# Communication-aware task placement for workflow scheduling on DaaS-based Cloud

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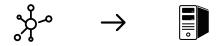
## Scheduling: Matching jobs and resources







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## Scheduling: Matching jobs and resources







- Jobs definition is changing,
- Resources are changing.



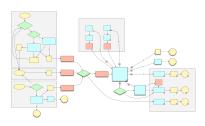




### Challenge

The matching logic need to consider chose changes.

Complex jobs (workflows);



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- Multi-Tenant (collaborative);



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- Dynamic platform (laaS Cloud with DaaS storage).

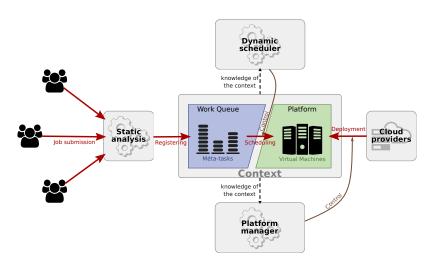


- Complex jobs (workflows);
- Multi-Tenant (collaborative);
- Dynamic platform (laaS Cloud with DaaS storage).

#### State of the art

Previous work considers at most 2 of those 3 factors.

Problem subdivision



Framework architecture

### DCP static scheduling algorithm

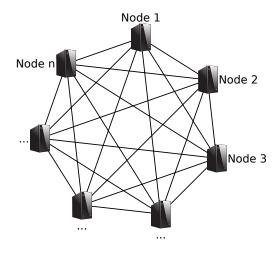
```
\mathcal{C} \leftarrow \text{empty clustering}
                                               compute BL and TL for each task using C
while \exists unmarked dependency between tasks do
    (u, v) \leftarrow \text{edge} with the largest path length (most critical).
Resolve ties by edge size (select largest).
   C' \leftarrow C.mergeClusters(u, v)
    compute BL' and TL' for each task using C'
    if DCPL(BL', TL') \leq DCPL(BL, TL) then
        (C, TL, BL) \leftarrow (C', TL', BL')
    end if
    mark(u, v)
end while
return \mathcal{C}
```

Y.-K. Kwok and I. Ahmad, "A static scheduling algorithm onto multiprocessors," in Proceedings of the 1994 International Conference on Parallel Processing.

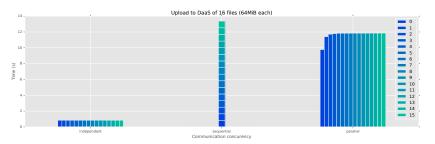
$$c(u, v) = \begin{cases} 0 & \text{if } \mathcal{C}(u) = \mathcal{C}(v) \\ \omega(u \to v) & \text{otherwise} \end{cases}$$

$$TL(v) = \begin{cases} 0 & \text{if } v \text{ has no predecessor} \\ \max_{u \in pred(v)} TL(u) + \omega(u) + c(u, v), \\ \text{avail}_{TL}(\mathcal{C}, v)) \end{cases}$$

$$BL(u) = \begin{cases} \omega(u) & \text{if } u \text{ has no successor} \\ \omega(u) + \max_{v \in succ(u)} c(u, v) + BL(v), \\ \text{avail}_{BL}(\mathcal{C}, u)) \end{cases}$$

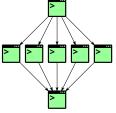


## Interference between concurrent communications



Transferring files between one node in sagittaire cluster (Grid'5000)//and a DaaS (storage5K)

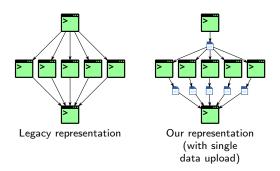
# See dependencies from the Dataflow point of view



Legacy representation

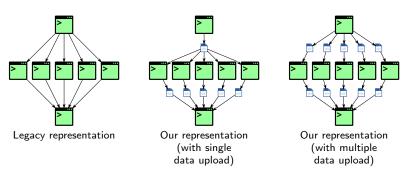
Representation of a fork-join DAG with n = 5 independent jobs.

## See dependencies from the Dataflow point of view



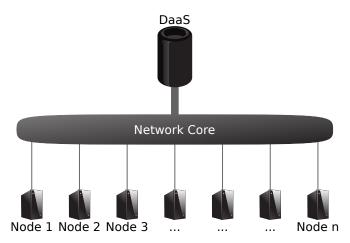
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## See dependencies from the Dataflow point of view



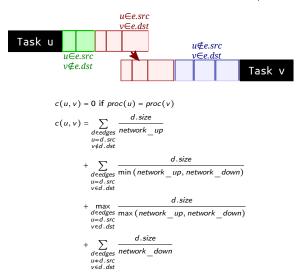
Representation of a fork-join DAG with n = 5 independent jobs.

# Reconsidering the network topology



A generic model of DaaS-based network topology.

#### Communications between two tasks on a DaaS-based platform.

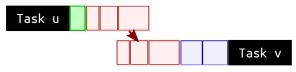


## Locality

$$islocal(d, v) = \begin{cases} 1 & \text{if } proc(d.src) = proc(v) \\ 0 & \text{otherwise} \end{cases}$$
$$islocal(d) = \prod_{v \in d.dst} islocal(d, v)$$

Adapting the Critical Path computation

Communications between two tasks on a DaaS-based platform (with locality).



$$c_{loc}(u, v) = 0 \text{ if } proc(u) = proc(v)$$

$$c_{loc}(u, v) = \sum_{\substack{d \in edges \\ u = d. src \\ v \notin d. dst \\ islocal(d) = 0}} \frac{d. size}{network\_up}$$

$$+ \sum_{\substack{d \in edges \\ u = d. src \\ v \in d. dst }} \frac{d. size}{min (network\_up, network\_down)}$$

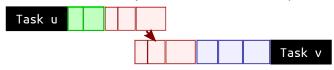
$$+ \sum_{\substack{d \in edges \\ u = d. src \\ v \in d. dst }} \frac{d. size}{max (network\_up, network\_down)}$$

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Adapting the Critical Path computation

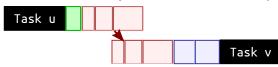
Different nodes w\ no locality, worst case communications i→j



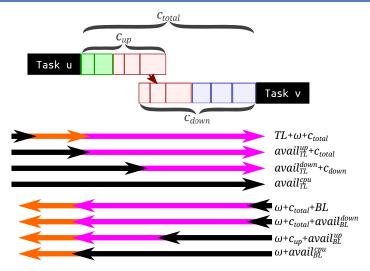
Same node, no communication i→j



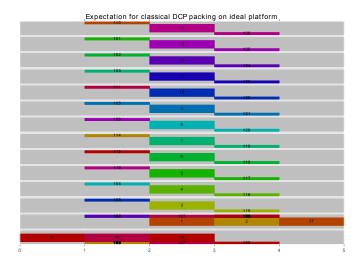
Different nodes w\ locality, worst case communication i→j

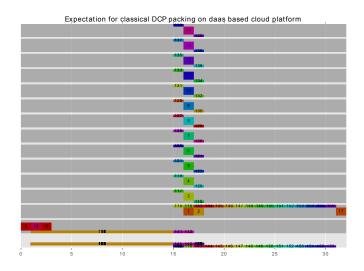


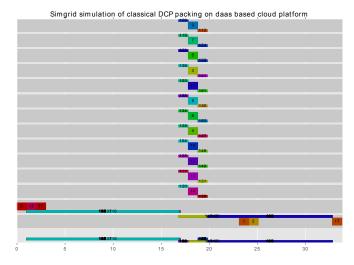
Communication aware task scheduling

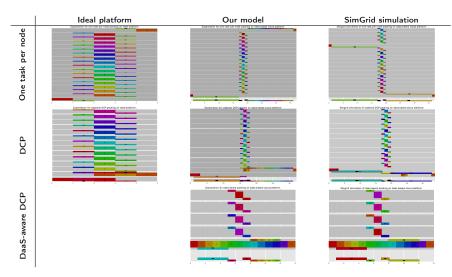


Preview of the critical path computation taking the machine network availability into account in DaaS-based platform.

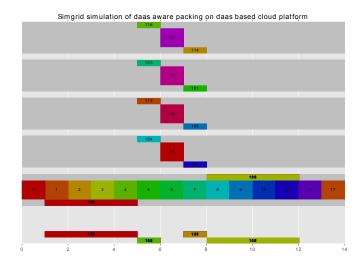








Comparison of the different clustering policies (Gantt charts and their associated makespan) for a multiple data fork-join DAG (n = 16).



Makespan Cost DAG Algorithm #Nodes (t)nodes One task per node 18 22.024 67,204 18.000 18.000 Single node Single Data Fork-join DCP 14 18.024 56.168 DaaS aware DCP 2 13.012 20.012 One task per node 18 37.024 82.204 18,000 Single node 18.000 Multiple Data Fork-join DCP 14 33.803 70.156 DaaS aware DCP 5 14.000 26.048

Cost and makespan details of the different clustering policies for single data or multiple data fork-join DAG (n = 16).

Results

## Conclusion

- Network infrastructures and communication pattern are key;
- Legacy algorithm is not necessarily bad and can be updated;
- Simulations of the resulting clustering gives better results;
- We keep the advantages of DCP: can be applied to any DAG.

### Future work

Regarding task clustering and Clouds:

Deciding which instance to use;

#### Broader:

- Designing the rest of our framework;
- Implementing everything in a user-friendly platform manager;
- Designing extra features: QoS description, instance selection,

. . .

Thank you

Thank you for your attention.

Do you have any question ?