

CLOUD COMPUTING CONCEPTS

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SCHEDULING

Lecture C

DOMINANT-RESOURCE
FAIR SCHEDULING

CHALLENGE

- What about scheduling VMs in a cloud (cluster)?
- Jobs may have multi-resource requirements
 - Job 1's tasks: 2 CPUs, 8 GB
 - Job 2's tasks: 6 CPUs, 2 GB
- How do you schedule these jobs in a “fair” manner?
- That is, how many tasks of each job do you allow the system to run concurrently?
- What does fairness even mean?

DOMINANT RESOURCE FAIRNESS (DRF)

- Proposed by researchers from U. California Berkeley
- Proposes notion of fairness across jobs with multi-resource requirements
- They showed that DRF is
 - Fair for multi-tenant systems
 - Strategy-proof: tenant can't benefit by lying
 - Envy-free: tenant can't envy another tenant's allocations

WHERE IS DRF USEFUL?

- DRF is
 - Usable in scheduling VMs in a cluster
 - Usable in scheduling Hadoop in a cluster
- DRF used in Mesos, an OS intended for cloud environments
- DRF-like strategies also used some cloud computing company's distributed OS's

How DRF Works

- Our example
 - Job 1's tasks: 2 CPUs, 8 GB
 - => Job 1's resource vector = <2 CPUs, 8 GB>
 - Job 2's tasks: 6 CPUs, 2 GB
 - => Job 2's resource vector = <6 CPUs, 2 GB>
- Consider a cloud with <18 CPUs, 36 GB RAM>

How DRF Works (2)

- Our example
 - Job 1's tasks: 2 CPUs, 8 GB
 - \Rightarrow Job 1's resource vector = $\langle 2 \text{ CPUs}, 8 \text{ GB} \rangle$
 - Job 2's tasks: 6 CPUs, 2 GB
 - \Rightarrow Job 2's resource vector = $\langle 6 \text{ CPUs}, 2 \text{ GB} \rangle$
- Consider a cloud with $\langle 18 \text{ CPUs}, 36 \text{ GB RAM} \rangle$
- Each Job 1's task consumes % of total CPUs = $2/18 = 1/9$
- Each Job 1's task consumes % of total RAM = $8/36 = 2/9$
- $1/9 < 2/9$
 - \Rightarrow Job 1's dominant resource is RAM, i.e., Job 1 is more memory-intensive than it is CPU-intensive

How DRF Works (3)

- Our example
 - Job 1's tasks: 2 CPUs, 8 GB
 - \Rightarrow Job 1's resource vector = $\langle 2 \text{ CPUs}, 8 \text{ GB} \rangle$
 - Job 2's tasks: 6 CPUs, 2 GB
 - \Rightarrow Job 2's resource vector = $\langle 6 \text{ CPUs}, 2 \text{ GB} \rangle$
- Consider a cloud with $\langle 18 \text{ CPUs}, 36 \text{ GB RAM} \rangle$
- Each Job 2's task consumes % of total CPUs = $6/18 = 6/18$
- Each Job 2's task consumes % of total RAM = $2/36 = 1/18$
- $6/18 > 1/18$
 - \Rightarrow Job 2's dominant resource is CPU, i.e., Job 1 is more CPU-intensive than it is memory-intensive

DRF FAIRNESS

- For a given job, the % of its dominant resource type that it gets cluster-wide, is the same for all jobs
 - Job 1's % of RAM = Job 2's % of CPU
- Can be written as linear equations, and solved

DRF SOLUTION, FOR OUR EXAMPLE

- DRF Ensures
 - Job 1's % of RAM = Job 2's % of CPU
- Solution for our example:
 - Job 1 gets 3 tasks each with <2 CPUs, 8 GB>
 - Job 2 gets 2 tasks each with <6 CPUs, 2 GB>
 - Job 1's % of RAM
 - = Number of tasks * RAM per task / Total cluster RAM
 - = $3 * 8 / 36 = 2/3$
 - Job 2's % of CPU
 - = Number of tasks * CPU per task / Total cluster CPUs
 - = $2 * 6 / 18 = 2/3$

OTHER DRF DETAILS

- DRF generalizes to multiple jobs
- DRF also generalizes to more than 2 resource types
 - CPU, RAM, Network, Disk, etc.
- DRF ensures that each job gets a fair share of that type of resource which the job desires the most
 - Hence fairness

SUMMARY: SCHEDULING

- Scheduling very important problem in cloud computing
 - Limited resources, lots of jobs requiring access to these jobs
- Single-processor scheduling
 - FIFO/FCFS, STF, Priority, Round-Robin
- Hadoop scheduling
 - Capacity scheduler, Fair scheduler
- Dominant-Resources Fairness