On the complexity of scheduling checkpoints for computational workflows

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## Motivation

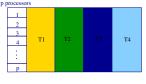
#### Framework

- Application task graph, a DAG where nodes represent tasks and edges correspond to dependences between them.
- Application DAG to be executed on a failure-prone platform of p identical processors.

## Motivation

#### Framework

- Application task graph, a DAG where nodes represent tasks and edges correspond to dependences between them.
- Application DAG to be executed on a failure-prone platform of p identical processors.
- Each task is executed in parallel on the p processors.





• Resilience provided through coordinated checkpointing.

#### Objective : Minimizing the expectation of the total execution time.

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- In which order should we execute the tasks?
- At the end of the execution of each task Ti, should we perform a checkpoint or should we proceed directly with the computation of another task?



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- Checkpointing strategies for computational tasks with linear chains (with single processor).
- They cannot aim at minimizing the expected execution time.
- Maximizing the amount of work done before the first failure.

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- Maximizing the amount of work done before the first failure.

We solve the original problem that is minimizing the expected execution time (At least for Exponential failures)



#### Expected time needed to execute a work and to checkpoint it

2 Complexity of the general scheduling problem

Oynamic Programming algorithm for linear chains





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4 Conclusion and extensions



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## Hypothesis

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- $\mathcal{W}$  : Duration of Work
- C: Checkpoint cost
- *D* : Downtime (hardware replacement by spare, or software rejuvenation via rebooting)
- R: Recovery cost after failure

• Recursive Approach :

 $\mathbb{E}(T(\mathcal{W},C,R)) =$ 

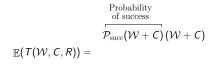
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 $\mathcal{P}_{succ}(\mathcal{W} + C) \underbrace{(\mathcal{W} + C)}_{F}$  $\mathbb{E}(T(\mathcal{W}, C, R)) =$ 

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$$\begin{split} & \mathcal{P}_{\text{succ}}(\mathcal{W}+\mathcal{C})\left(\mathcal{W}+\mathcal{C}\right) \\ & \mathbb{E}(\mathcal{T}(\mathcal{W},\mathcal{C},\mathcal{R})) = & + \\ & \left(1-\mathcal{P}_{\text{succ}}(\mathcal{W}+\mathcal{C})\right)\left(\mathbb{E}(\mathcal{T}_{\textit{lost}}(\mathcal{W}+\mathcal{C})) + \mathbb{E}(\mathcal{T}_{\textit{rec}}) + \mathbb{E}(\mathcal{T}(\mathcal{W},\mathcal{C},\mathcal{R}))\right) \end{split}$$

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$$\begin{aligned} \mathcal{P}_{\text{succ}}(\mathcal{W} + \mathcal{C}) \left( \mathcal{W} + \mathcal{C} \right) \\ \mathbb{E}(\mathcal{T}(\mathcal{W}, \mathcal{C}, \mathcal{R})) &= + \\ & \left( 1 - \mathcal{P}_{\text{succ}}(\mathcal{W} + \mathcal{C}) \right) \underbrace{\left( \mathbb{E}(\mathcal{T}_{\textit{lost}}(\mathcal{W} + \mathcal{C})) + \mathbb{E}(\mathcal{T}_{\textit{rec}}) + \mathbb{E}(\mathcal{T}(\mathcal{W}, \mathcal{C}, \mathcal{R})) \right)}_{\text{Time elapsed before the failure occurred}} \end{aligned}$$

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$$\mathcal{P}_{succ}(\mathcal{W} + C) (\mathcal{W} + C)$$

$$\mathbb{E}(T(\mathcal{W}, C, R)) = + (1 - \mathcal{P}_{succ}(\mathcal{W} + C)) (\mathbb{E}(T_{lost}(\mathcal{W} + C)) + \mathbb{E}(T_{rec}) + \mathbb{E}(T(\mathcal{W}, C, R)))$$

$$Time needed to compute W from scratch$$

• Recursive Approach :

$$\mathbb{E}(T(\mathcal{W}, C, R)) = \begin{array}{l} \mathcal{P}_{succ}(\mathcal{W} + C) (\mathcal{W} + C) \\ + \\ \underbrace{(1 - \mathcal{P}_{succ}(\mathcal{W} + C)) (\mathbb{E}(T_{lost}(\mathcal{W} + C)) + \mathbb{E}(T_{rec}) + \mathbb{E}(T(\mathcal{W}, C, R)))}_{Probability of failure} \end{array}$$

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# Computation of $\mathbb{E}(\mathcal{T}(\mathcal{W}, C, R))$

$$\mathbb{E}(T(\mathcal{W}, C, R)) = \mathbb{P}_{suc}(\mathcal{W} + C)(\mathcal{W} + C)$$

 $+(1-\mathbb{P}_{suc}(\mathcal{W}+C))\left[\mathbb{E}(T_{lost}(\mathcal{W}+C))+E(T_{rec})+\mathbb{E}(T(\mathcal{W},C,R))\right]$ 

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With an exponential failure distribution, we have :

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$$\mathbb{E}(T_{lost}(W+C)) = \int_0^\infty x \mathbb{P}(X = x | X < W + C) dx$$
  
 $\mathbb{E}(T_{lost}(W+C)) = \frac{1}{\lambda} - \frac{W+C}{e^{\lambda(W+C)}-1}$ 

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•  $\mathbb{E}(T_{rec}) = e^{-\lambda R} (D+R) + (1-e^{-\lambda R}) (D+\mathbb{E}(T_{lost}(R)) + \mathbb{E}(T_{rec}))$ 

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 $\mathbb{E}(T(\mathcal{W}, C, R)) = e^{\lambda R} \left(\frac{1}{\lambda} + D\right) \left(e^{\lambda(\mathcal{W}+C)} - 1\right)$ 



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### Problem statement

The general scheduling problem is :

• Given a time bound *K*, can we find an ordering for the execution of several independent tasks, and decide after which tasks to checkpoint, so that the expected execution time does not exceed *K*?

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#### Proposition

Consider n independent tasks,  $T_1, ..., T_n$ , with task  $T_i$  of duration  $W_i$  for  $1 \le i \le n$ . All checkpoint and recovery times are equal to C, and there is no downtime (D = 0). The problem to schedule these tasks, and to decide after which tasks to checkpoint, so as to minimize the expected execution time, is NP-complete in the strong sense.

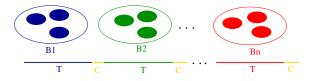
We use a reduction from 3-PARTITION, which is NP-complete in the strong sense.

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- General instance  $\mathcal{I}_1$  of 3-PARTITION:
  - 3n integers a<sub>1</sub>,..., a<sub>3n</sub>;
    ∑<sub>1≤j≤3n</sub> a<sub>j</sub> = nT, and T/4 < a<sub>j</sub> < T/2 for 1 ≤ j ≤ 3n,</li>
    Does there exist a partition in n subsets B<sub>1</sub>,..., B<sub>n</sub> of {a<sub>1</sub>,..., a<sub>3n</sub>} such that for all 1 ≤ i ≤ n, ∑<sub>a<sub>i</sub>∈B<sub>i</sub></sub> a<sub>j</sub> = T.

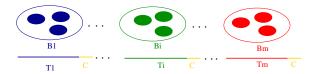
Note that necessarily in any solution, each  $B_i$  has cardinal 3.



## Proof of NP-completeness

• Instance  $\mathcal{I}_2$  of our problem :

- 3n independent tasks: task<sub>1</sub>, ..., task<sub>3n</sub>, task<sub>i</sub> being of size W<sub>i</sub> = a<sub>i</sub>.
- Does there exist a partition in m subsets  $B_1, \ldots, B_m$
- $\sum_{i=1}^{m} T_i = nT$  and *m* checkpoints.
- We let:  $\lambda = \frac{1}{2T}$ ,  $C = R = \frac{1}{\lambda}(\ln(2) \frac{1}{2})$ , and D = 0,  $K = n \frac{e^{\lambda C}}{\lambda} (e^{\lambda(T+C)} 1)$ .





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We want to compute the optimal expected execution time, that is:

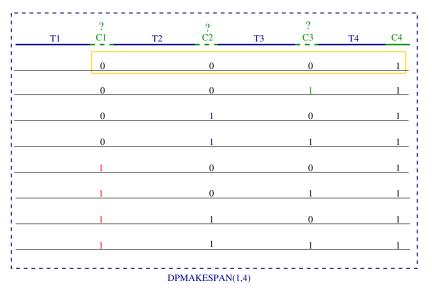
• the expectation *E* of the time needed to process all the tasks of an applications whose DAG is a linear chain.

Problem:

• Decide whether to checkpoint or not after the completion of each given task.



# Dynamic programming



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#### **Algorithm 1:** DPMAKESPAN(x, n)

$$\begin{aligned} & \text{if } x = n \text{ then} \\ & | \text{ return } (\mathbb{E}(T(\mathcal{W}_n, C_n, R_{n-1})), n) \\ & best \leftarrow \mathbb{E}(T(\sum_{i=x}^n \mathcal{W}_i, C_n, R_{x-1})) \\ & numTask \leftarrow n \\ & \text{for } j = x \text{ to } n - 1 \text{ do} \\ & \text{ (exp\_succ, num\_Task)} \leftarrow DPMAKESPAN(j+1, n) \\ & Cur \leftarrow exp\_succ \\ & + \mathbb{E}(T(\sum_{i=x}^j \mathcal{W}_i, C_j, R_{x-1})) \\ & \text{ if } Cur < best \text{ then} \\ & | best \leftarrow Cur \\ & numTask \leftarrow j \\ & \text{return } (best, numTask) \end{aligned}$$

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## Linear complexity

#### Proposition

Algorithm1 provides the optimal solution for a linear chain of n tasks. Its complexity is  $O(n^2)$ .

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### Extensions

- General model of checkpointing costs:
  - The checkpoint after a task  $T_i$  may depend on  $T_i$  and on some other tasks that have been executed since the last checkpoint.
- Alleviating the *full parallelism* assumption:
  - Variable parallelism.
  - Ressource allocation problem.
- Using general failure laws than Exponential distributions:
  - First difficulty: Approximating the failure distribution of a platform of p processors.
  - Second difficulty: Estimating the expected execution time of a work W.

Important results:

- Closed-form formula for the expected execution time of a computational workflows followed by its checkpoint (using Exponential failure distribution).
- The strong NP-hardness of the problem for independent tasks and constant checkpoint costs.

• Dynamic programming algorithm for linear chains of tasks with arbitrary checkpoint costs.

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