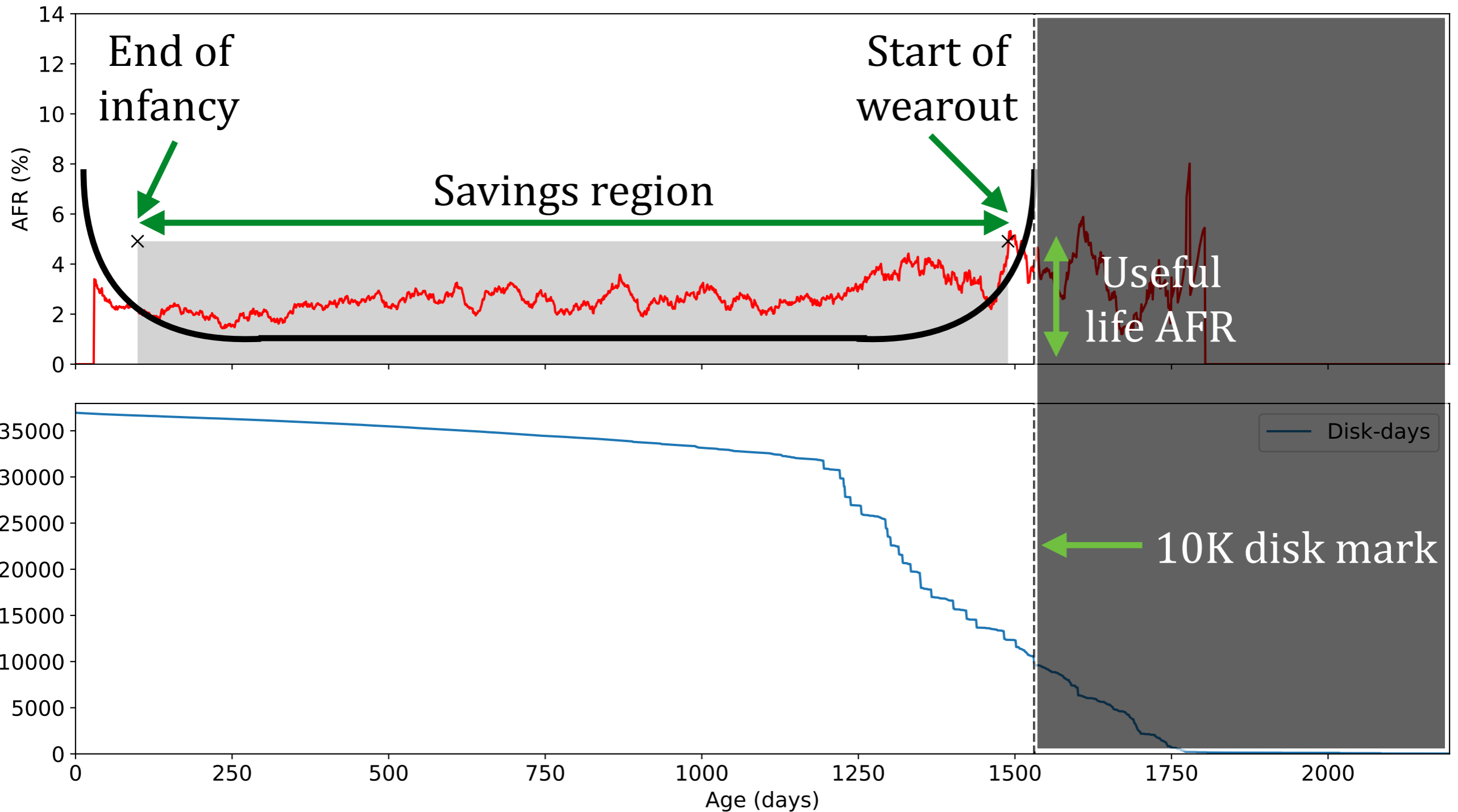
# HeART in action on a disk-group

## S-4 AFR details



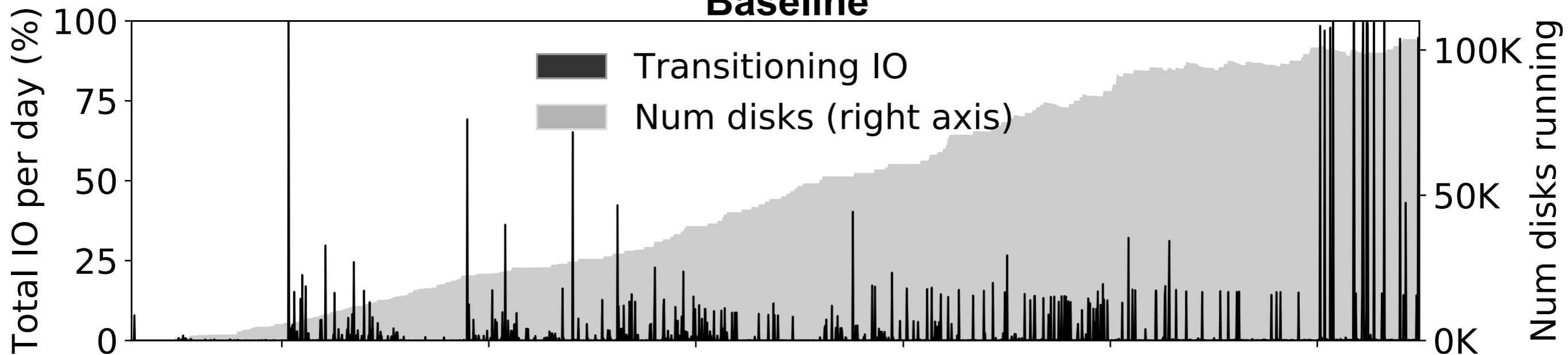
# HeART in action on a disk-group

## S-4 AFR details

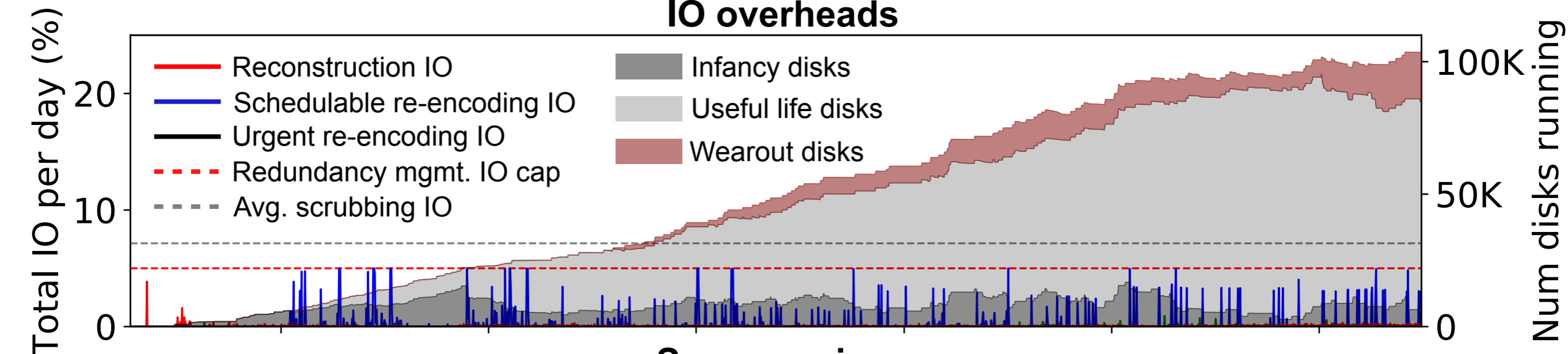


# HeART + Pacemaker on Backblaze

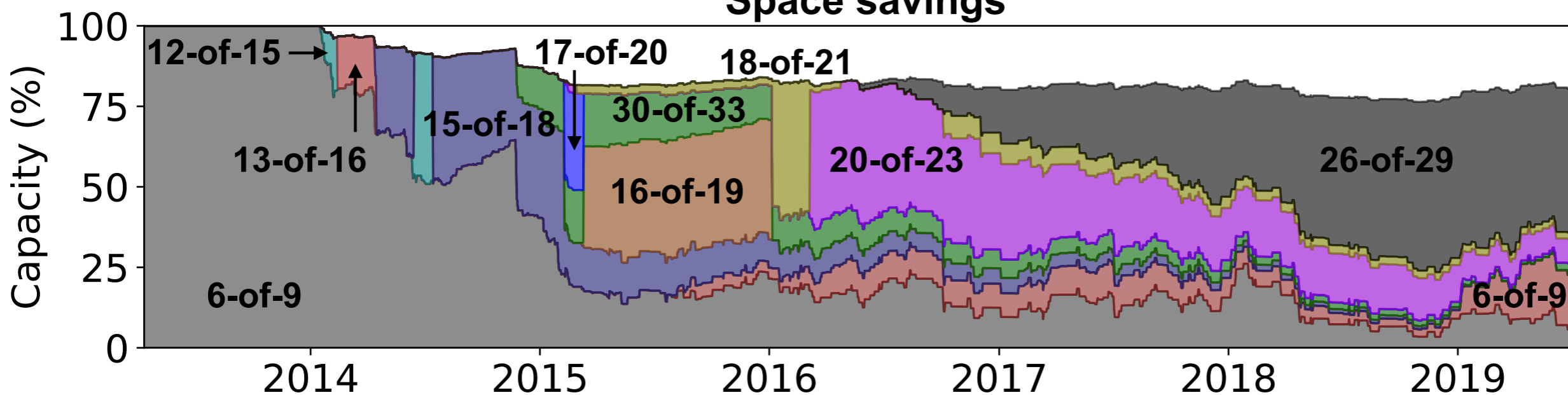
## Baseline



## IO overheads



## Space savings





# HeART summary

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- Exploiting reliability heterogeneity reduces storage cost
- Overall >20% space savings observed on production dataset
- Less than 5% IO bandwidth spent in redundancy mgmt
- **HeART**: an online heterogeneity-aware redundancy tuner
  - actively engages with disk bathtub curves
  - built-in online anomaly and change point detector
- **Pacemaker**: performs efficient redundancy management
  - data redistribution instead of data reencoding
  - converts urgent redundancy mgmt IO into schedulable IO

# References

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2. Sathiamoorthy, Maheswaran, et al. "**Xoring elephants: Novel erasure codes for big data.**" *VLDB*. 2013.
3. Guha, Sudipto, et al. "**Robust random cut forest based anomaly detection on streams.**" *ICML*. 2016.
4. Truong, Charles, Laurent Oudre, and Nicolas Vayatis. "**ruptures: change point detection in Python.**" *arXiv:1801.00826*. 2018.
5. Rashmi, K. V., et al. "**A hitchhiker's guide to fast and efficient data reconstruction in erasure-coded data centers.**" *ACM SIGCOMM Computer Communication Review*. 2015.



“My heart is in the work”

“My work is in the HeART”

