



LCI Advanced Workshop 2025: Fair Resource Allocation

Alan Chapman
HPC Systems Analyst
Arizona State University Research Computing
alan.chapman@asu.edu

This document is a result of work by volunteer LCI instructors and is licensed under CC BY-NC 4.0 (https://creativecommons.org/licenses/by-nc/4.0/



AGENDA



- Fairshare Fundamentals & Why It Matter
- Classic Fairshare Algorithm Deep-Dive
- Fairtree Algorithm Architecture
- Comparative Analysis & Trade-offs
- Configuration & Tuning Best Practice
- Q&A + Lab Preview



WHY FAIRSHARE MATTERS



The Problem: Shared Resources

Why It Matters

- Multi-tenant clusters: Fair allocation across departments/project
- Prevents hoarding: Users can't monopolize resources long-term
- Incentivizes efficiency: Heavy users get lower priority (temporary)
- Academic/Industry: Standard practice for HPC governance



FAIRSHARE FUNDAMENTALS



Concept	Definition	l
		l
Fairshare Factor	User's fair allocation ÷ actual consumption (0.0-1.0)	l
Decay Half-Life	Time for past usage to lose 50% weight (e.g., 7 days)	l
Account Hierarchy	Tree structure: root → dept → project → user	l
Priority Formula	Job priority = (fairshare × wieght) + (age × weight) + (QOS × weight)	i
Usage Tracking	Slurm tracks CPU-minutes, memory, GPU-minutes per account	l

Example Calculation

Account: lci/ml

Fair Share: 50% (assigned)

Usage: 10,000 CPU-min over 7 days

Fairshare Factor = (Fair Share - Usage) / Fair Share

= (50% - 45%) / 50% = 0.1 (LOW priority)

Account: lci/io Fair Share: 50%

Usage: 5,000 CPU-min

Fairshare Factor = (50% - 22.5%) / 50% = 0.55 (HIGHER priority)







```
The Algorithm (Slurm ≤ 22.04)
FOR each pending job:
fairshare_factor = calculate_factor(job.account)
priority = (fairshare_factor × weight_fairshare)
+ (age_factor × weight_age)
+ (qos_factor × weight_qos)
queue_job_by_priority(job, priority)
FUNCTION calculate factor(account):
# Recursive: parent's factor affects children
parent factor = calculate factor(parent account)
my_usage = sum(usage_since_decay_time)
my_share = my_fairshare_pct * parent_factor
factor = (my_share - my_usage) / my_share
return max(0, factor) # Clamp to [0, 1]
```





CLASSIC FAIRSHARE - REAL EXAMPLE

Scenario: 2-Level Hierarchy

root (100% fair share)

lci (50% share, used 30% of cluster)

ml (30% of lci's share)

io (20% of lci's share)

admin (50% share, used 10% of cluster)

Job Priority Calculation

Job A: account=lci/ml Parent factor (lci) = (50% - 30%) / 50% = 0.4 My usage (ml) = 5% of cluster My fair share = $30\% \times 0.4 = 12\%$ My factor = $(12\% - 5\%) / 12\% = 0.58 \rightarrow MEDIUM$ priority

Job B: account=lci/io Parent factor (lci) = $0.4 \leftarrow$ SAME as ml! My usage (io) = 8% of cluster My fair share = $20\% \times 0.4 = 8\%$ My factor = $(8\% - 8\%) / 8\% = 0.0 \rightarrow$ LOW priority

Job C: account=admin Parent factor (root) = (100% - 10%) / 100% = 0.9 My usage (admin) = 10% of cluster My fair share = $50\% \times 0.9 = 45\%$ My factor = $(45\% - 10\%) / 45\% = 0.78 \rightarrow HIGH$ priority



FAIRTREE ALGORITHM (INTRO)



What Changed in Slurm 22.05+

Goal: Eliminate cascade penalty; compute fairshare locally per node

Classic: Fairtree:

Key Innovation: Sibling Fairness

In Fairtree, each account's factor depends ONLY on:

- 1. Its own usage
- 2. Its own fair share
- 3. Its siblings' usage (not ancestors)

Algorithm Sketch

```
FUNCTION fairtree_factor(account):
my_usage = sum(usage_since_decay_time)
my_share = account.fairshare_pct
# Key: compute against siblings, not parent
sibling_total_share =
sum(sibling.fairshare_pct)
sibling_total_usage = sum(sibling.usage)
# Normalize my share within sibling group
my_normalized_share = my_share /
sibling_total_share
factor = (my_normalized_share - my_usage) /
my_normalized_share
return max(0, factor)
```



FAIRTREE - REAL EXAMPLE



Same Scenario, Fairtree Algorithm

```
root (100% fair share)

├─ lci (50% share, used 30% of cluster)

├─ ml (30% of lci's share)

└─ io (20% of lci's share)

└─ admin (50% share, used 10% of cluster)
```

Job Priority Calculation (Fairtree)

Job A: account=lci/ml

Sibling group: {ml, io}

Sibling total share: 30% + 20% = 50%

My normalized share: 30% / 50% = 60%

My usage: 5% of cluster

My factor = $(60\% - 5\%) / 60\% = 0.92 \rightarrow HIGH$

priority ✓

Job B: account=lci/io
Sibling group: {ml, io}
Sibling total share: 50%
My normalized share: 20% / 50% = 40%
My usage: 8% of cluster
My factor = (40% - 8%) / 40% = 0.80 →
MEDIUM priority ✓

Job C: account=admin
Sibling group: {lci, admin}
Sibling total share: 50% + 50% = 100%
My normalized share: 50% / 100% = 50%
My usage: 10% of cluster
My factor = (50% - 10%) / 50% = 0.80 →
MEDIUM priority ✓



COMPARATIVE ANALYSIS - COMPLEXITY



Algorithm Complexity

Metric	Classic	Fairtree
Time per job	 O(depth × n)	O(children)
Scheduling cycle (1000 jobs, 200 accounts)	~500ms	~50ms
Scalability	Poor (deep trees)	Excellent
Memory overhead	Low	Medium (sibling caches)
Decay recalc	Global (expensive)	Local (cheap)

Behavior Differences

Scenario	Classic 	Fairtree
•	•	Each child judged fairly Direct
New account (zero usage)	High priority	High priority
Starvation risk	YES (cascade)	NO (local max)
Predictability	Low	High



IMPLEMENTATION - SLURM CONFIG



Enable Fairtree (Slurm 22.05+)

```
# In slurm.conf
PriorityType=priority/multifactor
                                  # Fair Tree
PriorityFlags=FAIR_TREE
#PriorityFlags=DEPTH OBLIVIOUS
                               # Classic Fairshare
# Tune weights (example for CPU cluster)
PriorityWeightFairshare=100000000 # Fairshare dominates priority
PriorityWeightAge=10000000 # Age: jobs wait, priority increases
PriorityWeightQOS=110000000 # QOS: can override fairshare
# Decay half-life (past usage fades over time)
PriorityDecayHalfLife=7-0 # 7 days; usage older than this loses weight
# Usage reset (optional; default: NONE)
PriorityUsageResetPeriod=NONE # Don't reset; continuous decay
# Backfill (allows smaller jobs to fill gaps)
SchedulerParameters=bf window=43200,bf resolution=300
```



ACCOUNT HIERARCHY DESIGN



```
Good: Hierarchical (Recommended)
root (100% fair share)
  – admin (5% share)

— sysadmin (100% of admin's share)

 – research (70% share)
   - ml_team (40% of research)
    — alice (50% of ml_team)
    bob (50% of ml_team)
 bio_team (30% of research)
  — charlie (60% of bio_team)
  └─ diana (40% of bio team)
teaching (25% share)
class 2024 (100% of teaching)
```

PRACTICAL IMPACTS - ACADEMIC WORKLOAD



Scenario: Multi-PI Department

Department: 16 cores, 64 GB RAM (shared)

```
Account tree:

root

— pi_alice (50% fair share)

| — phd_student_1 (50% of alice's)

| — phd_student_2 (50% of alice's)

— pi_bob (50% fair share)

— phd_student_3 (50% of bob's)

— phd_student_4 (50% of bob's)

— postdoc_1 (0%) ← Temporary, gets scraps
```



PRACTICAL IMPACTS - ACADEMIC WORKLOAD



Workload Pattern

Time Job Submitter Cores Status Classic Factor Fairtree Factor 09:00 phd_student_1 4 RUNNING 1.0 1.0 09:30 phd_student_3 4 RUNNING 1.0 1.0 10:00 phd_student_2 4 PENDING 0.5 0.9 ← Fairtree higher! (reason: alice over-allocated in classic) 10:15 postdoc_1 2 PENDING 0.0 0.7 ← Fairtree higher! (postdoc has no fair share, but Fairtree still gives chance)



PRACTICAL IMPACTS - ACADEMIC WORKLOAD



Wait Time Impact:

Classic:

phd_student_2: waits 45 min (cascade penalty from alice over-use)
postdoc_1: waits indefinitely (zero fair share)

Fairtree:

phd_student_2: waits 10 min (fair within ml_team)
postdoc_1: waits 20 min (gets some priority, prevents starvation)



PRACTICAL IMPACTS - CPU WORKLOADS



Scenario: Batch Job Farm

```
Cluster: 128 cores, 2 accounts

root (100%)

├── lci (60% share)

│ ├── batch_inference (40% of lci)

│ └── data_processing (60% of lci)

└── admin (40% share)
```

Job Submission Pattern

```
09:00 batch_inference submits 50 × 4-core jobs (200 cores needed)
Only 4 cores available; 49 jobs queued

09:05 data_processing submits 10 × 2-core jobs (20 cores)
All queued behind batch_inference

09:10 admin submits 1 × 8-core job
All queued

09:15 First batch_inference job finishes (4 cores freed)
```



PRACTICAL IMPACTS - CPU WORKLOADS



Scheduling Decision

Classic Fairshare:

- Ici parent factor: (60% 4%) / 60% = 0.93
- batch_inference factor: 0.85 (over-used relative to share)
- data_processing factor: 0.95 (under-used)
- → Schedules: data_processing (0.95 > 0.85)
- → Batch job waits longer X

Fairtree:

- batch_inference vs. data_processing (siblings)
- batch inference used 4%, share 40% → factor 0.90
- data_processing used 0%, share 60% → factor 1.0
- → Schedules: data_processing (1.0 > 0.90)
- → Same decision, but for right reason



TUNING FAIRSHARE - WEIGHTS



Priority Formula (Multifactor)

```
job_priority = (fairshare_factor × weight_fs)
+ (age_factor × weight_age)
+ (qos_factor × weight_qos)
+ (job_size_factor × weight_size)
```

Weight Scenarios

Scenario 1: Fairshare-Dominant (Research Cluster)

PriorityWeightFairshare=100000000 PriorityWeightAge=10000000 PriorityWeightQOS=1000000

Effect: Historical usage is primary; age and QOS minor.

Use case: Multi-PI research; want fair allocation of resources.



TUNING FAIRSHARE - WEIGHTS



Scenario 2: Age-Dominant (Responsiveness)

PriorityWeightFairshare=1000000

PriorityWeightAge=100000000

PriorityWeightQOS=1000000

Effect: Jobs waiting longer get boosted; fairshare secondary.

Use case: Interactive clusters; want quick turnaround for all users.

Scenario 3: QOS-Dominant (SLAs)

PriorityWeightFairshare=1000000

PriorityWeightAge=1000000

PriorityWeightQOS=100000000

Effect: QOS classes override fairshare.

Use case: Production systems; high-priority jobs must run.







Decay Half-Life Impact

PriorityDecayHalfLife=7-0 # 7 days

Timeline Example

Day 0: User runs 1000 CPU-min job

Recorded usage: 1000 CPU-min

Day 7: (half-life reached)

Effective usage: 500 CPU-min (50% decay)

Day 14: (2× half-life)

Effective usage: 250 CPU-min (75% decay)

Day 21: (3× half-life)

Effective usage: 125 CPU-min (87.5% decay)



TUNING FAIRSHARE - DECAY



Policy Implications

Lab Comparison

Config A: PriorityDecayHalfLife=1-0 Config B: PriorityDecayHalfLife=7-0 Config C: PriorityDecayHalfLife=NONE

Submit workload, measure priority evolution over time.



BEST PRACTICES



✓ Do's

- 1. Use Fairtree (Slurm 22.05+)
- Migrate from classic; performance + fairness gains
- 2. Design hierarchy to match org structure
- root → department → project → user
- Typically 3-4 levels; avoid >5 (diminishing returns)
- 3. Allocate fair shares proportionally to allocation
- If ml_team gets 40% of budget, give 40% fair share
- Not 50% (too generous) or 30% (too stingy)
- 4. Set decay half-life to your cluster culture
- Research: 7 days (balanced)
- Production: 14-30 days (strict)
- Dev: 1 day (forgiving)
- 5. Use PriorityWeightFairshare >> other weights
- Fairshare should dominate (100M vs. 10M for age)
- 6. Monitor fairshare factors regularly
- `sshare -l` or `sacctmgr show assoc`
- Alert if any account hits factor 0.0 (starvation risk)

X Don'ts

- 1. Don't use classic fairshare (pre-22.05)
- Upgrade; Fairtree is proven, faster
- 2. Don't create >100 accounts per level
- Scheduling overhead; merge into fewer buckets
- 3. Don't set fair share >100% at any level
- Violates fairness semantics; set to 100% or less
- 4. Don't use NONE for decay half-life on interactive clusters
- Users get "locked out" after big job; use 1-7 days
- 5. Don't ignore fairshare in QOS
- QOS can override fairshare; use sparingly (SLAs only)
- 6. Don't change hierarchy without planning
- Moving accounts can cause starvation; test first



MIGRATION GUIDE



From Classic to Fairtree

Step 1: Backup current slurm.conf cp /etc/slurm/slurm.conf /etc/slurm/slurm.conf.bak

Step 2: Add FAIR_TREE flag sed -i 's/PriorityFlags=DEPTH_OBLIVIOUS/PriorityFlags=FAIR_TREE/' /etc/slurm/slurm.conf

Step 3: Reload Slurm config (no restart needed) scontrol reconfigure

Step 4: Verify scontrol show config | grep PriorityFlags

Step 5: Monitor for 1 week watch -n 60 'sshare -l | head -20'

Rollback if needed:

sed -i 's/PriorityFlags=FAIR_TREE/PriorityFlags=DEPTH_OBLIVIOUS/' /etc/slurm/slurm.conf scontrol reconfigure



MIGRATION GUIDE



Expected Changes in First Week

Before (Classic):

- Fairshare factors: erratic, hard to explain
- Priority jumps: large, unpredictable
- Wait times: variable for same-sized jobs

After (Fairtree):

- Fairshare factors: stable, intuitive
- Priority progression: smooth
- Wait times: more predictable
- Scheduling decisions: faster (50ms vs. 500ms per cycle)



SUMMARY



Key Takeaways

- 1. Fairshare is essential for multi-tenant clusters
 - Prevents starvation, incentivizes efficiency, enables governance
- 2. Fairtree (22.05+) beats classic
 - O(n) vs. O(n²), local fairness vs. cascading penalties, predictable
- 3. Config matters
 - Hierarchy design, fair share allocation, weights, decay half-life
- 4. Monitoring is critical
 - Watch fairshare factors; alert on starvation
- 5. Migration is easy
 - One flag; no restart; immediate benefits



QUESTIONS & RESOURCES



Questions?

Resources:

Fair Tree - https://slurm.schedmd.com/fair_tree.html
Classic Fair Share - https://slurm.schedmd.com/classic_fair_share.html
Multifactor Priority Plugin - https://slurm.schedmd.com/priority_multifactor.html

Lab Setup:

- Cluster: lciadv (2 compute nodes, 2 cores each)
- Accounts: root, lci
- Users: root, rocky (lci)
- Duration: 60 min
- All CPU workloads (no GPU)

