

Introduction

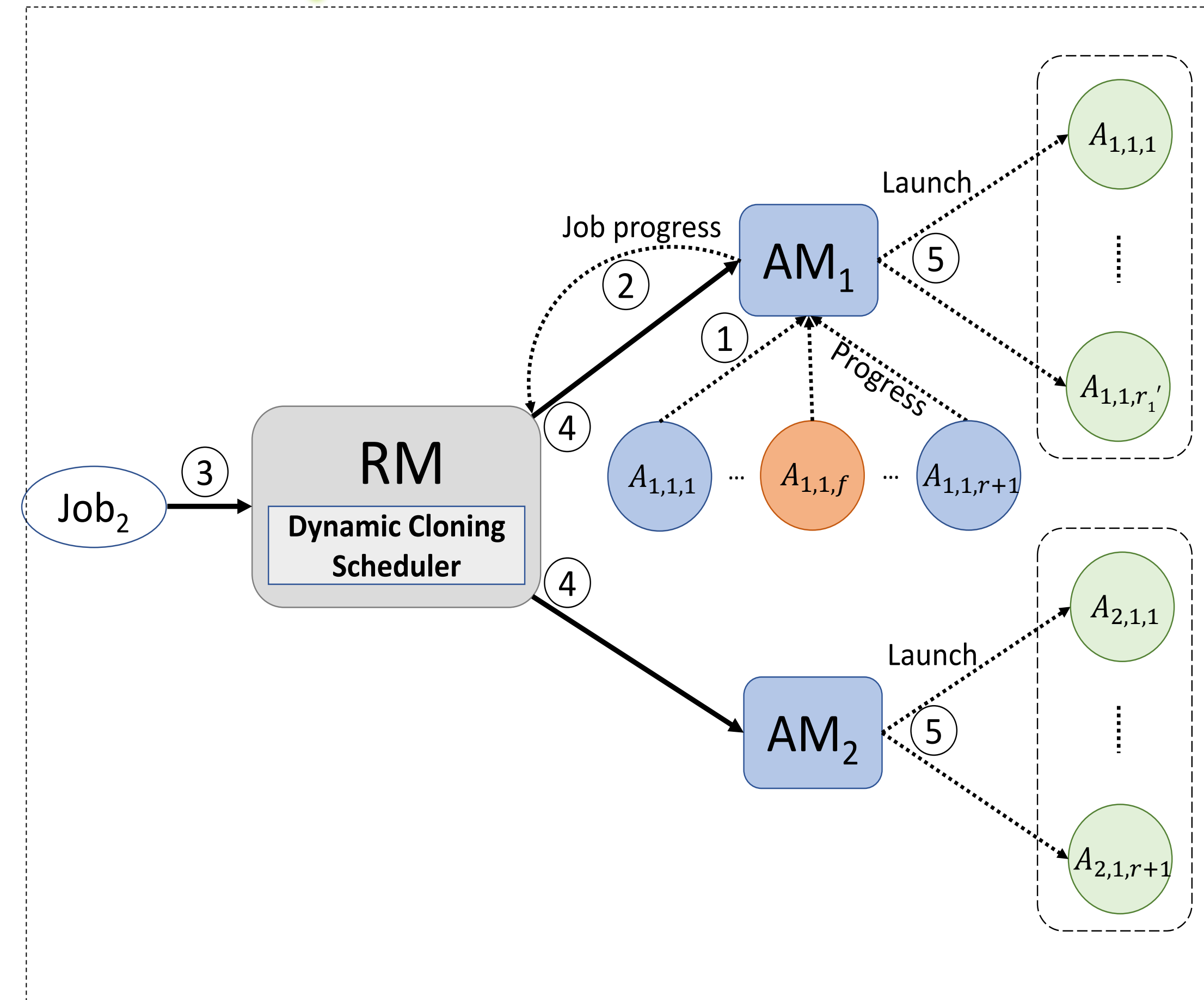
- Demand for cloud-based processing frameworks continues to grow
- Cloud applications is becoming increasingly mission-critical and deadline sensitive, especially in shared clusters
- Cloud providers seek efficient techniques to meet the SLA
- A few slow tasks, called stragglers, could significantly impact job execution time
- Launching extra attempts (clones) for each task upon submission can mitigate stragglers
- This work proposes Shed, an optimization framework that leverages dynamic cloning to jointly maximize jobs' Probability of Completion before Deadline (PoCD) by fully utilizing the available resources

System Model

Consider J jobs submitted to the Mapreduce processing framework. Each job j is associated with a deadline D_j that is determined by a user. Each job j consists of N_j tasks, and it is considered successful if all its N_j tasks are executed and completed before the job deadline D_j . Let T_j denote job j 's completion time, and $T_{j,i}$ for $i = 1, \dots, N_j$ be the (random) completion times of tasks belongs to job j . Any task whose execution time exceeds the deadline is considered a straggler. Our dynamic cloning approach mitigate the effect of stragglers by proactively launching r_j extra attempts for each task. A task is finished once any one of the $r_j + 1$ attempts finishes execution, and then the other copies are killed. Thus, task's completion time $T_{j,i}$ is determined by the completion time of the fastest attempt, i.e.,

$$T_{j,i} = \min_{k=1, \dots, r_j+1} T_{j,i,k}, \forall i, j.$$

System Architecture



Joint PoCD Optimization

$$\begin{aligned} & \text{maximize} && \sum_{j=1}^J U(p_j), \\ & \text{s.t.} && \sum_{j=1}^J N_j \cdot (r_j + 1) + |J| \leq \lambda \cdot m \\ & && p_j = R_j(r_j), \forall j \\ & && r_j \geq 0, \forall j \end{aligned}$$

Proposed Online Algorithm

Algorithm 1: Proposed Online Algorithm

- 1: Upon submission of a new job:
- 2: Kill all jobs which missed their deadlines
- 3: $J = \{j_1, j_2, j_3, \dots\}$
- 4: if $|J| == 1$ then
- 5: $r_{\max} = \lfloor \frac{\lambda \cdot m - N_1 - 1}{N_1} \rfloor$
- 6: $r_1 = r_{\max}$
- 7: else
- 8: $r_j = 0 \forall j$
- 9: $\omega = 0$
- 10: $\kappa = \lambda \cdot m - \sum_{j=1}^J N_j - |J|$
- 11: Calculate $R_j \forall j$
- 12: while $J \neq \{\emptyset\}$ do
- 13: $j' = \arg \min_j \{R_j\}$
- 14: if $N_{j'} + \omega > \kappa$ then
- 15: $J = J - \{j'\}$
- 16: else
- 17: $r_{j'} = r_{j'} + 1$
- 18: $\omega = \omega + N_{j'}$
- 19: Calculate $R_j \forall j$
- 20: end if
- 21: end while
- 22: end if

Theorems

$$R_{su} = \left[1 - \left(\frac{t_{\min}}{D_j} \right)^{\beta \cdot (r_j + 1)} \right]^{N_j}$$

$$R_{ru} = \left[1 - \left(\frac{(1 - \phi_i) t_{\min}}{D_j - \tau_j} \right)^{\beta \cdot (r_j + 1)} \right]^{N_j}$$

Conclusion

In this work, we propose Shed, an optimization framework that leverages dynamic cloning to jointly maximize PoCD and cluster utilization. We also present an online scheduler that dynamically optimize resources upon new job arrival. Our solution includes an online greedy algorithm to find the optimal number of clones needed for each job. Our results show, in some cases, that Shed can achieve 100% PoCD compared to Dolly and Hadoop with speculation. The proposed algorithm is able to achieve more than 90% utilization of available cloud resources, whereas Dolly and Hadoop achieves only about 22%.

Reference

- J. Dean *et al*, 2008
- G. Ananthanarayanan *et al*, 2013

Evaluation

