University of Glasgow

Evaluation of Memory and CPU usage via Cgroups of ATLAS workloads running at a Tier-2

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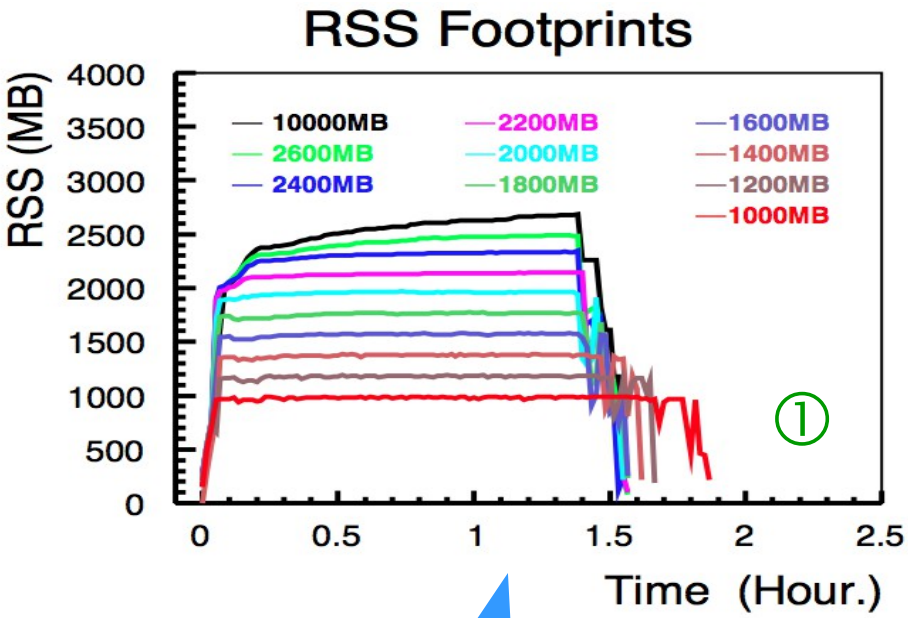
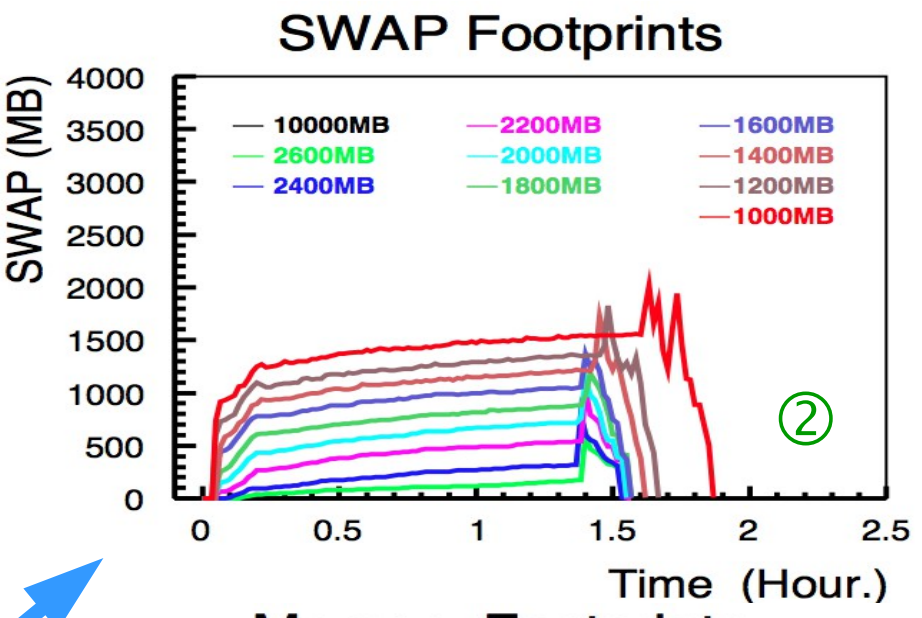
March. 25th , 2015

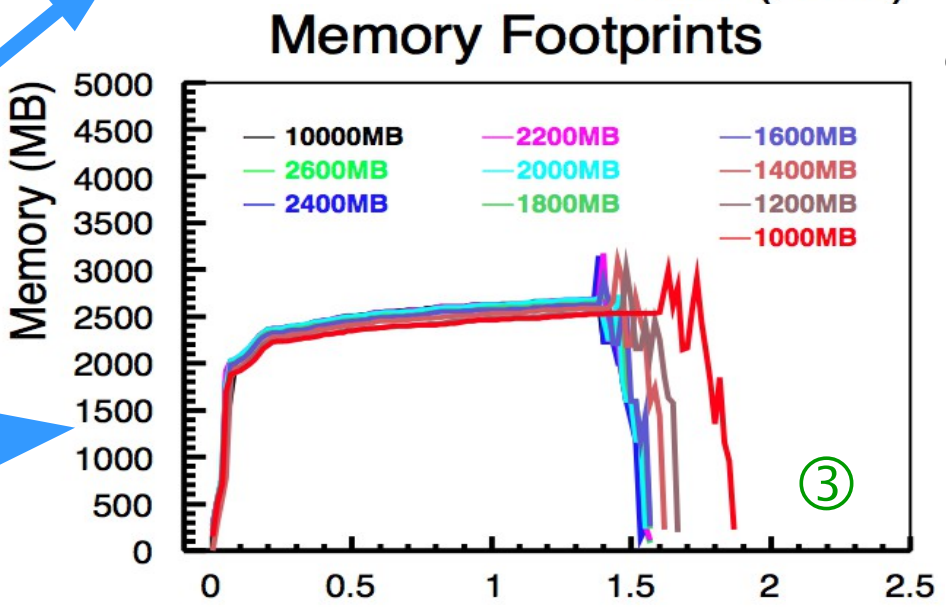
Condor Cgroups

- Control Groups (Cgroups)
  - Linux kernel feature to limit/isolate resource usage among user-defined groups of tasks(processes).
  - Available Resource Controllers (or subsystems):
    - Block-I/O,cpu/cpuacct/cpuset/devices/freezer/memory/net_cls/net_prio/ns
- Condor Cgroups
  - Condor puts each job into a dedicated cgroup for a selected subsystem
  - Can be used to track the subsystem usage of all processes started by a job
  - control cpu usage at job level:
    - Jobs can use more cpu than allocated if there are still free cpu
  - control RSS (physical memory) usage at job level:
    - **soft**: jobs can access more memory than allocated if there is still free physical memory available in the system
    - **hard**: jobs can't access more physical memory than allocated

Condor Cgroups







- Can be used to track the subsyste
- control cpu usage at job level:
  - Jobs can use more cpu than a
- control RSS (physical memory)
  - **soft**: jobs can access more m physical memory available i
  - **hard**: jobs can't access more

Glasgow Tier-2

- Glasgow Condor Cluster
  - Production instance since Aug 2014, has received ~ 1.3 million jobs, comprises 2 ARC-CE(8/16 core), 1 condor central server (8core), 42 worker-nodes (1456 logical cores)
- MySQL Databases:
  - Condor: select/record historical info of condor jobs, incuding ClusterId/GlobalJobId/JobStatus/ExitCode/LastJobStatus/RequestCpus/RequestMemory/JobMemoryLimit/JobTimeLimit/Use/RemoteWallClockTime/RemoteSysCpu and etc..
  - Panda: PandaID of finished/failed jobs in GLASGOW panda queues
- Memory/Cpu info collection from cgroups:
  - Cgmemd collects the following every minute for each job (on each WN):
    - Cputime: total CPU time consumed by all tasks in the job (later converted into the regular CPU usage by comparing 2 neighbouring sampling points):
    - RSS: instantaneous physical memory usage of the job
    - SWAP: instantaneous swap usage of the job
  - Memory = RSS + SWAP

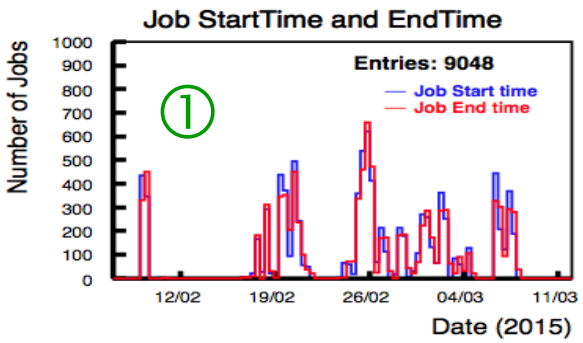
$$cpu\_usage_i = \frac{(cpu\_time_i - cpu\_time_{i-1})/10^9}{(Time_i - Time_{i-1})}$$

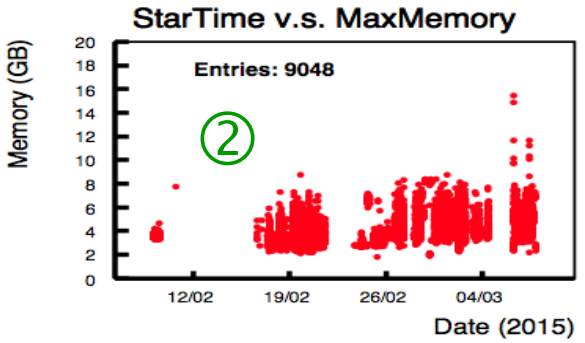
- ATLAS job Info tracking:
  - GlobalJobId, trfName(last one), PandaID (last one)
- Analysis
  - Currently focus on ATLAS good jobs

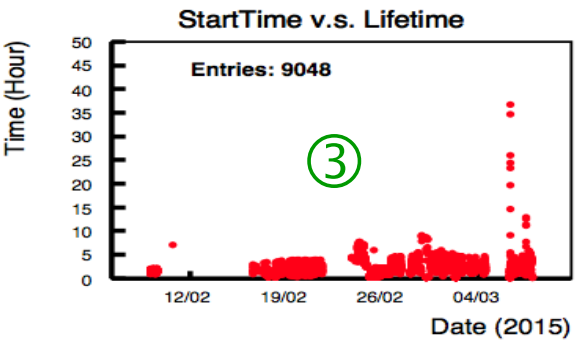
ATLAS jobs at Glasgow Tier-2 *(4)*

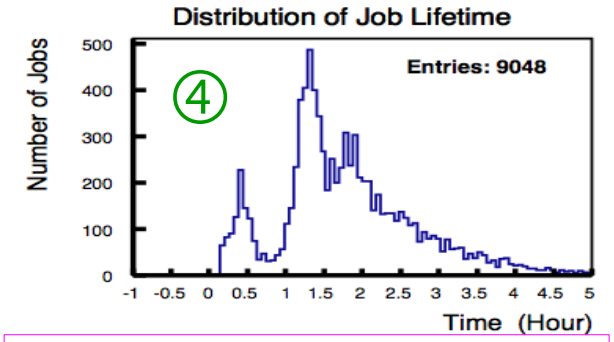
- Empty Pilots
  - runs 2 or 3 minutes, using ~25MB memory with CPU usage < 0.1
- Production jobs:
  - Multi-core:
    - panda_queue = UKI-SCOTGRID-GLASGOW_MCORE
    - Request_cpu = 8 & Req_memory = 16 GB
    - Site policy: RSS > 16GB not allowed, no restriction on SWAP
  - Single-core:
    - panda_queue = UKI-SCOTGRID-GLASGOW_SL6
    - Request_cpu = 1 & Request_memory = 3 GB
    - Site policy: RSS > 3GB not allowed, no restriction on SWAP
  - GlobalJobId – PandaID: one pilot picks <= 1 production job
- Analysis jobs:
  - panda_queue = ANALY_GLASGOW_SL6
  - Request_cpu = 1 & Request_memory = 4GB
  - Site restriction: RSS > 4 GB not allowed, no restriction on SWAP
  - GlobalJobId – PandaID : one pilot could pick >1 analysis jobs
- Selection of Good Jobs:
  - finished successfully both in condor and Panda

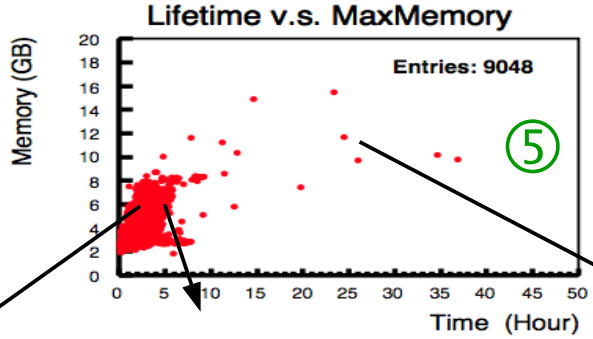
Multicore Simulation Jobs (5)

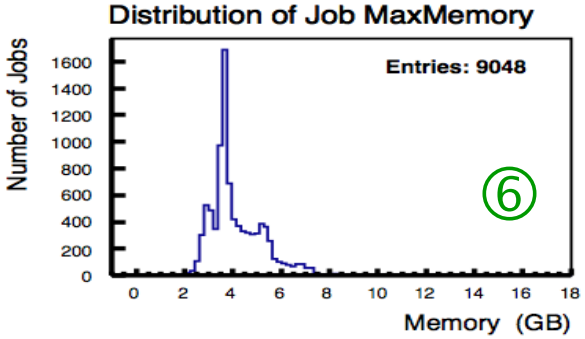


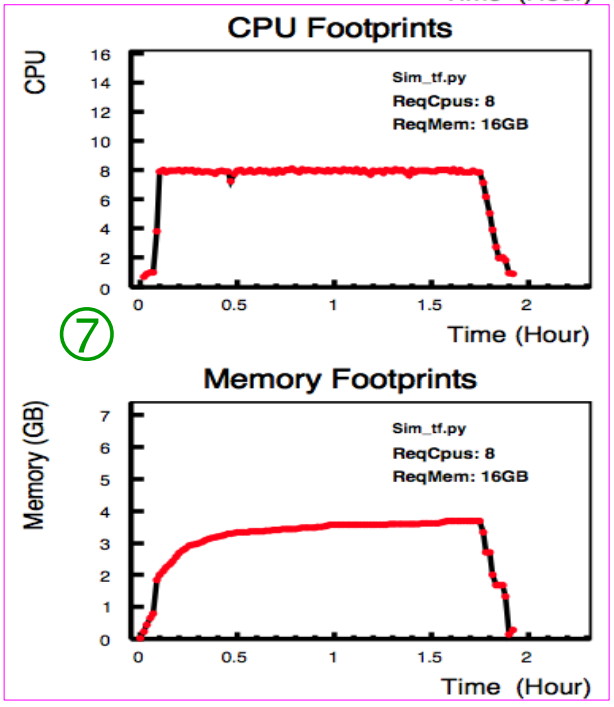


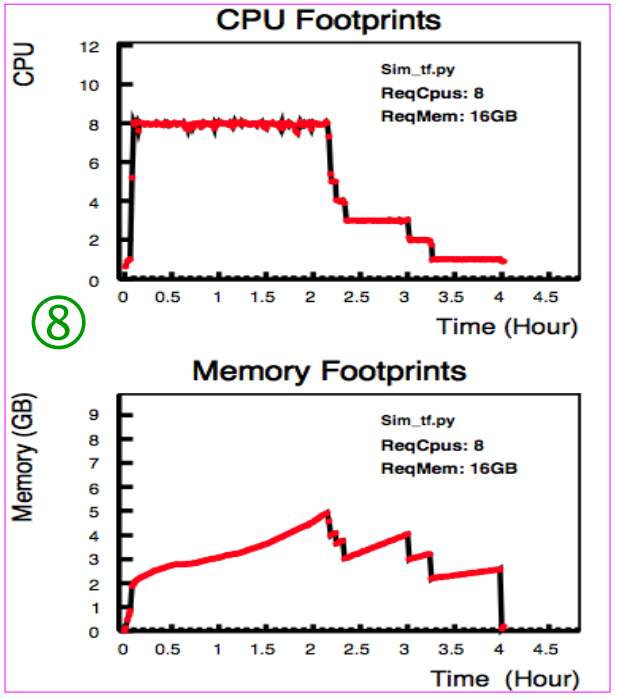


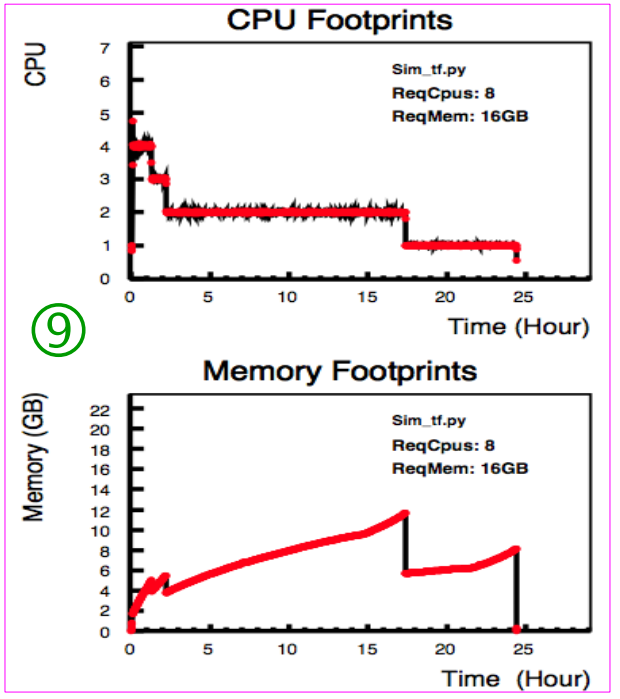




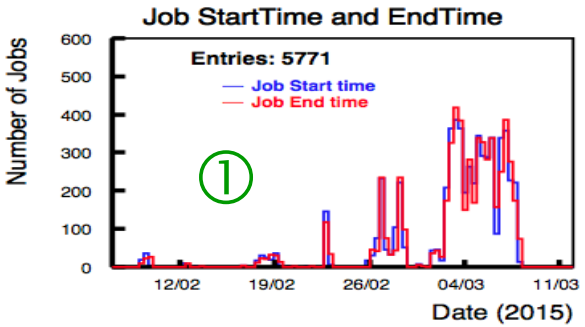


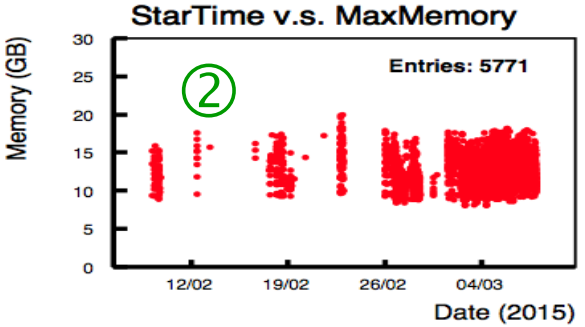


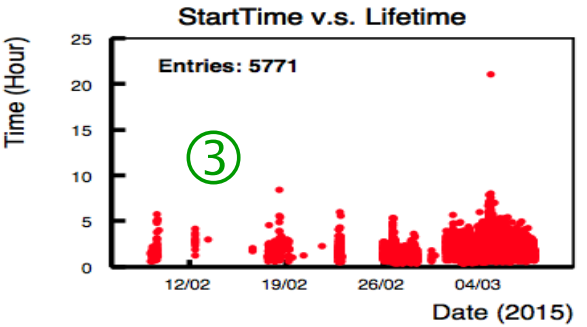


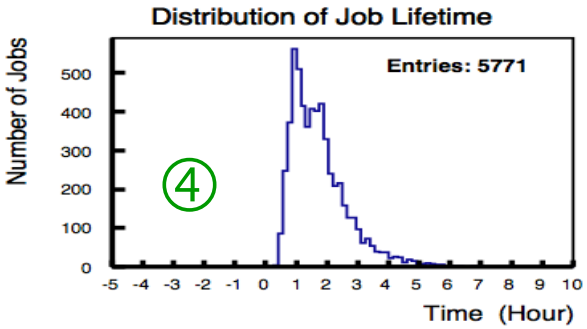


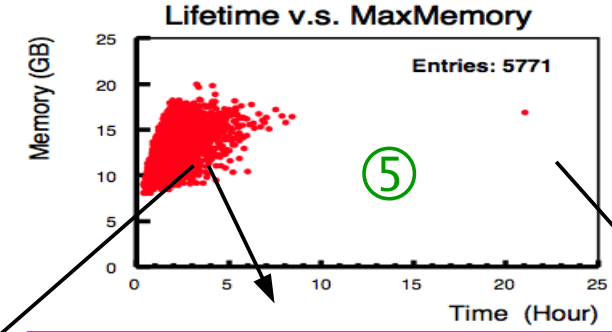
Multicore Reconstruction Jobs (6)

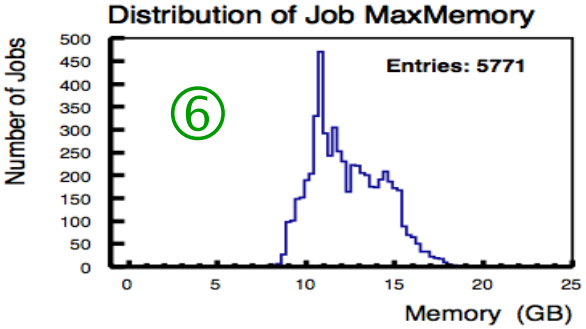


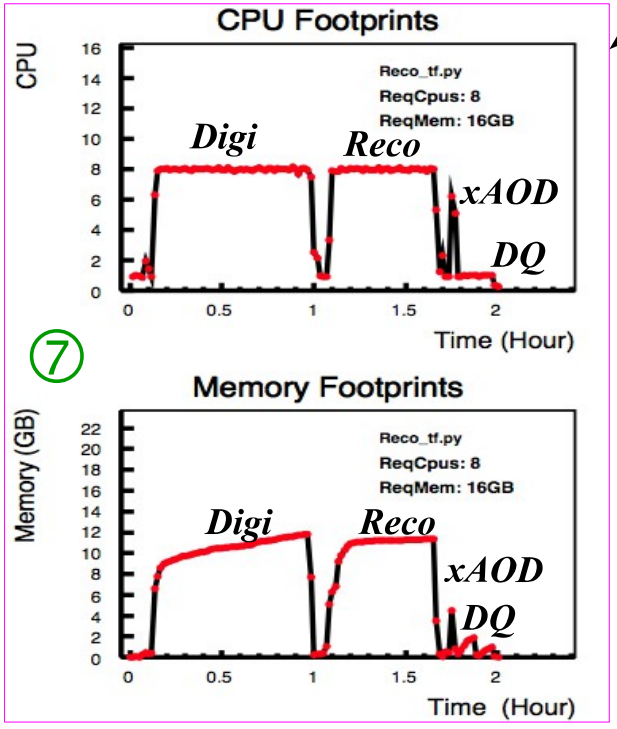


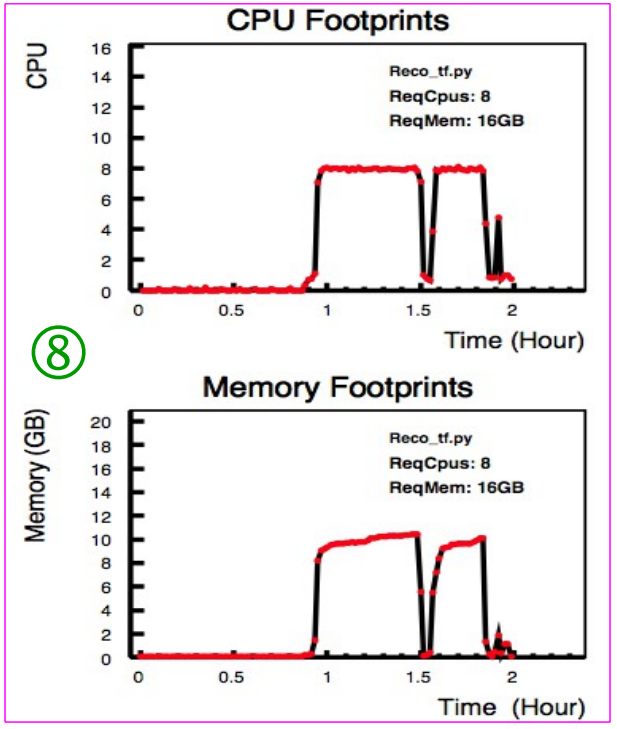


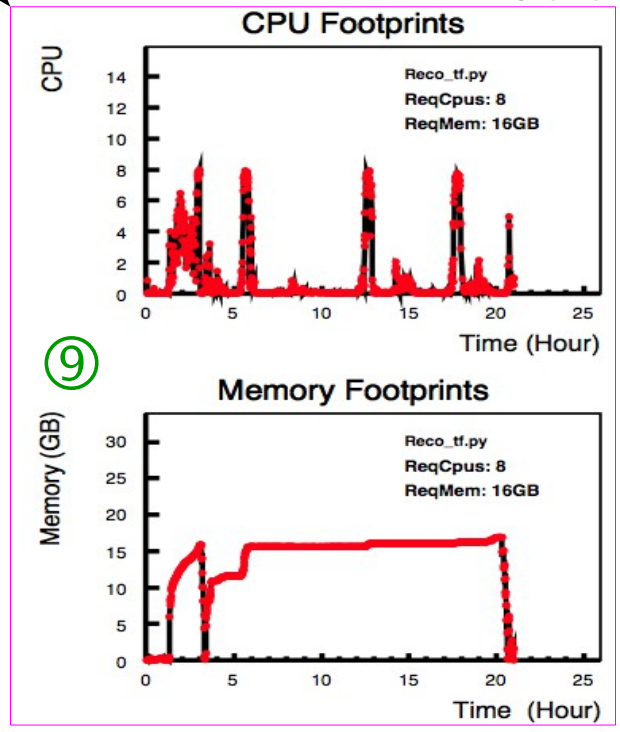




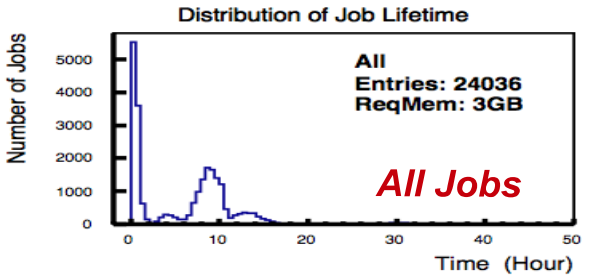


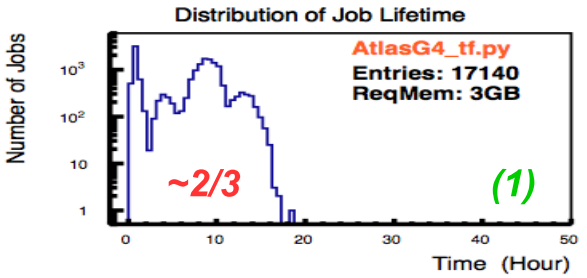


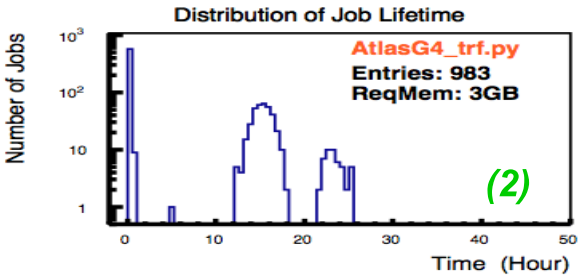


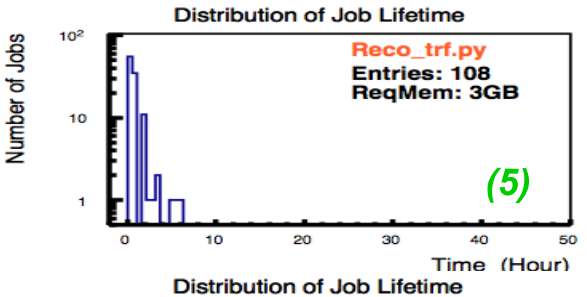


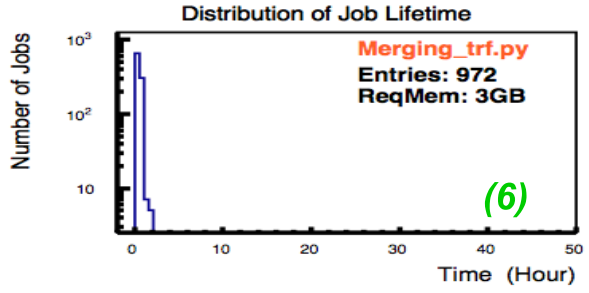
Lifetime of Singlecore Production Jobs (7)

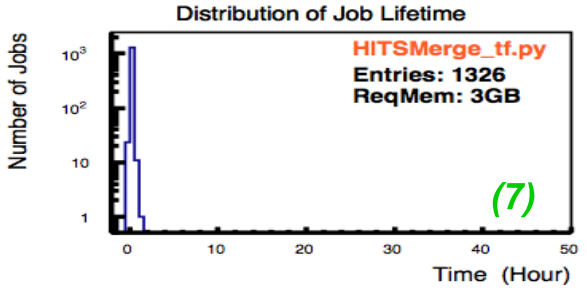


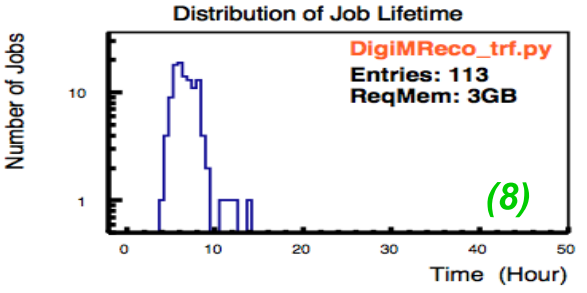


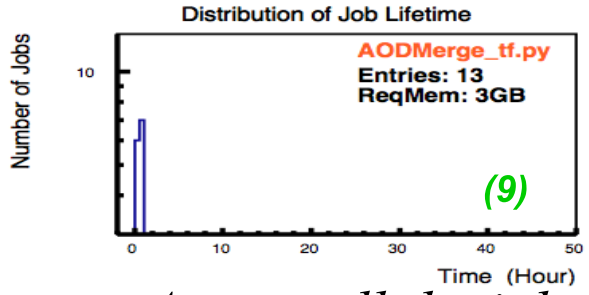


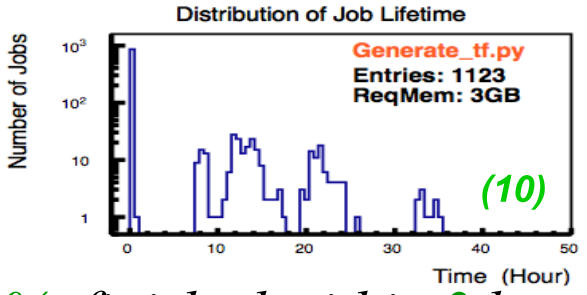






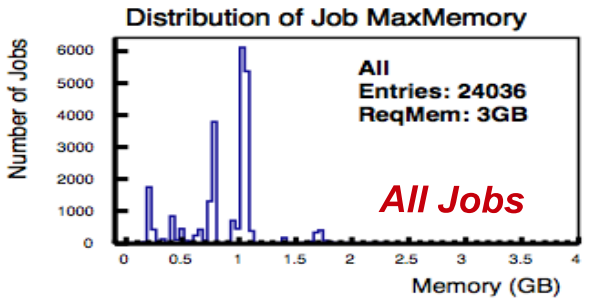


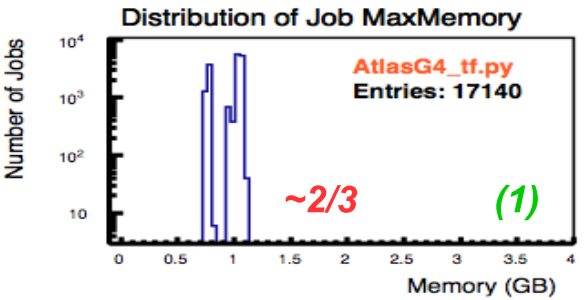


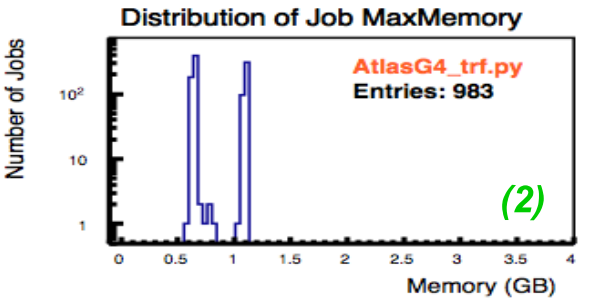


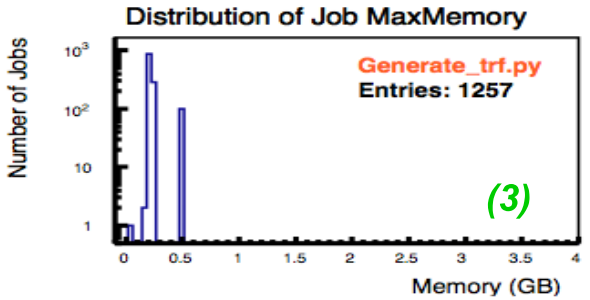
- *Among all the jobs, ~41% finished within 2 hours, ~56% finished between 2 and 20 hours, ~2.3% runs >20 hours.*

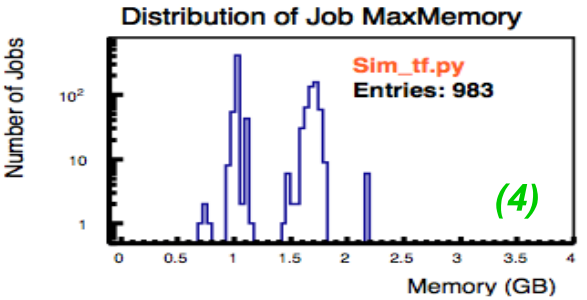
MaxMemory of Singlecore Production Jobs (8)

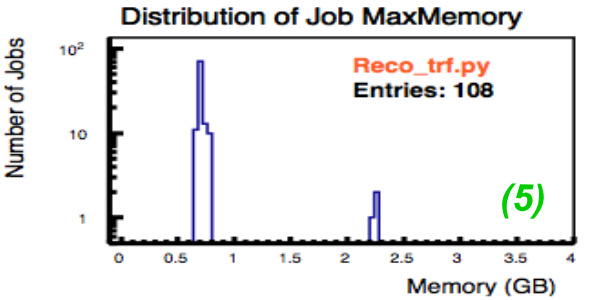


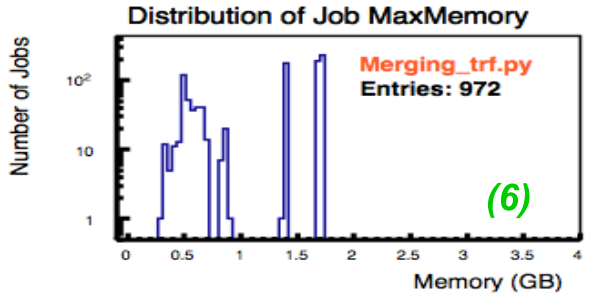


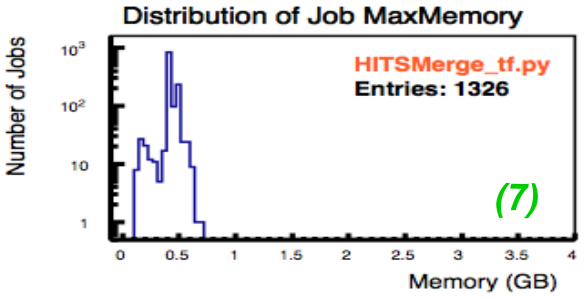


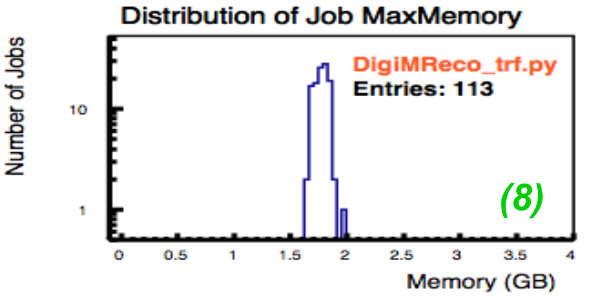


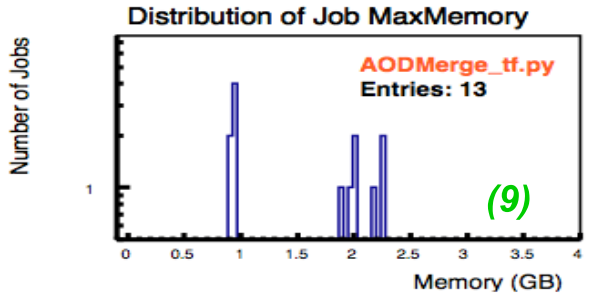


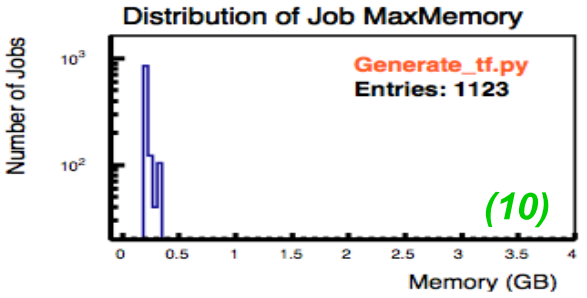


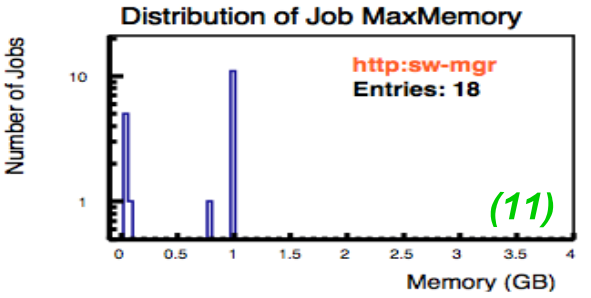






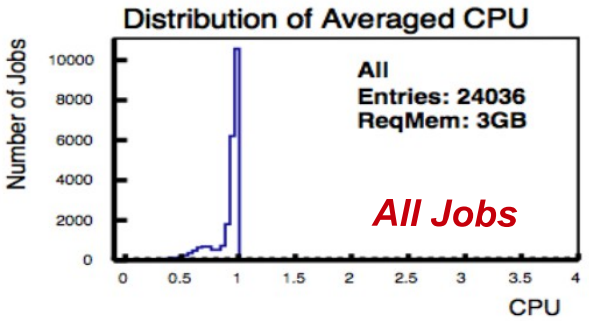


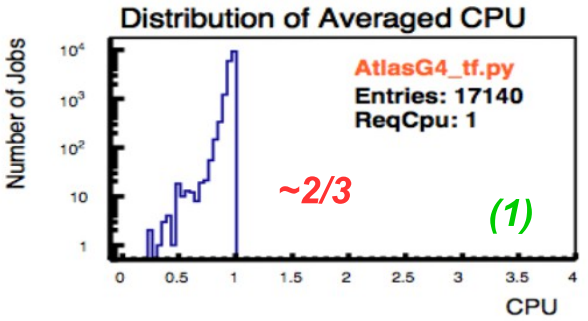


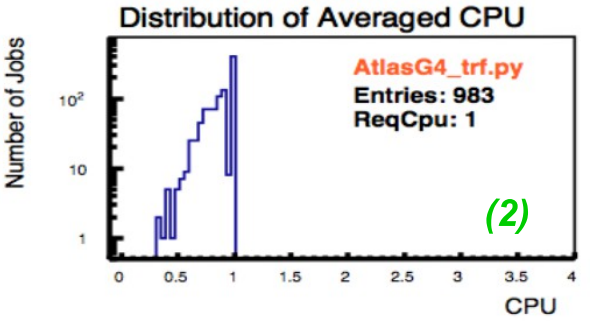


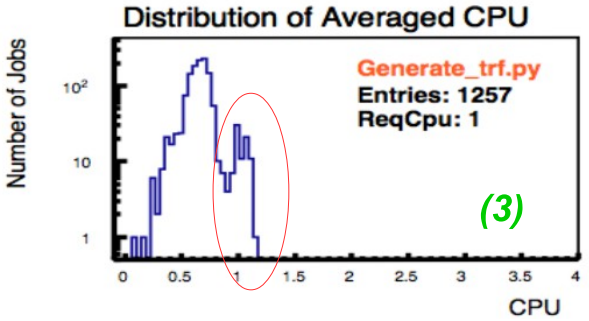
- *Among all the jobs, ~95% use < 1.2 GB, ~ 4.9% use within 1.2GB and 2GB range, <0.1% uses >2GB, none uses > 4GB*

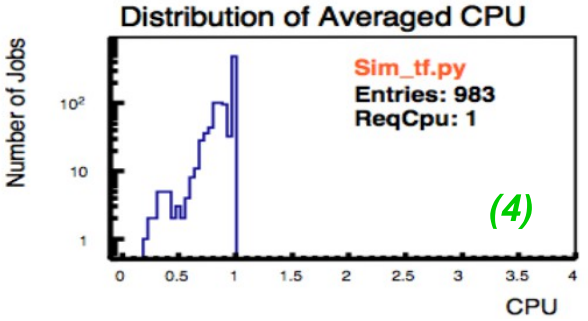
Averaged Cpu of Singlecore Production Jobs (9)

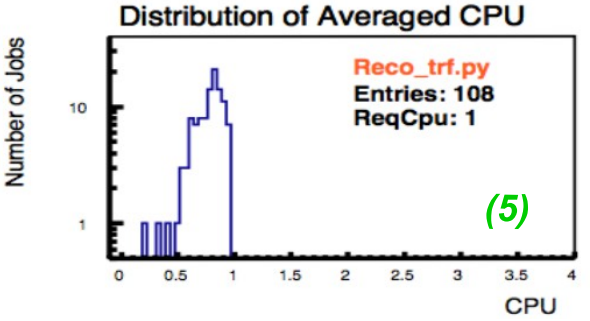


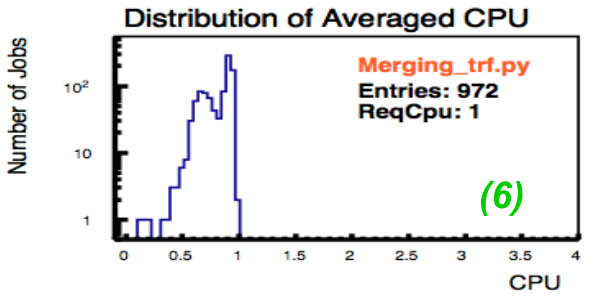


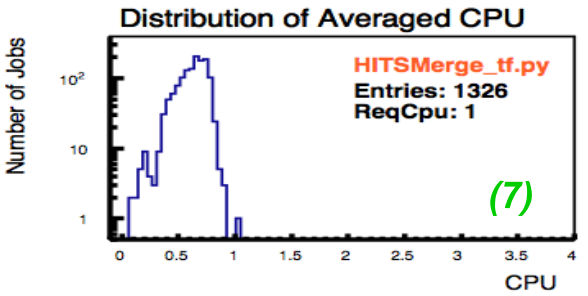


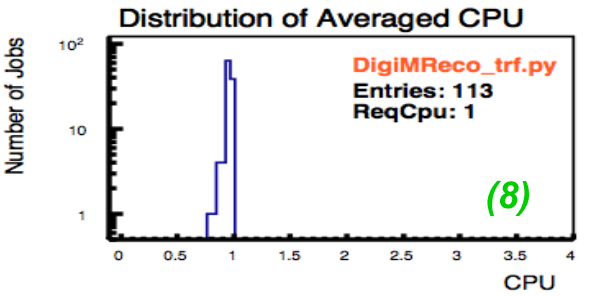


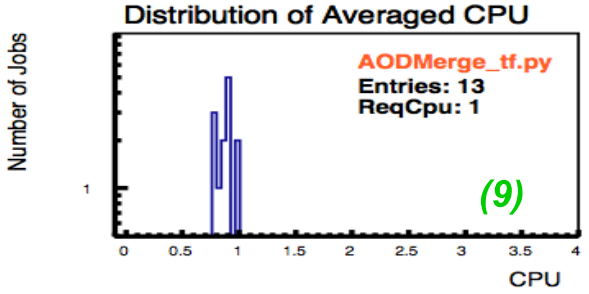


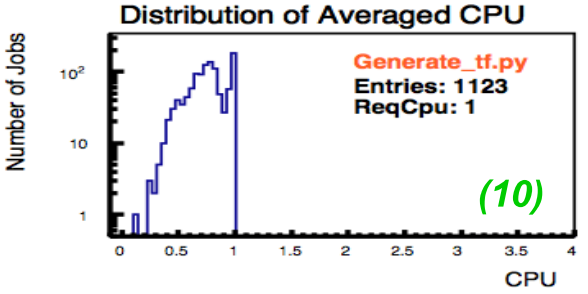


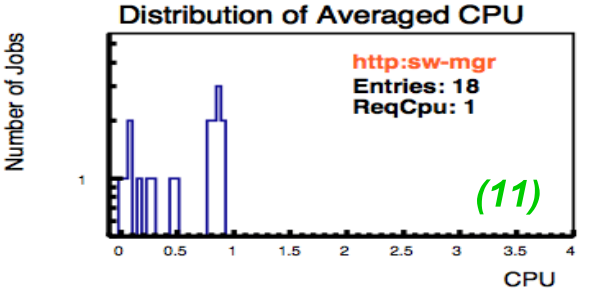






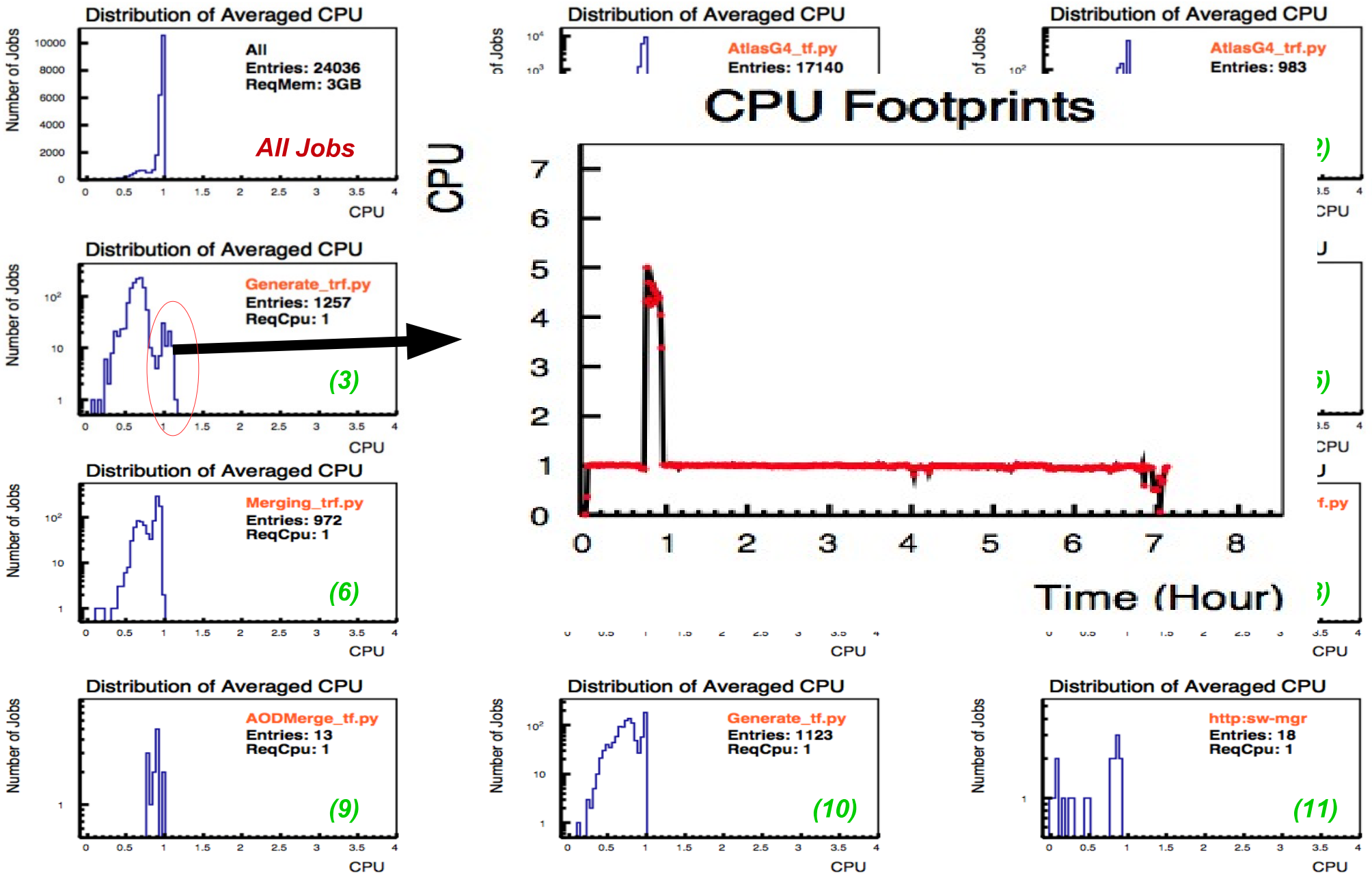






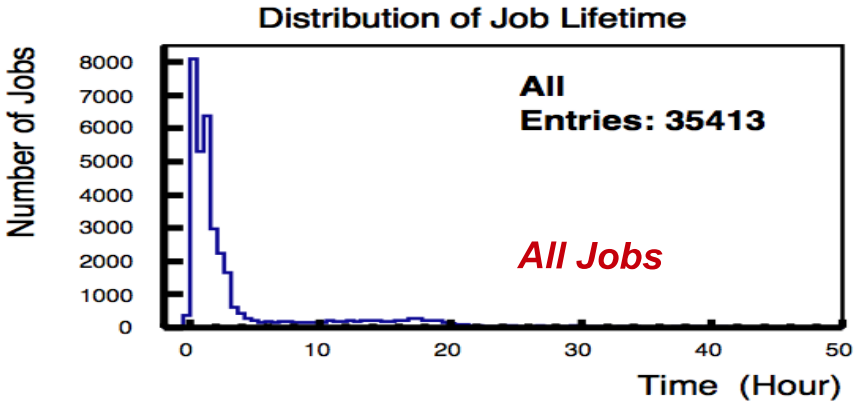
- *All < 1 except a few Generate_trf.py jobs*

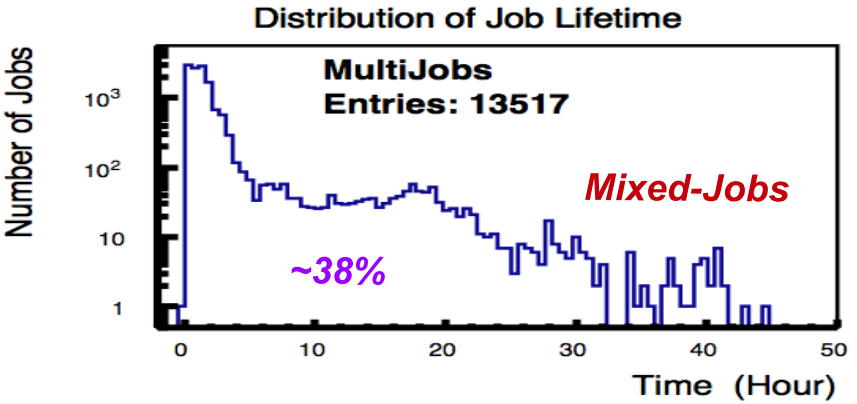
Averaged Cpu of Singlecore Production Jobs (9)

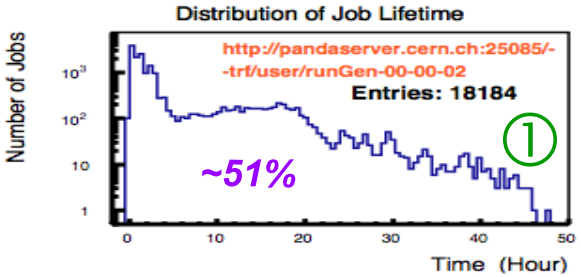


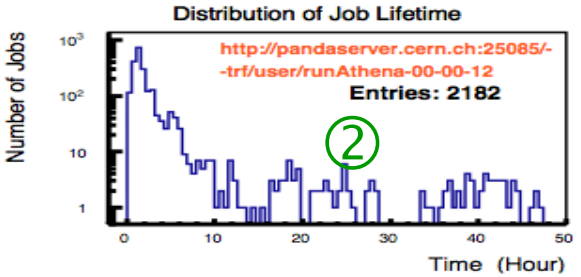
- *All < 1 except a few Generate_trf.py jobs*

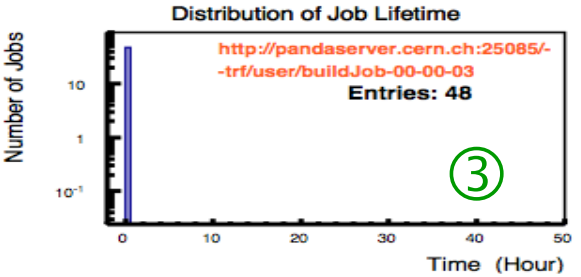
Lifetime Distribution of Analysis Jobs (10)

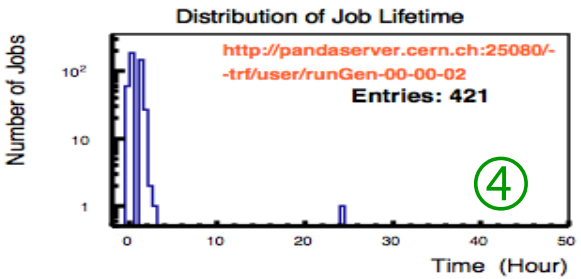


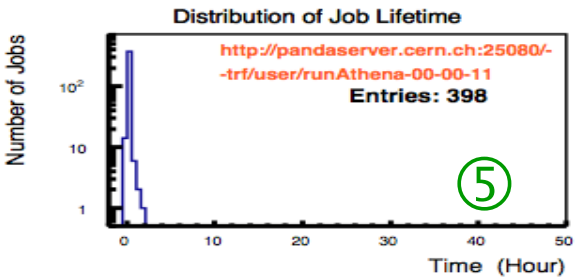


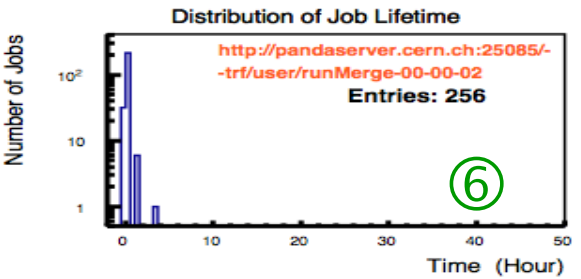


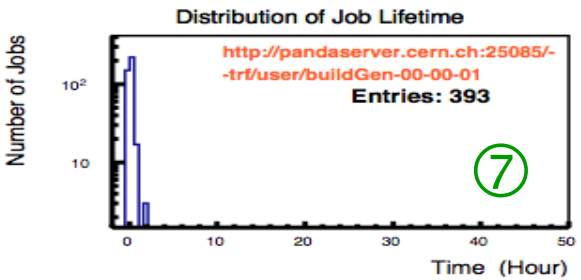


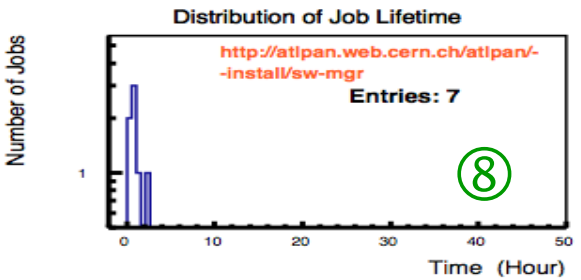


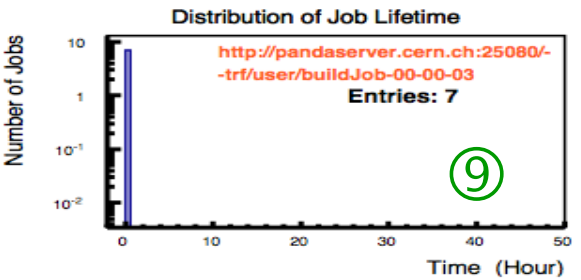






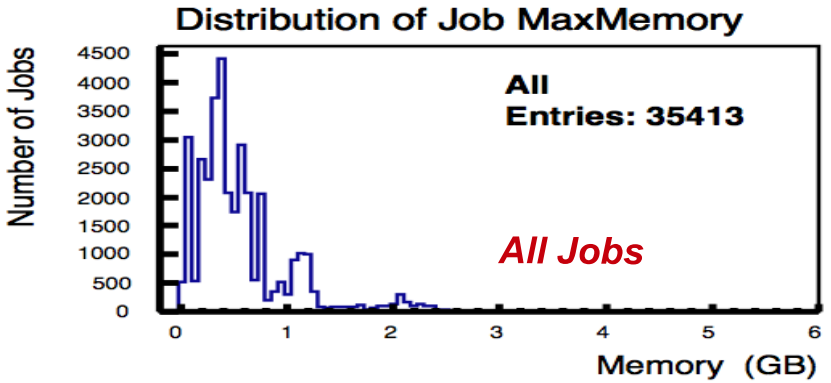


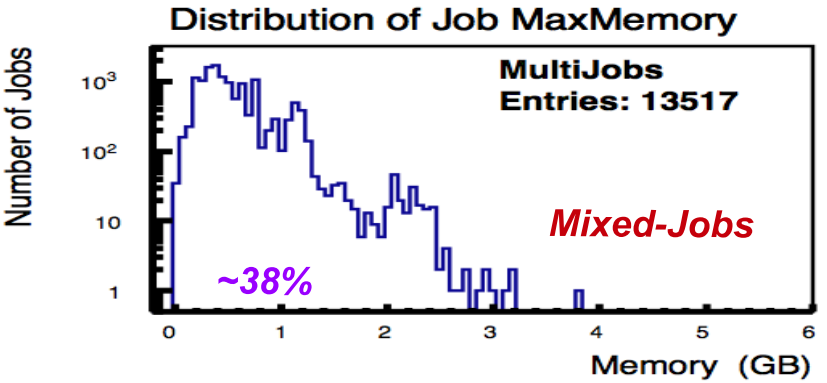


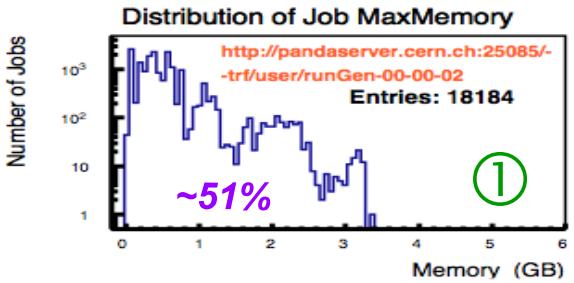


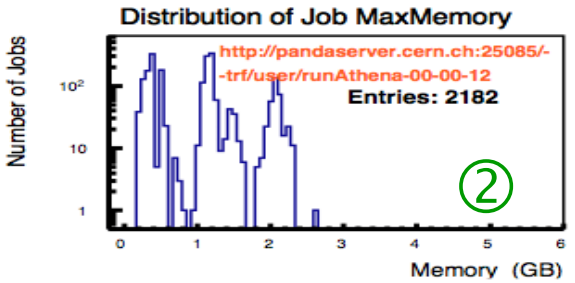
- *Among all the analysis jobs, ~63% finished within 2 hours, ~33% finished between 2 and 20 hours, 4% runs >20 hours.*

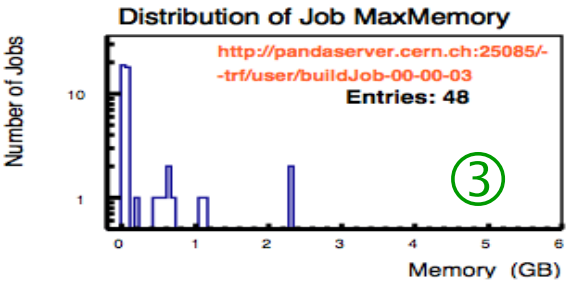
MaxMemory of Analysis Jobs

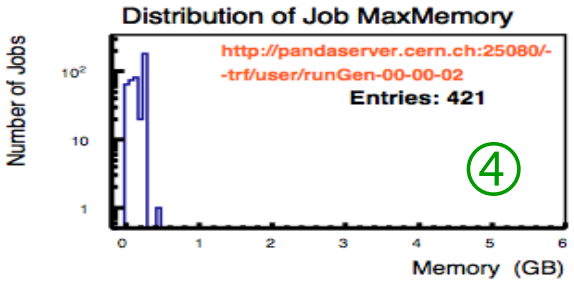


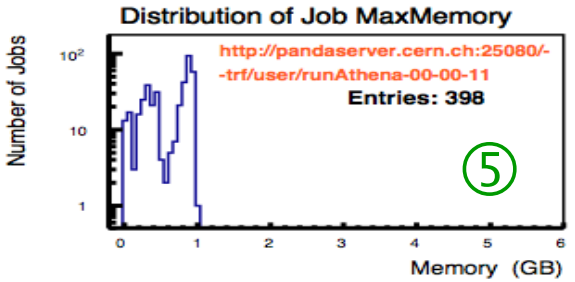


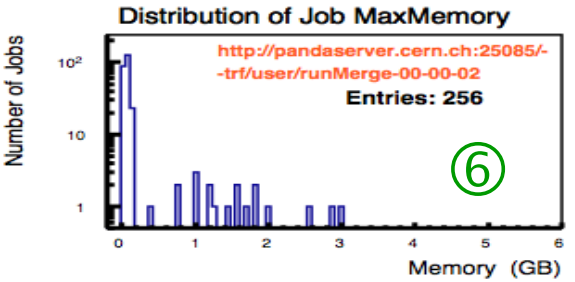


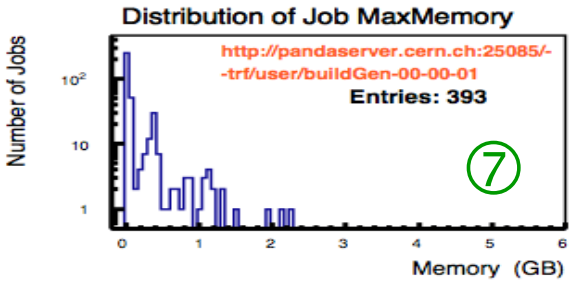


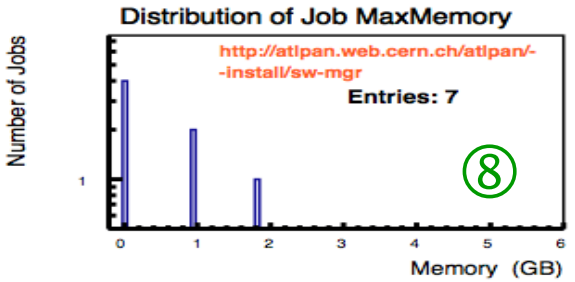


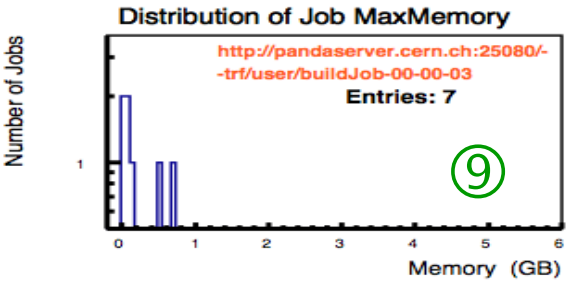






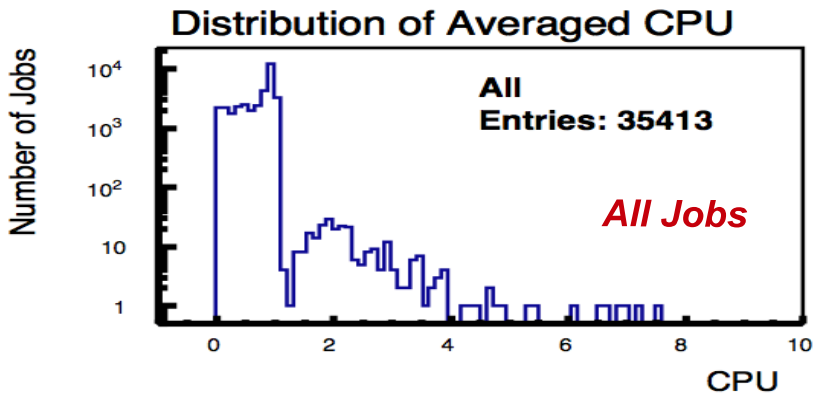


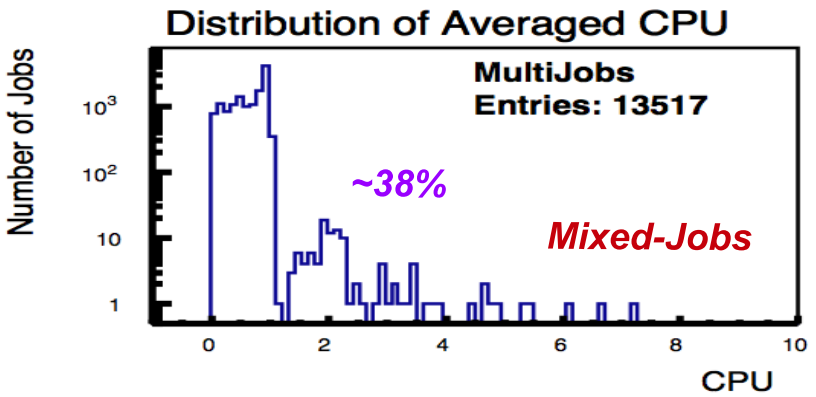


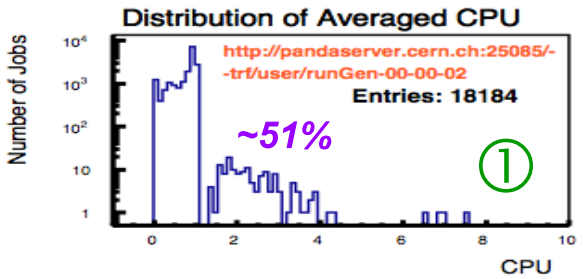


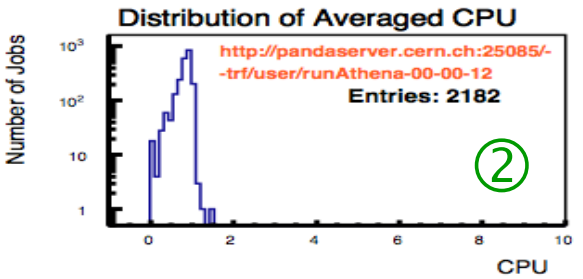
- *Among all the analysis jobs, ~94.5% use <1.5GB, ~5% uses 1.5-3GB, 0.2% uses >3GB.*

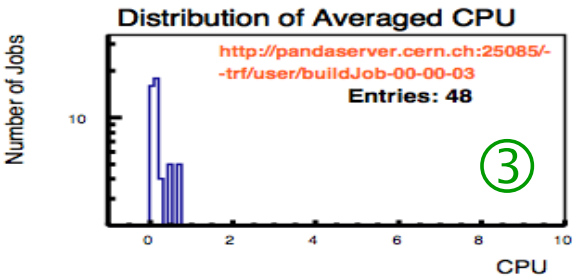
Averaged CPU of Analysis Jobs

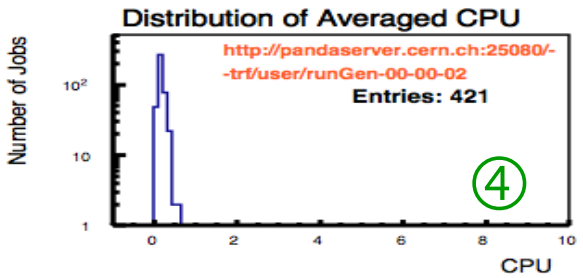


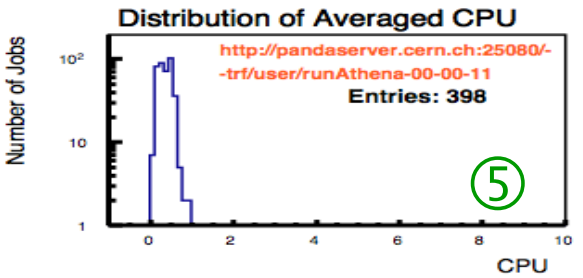


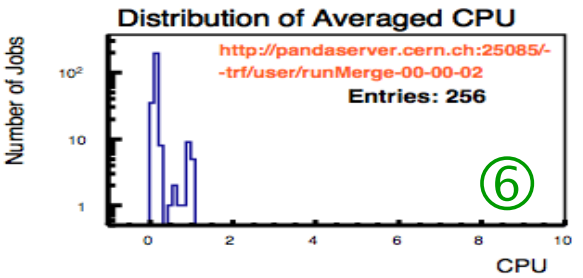


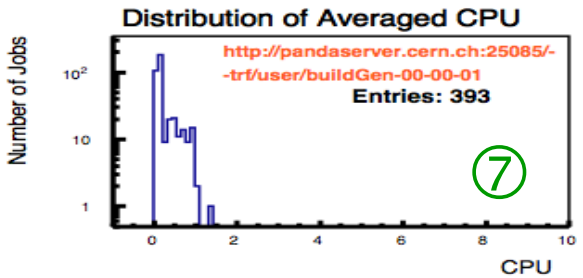


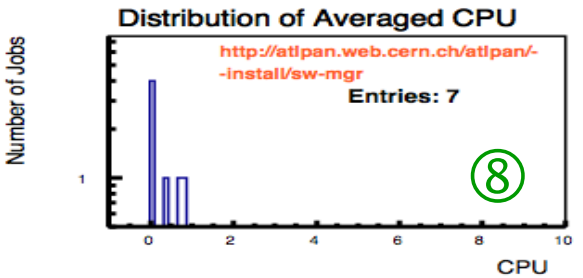


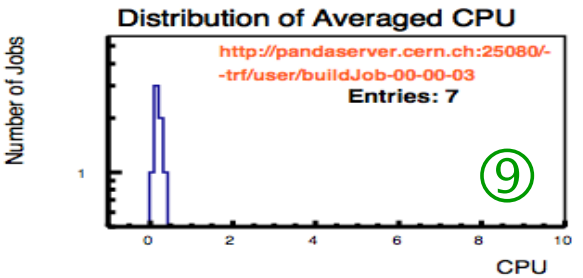






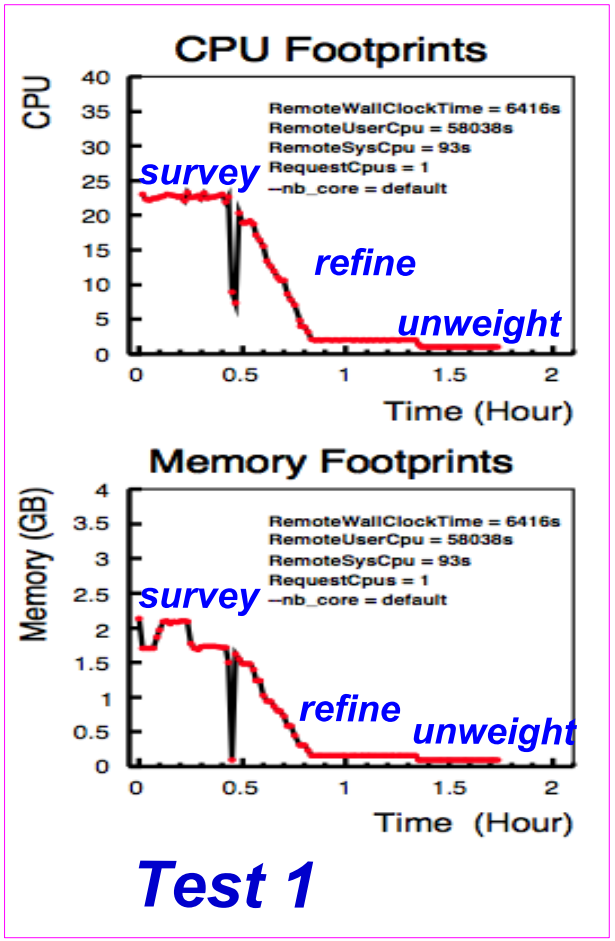


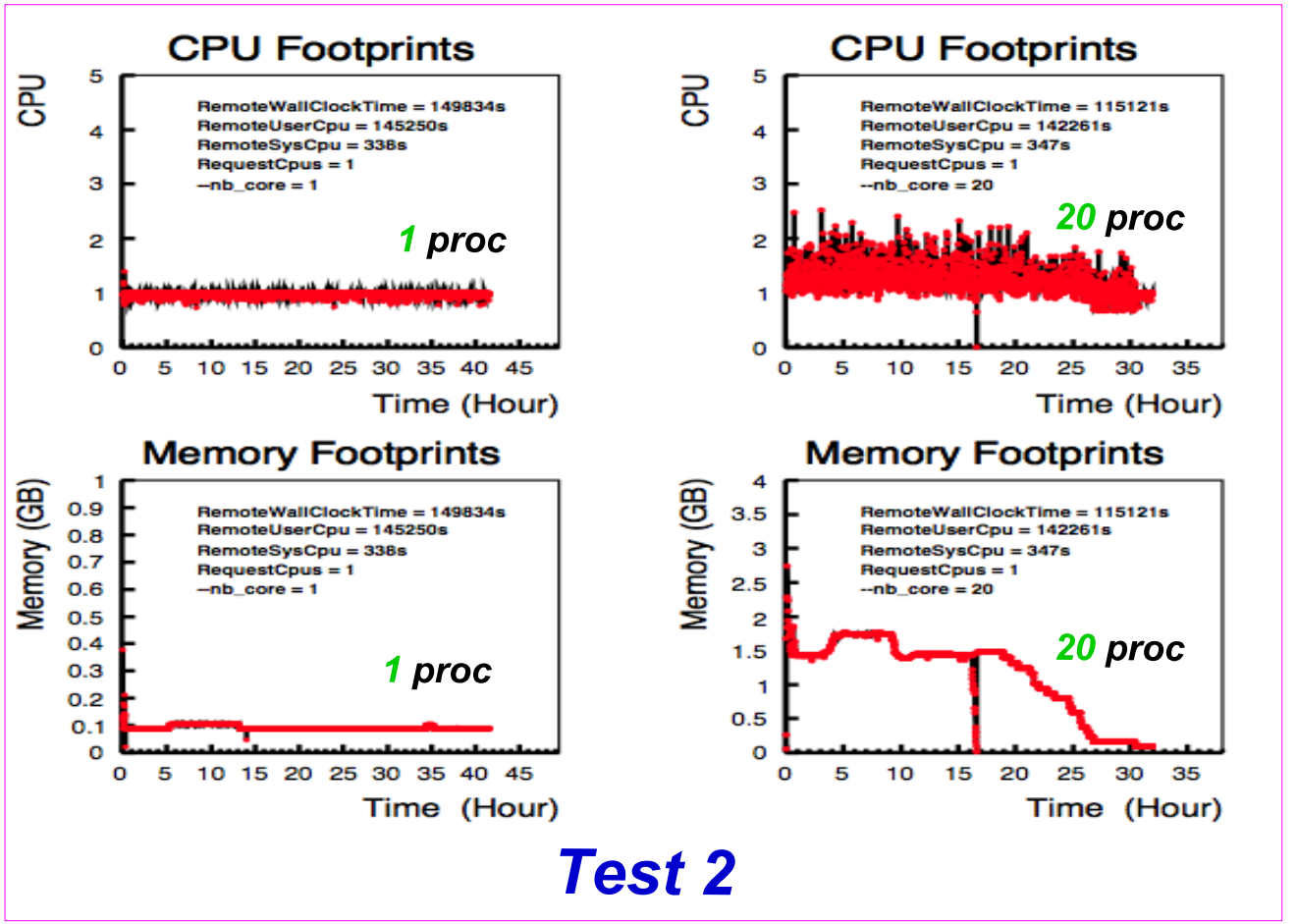




- *~1% jobs use > 1 cpu, known as Madevents jobs created by Madgraph*

Test of Madgraph jobs





- *Test node: node046, 24 core, 24GB physical memory, 24GB swap*
- *Test 1: run madgraph in default mode, e.g. without setting –nb_core*
- *Test 2: run madgraph with 1 and 20 processes separately, both in parallel with another condor job which stressed the other 23 cores all the time.*

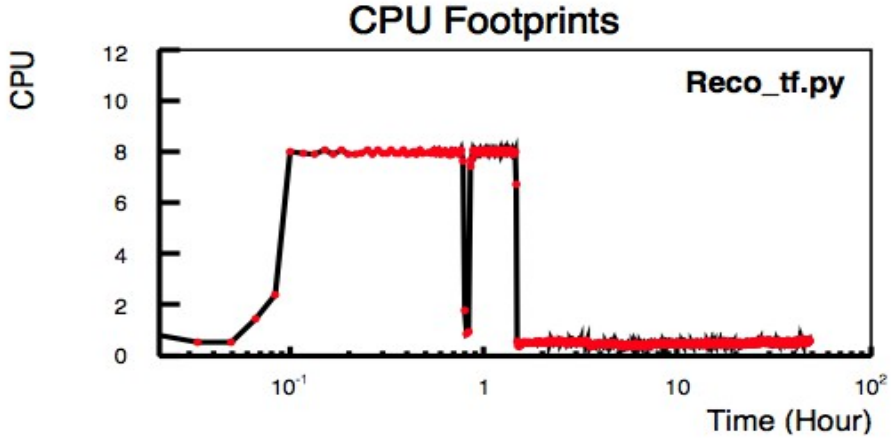
Memory Overcommit

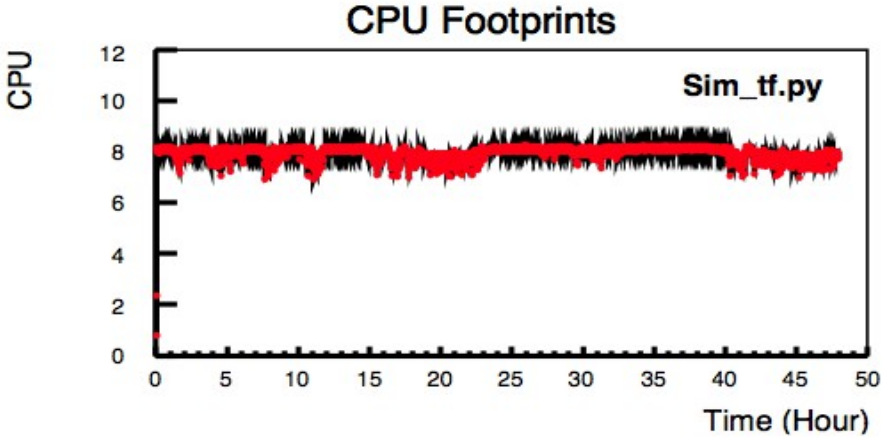
- Balance between Job's memory over-requesting and high resource usage
  - Optimization is difficult due to job's complexity and WN variation
- ATLAS jobs:

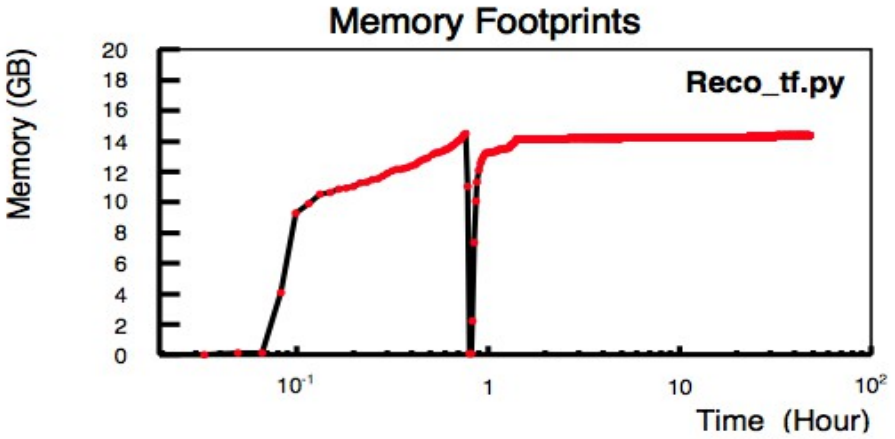
	Request_CPU	Request_MEM	MAX_MEM_used	Request Mem/cpu	Ideal Mem/cpu
PRODUCTION	1	3GB	95% < 1.2GB	3 GB	1.2 GB
	8	16GB	[2-8],[8-16]	2 GB	2 GB
ANALYSIS	1	4GB	94.5% < 1.5GB	4 GB	1.5 GB

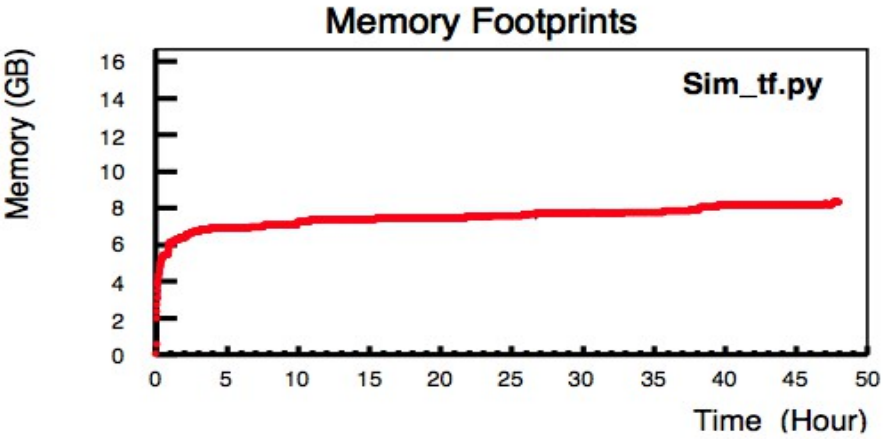
  - Multicore production jobs: request memory close to real usage
  - Single core jobs: request ~150% than real usage
  - ATLAS is planning to use RSS & SWAP to replace MEMORY in job description
- Site policy:
  - RSS > 2GB/CPU: overcommit_factor >= 2
  - RSS < 2GB/CPU: overcommit_factor < 2

Broken Multicore Jobs









- *Jobs could get broken at any step, and a broken job takes 48 hours (384 cpu-hours) while a normal multicore job only takes ~ 2 hours*
- *Suspicous job detecting system setup to track/kill suspicious multicore jobs*
  - *Reco_tf.py: cpu usage dropped to ~0.4 and memory usage unchanged*
  - *Sim_tf.py: running time > 10 hours (old Sim_tf.py jobs)*

Future Work

- Rerun the analysis on larger time scale
- Monthly calibation
  - To detect the change of Job properties
  - Update local resource policy correspondingly
- Integration into site monitoring/security tools
  - Online tracking of cpu/memory footprints of selected job
  - Online tracking of suspicous jobs
    - Enable the killing of broken multicore reconstruction jobs
- Expand the analysis to more VOs
  - CMS?
  - Small VOs?

Questions?