# A nice little scheduling problem

### Yves Robert Ecole Normale Supérieure de Lyon & Institut Universitaire de France

CCGSC'2010 Asheville

### A few nice little scheduling problems

- I made it to the 10 CCGSC workshops!
- I talked about a nice little scheduling problem in 1992
- I talked about a nice little scheduling problem in 1994
- I talked about a nice little scheduling problem in 1996
- I talked about a nice little scheduling problem in 1998
- I talked about a nice little scheduling problem in 2000
- I talked about a nice little scheduling problem in 2002
- I talked about a nice little scheduling problem in 2004
- I talked about a nice little scheduling problem in 2006
- I talked about a nice little scheduling problem in 2008



Sequential jobs

Parallel jobs

### A ferrinice little scheduling problems



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- I talked about a nice attle scheduling coblem in 1996
- I talked about zerice little scheduling protein in 1998
- I talked abe a nice little scheduling problem in 2000
- I talked At last

**J**ked

- a fundamental problem
- talked in exascale computing!!

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2001

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308

**N**4

Checkpointing versus Migration for Post-Petascale Machines

### Franck Cappello INRIA-Illinois Joint Laboratory for Petascale Computing

Henri Casanova University of Hawaiʻi

Yves Robert Ecole Normale Supérieure de Lyon & Institut Universitaire de France

#### CCGSC'2010 Asheville

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• Fault tolerant computing becomes **unavoidable** Caveat: same story told for a very long time! 😳

• Coming for real on future machines, e.g. **Blue Waters** INRIA-Illinois Joint Laboratory for Petascale Computing

- Techniques:
  - failure avoidance (as opposed to failure tolerance)
  - checkpointing, migration

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Framework	Sequential jobs	Parallel jobs	Results	No prediction
Outline				



- 2 Sequential jobs
- 3 Parallel jobs
- 4 Numerical results
- 5 To predict or not to predict

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Framework	Sequential jobs	Parallel jobs	Results	No prediction
Outline				

1 Framework

- 2 Sequential jobs
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(B)

 Framework
 Sequential jobs
 Parallel jobs
 Results
 No prediction

- Applications will face resource faults during execution
- Failure prediction available (e.g. alarm when a disk or CPU becomes unusually hot)
- Application must dynamically prepare for, and recover from, expected failures
- Compare two well-known strategies:
  - Checkpointing: purely local, but can be very costly
  - Migration: requires availability of a spare resource

• Remember, we assume accurate failure prediction

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- Applications will face resource faults during execution
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- D: length of downtime intervals
- $\mu$ : (average) length of execution intervals, a.k.a. MTTF
  - R: recovery time (beginning of interval)
  - C: checkpoint time (end of interval, just before failure)





- D: length of downtime intervals
- $\mu$ : (average) length of execution intervals
  - *M*: migration time (end of interval, just before failure)
  - Need spare node 😟

Framework	Sequential jobs	Parallel jobs	Results	No prediction
Notations				

- C: checkpoint save time (in minutes)
- R: checkpoint recovery time (in minutes)
- D: down/reboot time (in minutes)
- M: migration time (in minutes)
- μ: mean time to failure (e.g., 1/λ if failures are exponentially distributed)
- N: total number of cluster nodes
- *n*: number of spares (migration)



• Checkpointing/migration comparison makes sense only if

#### M < C + D + R

otherwise better use faulty machine as own spare

• *Live migration* without any disk access, thereby dramatically reducing migration time

Framework	Sequential jobs	Parallel jobs	Results	No prediction
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Probability of node being active

$$u_c = \max\left(0, \frac{\mu - R - C}{\mu + D}
ight)$$

#### **Global throughput**

$$\rho_c = u_c \times N = \max\left(0, \frac{\mu - R - C}{\mu + D}\right) \times N$$

∃ →

FrameworkSequential jobsParallel jobsResultsNo predictionMigration (1/2)



#### Probability of node being active

$$u_m = \max\left(0, \frac{\mu - M}{\mu + D}\right)$$

#### **Global throughput**

$$\rho_m = u_m \times (N - n) = \max\left(0, \frac{\mu - M}{\mu + D}\right) \times (N - n)$$

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FrameworkSequential jobsParallel jobsResultsNo predictionMigration (2/2)



No shortage of spare nodes?

$$success(n) = \sum_{k=0}^{n} {N \choose k} u_m^{N-k} (1-u_m)^k$$

- Find n = α(ε, N) that "guarantees" a successful execution with probability at least 1 − ε
- Solve numerically

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Number of processors required by typical jobs: *two-stage log-uniform distribution biased to powers of two* 

• Let  $N = 2^Z$  for simplicity

- Probability that a job is sequential:  $\alpha_0 = p_1 \approx 0.25$
- Otherwise, the job is parallel, and uses 2<sup>j</sup> processors with identical probability

$$\alpha_j = \alpha = (1 - p_1) \times \frac{1}{Z}$$

for  $1 \leq j \leq Z = \log_2 N$ 

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### • Steady-state utilization of whole platform:

- all processors always active
- constant proportion of jobs using any processor number
- Expectation of the number of jobs:
  - K total number of jobs running
  - $\beta_j$  jobs that use  $2^j$  processors exactly



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Framework	Sequential jobs	Parallel jobs	Results	No prediction
Distribution	(3/3)			

• Equations:

• 
$$K = \sum_{j=0}^{Z} \beta_j$$
  
•  $\beta_j = \alpha_j K$  for  $0 \le j \le Z$   
•  $\sum_{j=0}^{Z} 2^j \beta_j = N$ 

$$\frac{N}{K} = \sum_{j=0}^{Z} 2^{j} \alpha_{j} = p_{1} + \frac{1 - p_{1}}{Z} \sum_{j=1}^{Z} 2^{j} = p_{1} + \frac{1 - p_{1}}{Z} (2N - 2)$$

hence the value of K and the  $\beta_j$ 

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Framework	Sequential jobs	Parallel jobs	Results	No prediction
Failures				

- If a job uses two processors, what is the expected interval time between failures?
- $\mu_j$  mean of the minimum of  $2^j$  i.i.d. variables
- If the variables are exponentially distributed, with scale parameter  $\lambda,$  then

$$\mu_j = 1/(\lambda 2^j) = \mu/2^j$$

• If the variables are Weibull, with scale parameter  $\lambda$  and shape parameter a, then

$$\mu_j = \lambda \Gamma(1 + 1/(a2^j))$$

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Framework	Sequential jobs	Parallel jobs	Results	No prediction
Checkpointi	ng			

#### **Platform throughput**

$$\rho_{cp} = \sum_{j=0}^{Z} \beta_j \times 2^j \times \max\left(0, \frac{\mu_j - R - C}{\mu_j + D}\right)$$

For the exponential distribution:  $\mu_j = \mu/2^j$ 

Framework	Sequential jobs	Parallel jobs	Results	No prediction
Migration				

#### **Platform throughput**

$$\rho_{mp} = \left(\sum_{j=0}^{Z} \beta_j \times 2^j \times \max\left(0, \frac{\mu_j - M}{\mu_j + D}\right)\right) \times \frac{N - n}{N}$$

Probability of success: same as for independent jobs

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(B)

Framework	Sequential Jobs	Parallel Jobs	Results	No prediction
Scenarios				

- Understand the impact of checkpointing vs. migration
- All results are in percentage improvement of migration over checkpointing (negative or positive values)
- All results use the following values:
  - $\mu=1$  day, 1 week, 1 month, 1 year

• 
$$N = 2^{14}, 2^{17}, 2^{20}$$

• 
$$arepsilon=10^{-4}$$
,  $10^{-6}$ 

• Number of required spares in parentheses

# Scenario "today" – C = R = 10, D = 1, M = 0.33

Sequential Jobs		Parallel Jobs			
$\mu$	Ν	$arepsilon=10^4$	$arepsilon=10^6$	$arepsilon=10^4$	$arepsilon=10^{6}$
	2 <sup>14</sup>	1.19 (32)	1.16 (37)	3141.07 (32)	3140.08 (37)
1 day	2 <sup>17</sup>	1.26 (164)	1.25 (177)	3086.92 (164)	3086.61 (177)
	2 <sup>20</sup>	1.28 (1086)	1.28 (1119)	3033.16 (1086)	3033.07 (1119)
	2 <sup>14</sup>	0.14 (9)	0.12 (12)	3521.14 (9)	3520.47 (12)
1 week	2 <sup>17</sup>	0.17 (35)	0.16 (40)	3511.74 (35)	3511.61 (40)
	2 <sup>20</sup>	0.18 (184)	0.18 (198)	3501.72 (184)	3501.67 (198)
	2 <sup>14</sup>	0.02 (5)	0.00 (7)	1541.89 (5)	1541.69 (7)
1 month	2 <sup>17</sup>	0.04 (13)	0.03 (17)	3354.95 (13)	3354.84 (17)
	2 <sup>20</sup>	0.04 (55)	0.04 (63)	3352.86 (55)	3352.83 (63)
	2 <sup>14</sup>	-0.01 (2)	-0.01 (3)	69.22 (2)	69.21 (3)
1 year	2 <sup>17</sup>	0.00 (4)	-0.00 (6)	1037.00 (4)	1036.99 (6)
	2 <sup>20</sup>	0.00 (11)	0.00 (13)	3381.52 (11)	3381.52 (13)

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# Scenario "2011" – C = R = 5, D = 1, M = 0.33

Sequential Jobs		Parallel Jobs			
$\mu$	N	$arepsilon=10^4$	$arepsilon=10^{6}$	$arepsilon=10^4$	$arepsilon=10^{6}$
	2 <sup>14</sup>	0.48 (32)	0.45 (37)	1587.29 (32)	1586.78 (37)
1 day	2 <sup>17</sup>	0.55 (164)	0.54 (177)	1573.40 (164)	1573.24 (177)
	2 <sup>20</sup>	0.57 (1086)	0.57 (1119)	1558.96 (1086)	1558.91 (1119)
	2 <sup>14</sup>	0.04 (9)	0.02 (12)	1743.11 (9)	1742.77 (12)
1 week	2 <sup>17</sup>	0.07 (35)	0.07 (40)	1741.00 (35)	1740.93 (40)
	2 <sup>20</sup>	0.08 (184)	0.08 (198)	1738.54 (184)	1738.52 (198)
	2 <sup>14</sup>	-0.01 (5)	-0.02 (7)	734.36 (5)	734.26 (7)
1 month	2 <sup>17</sup>	0.01 (13)	0.01 (17)	1656.28 (13)	1656.23 (17)
	2 <sup>20</sup>	0.02 (55)	0.02 (63)	1655.80 (55)	1655.78 (63)
	2 <sup>14</sup>	-0.01 (2)	-0.02 (3)	25.16 (2)	25.15 (3)
1 year	2 <sup>17</sup>	-0.00 (4)	-0.00 (6)	477.62 (4)	477.61 (6)
	2 <sup>20</sup>	0.00 (11)	0.00 (13)	1668.73 (11)	1668.73 (13)

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Parallel jobs

## Scenario "2015" – C = 10R = 0.21, D = 0.25, M = 0.33

Sequential Jobs		Parallel Jobs			
$\mu$	N	$arepsilon=10^4$	$arepsilon=10^{6}$	$arepsilon=10^4$	$arepsilon=10^{6}$
	214	-0.12 (18)	-0.14 (22)	-27.96 (18)	-27.98 (22)
1 day	217	-0.07 (82)	-0.08 (91)	-27.92 (82)	-27.92 (91)
	220	-0.05 (501)	-0.06 (523)	-27.90 (501)	-27.90 (523)
	2 <sup>14</sup>	-0.04 (6)	-0.05 (8)	-13.14 (6)	-13.15 (8)
1 week	217	-0.02 (20)	-0.02 (24)	-29.07 (20)	-29.08 (24)
	2 <sup>20</sup>	-0.01 (91)	-0.01 (101)	-29.07 (91)	-29.07 (101)
	2 <sup>14</sup>	-0.02 (3)	-0.03 (5)	-2.63 (3)	-2.64 (5)
1 month	2 <sup>17</sup>	-0.01 (8)	-0.01 (11)	-30.74 (8)	-30.74 (11)
	2 <sup>20</sup>	-0.00 (30)	-0.00 (35)	-30.74 (30)	-30.74 (35)
	214	-0.01 (2)	-0.01 (2)	-0.22 (2)	-0.22 (2)
1 year	217	-0.00 (3)	-0.00 (4)	-1.69 (3)	-1.69 (4)
	2 <sup>20</sup>	-0.00 (7)	-0.00 (9)	-17.00 (7)	-17.00 (9)

Framework	Sequential jobs	Parallel jobs	Results	No prediction
Summary				

- Sequential jobs: comparable performance (within 2%)
- Parallel jobs, short term: prefer migration
- Parallel jobs, 2015: picture not so clear
- Good news for migration:
  - small number of spares
  - insensitive to target value of success probability

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(B)

### Checkpointing versus ... checkpointing

- No failure prediction available
- No more migration 😊
- Checkpoint periodically
- How to determine optimal period T?
- Impact on platform throughput?

Sequential jobs Framework Results No prediction Optimal period T (1/3)

W = expected percentage of time lost, or "wasted":

$$W = \frac{C}{T} + \frac{T}{2\mu} \tag{1}$$

• First term in (1) by definition: C time-steps devoted to checkpointing every T time-steps

• Every  $\mu$  time-steps, a failure occurs  $\Rightarrow$  loss of T/2 time-steps in average

$$W_{min} = \sqrt{\frac{2C}{\mu}}$$

Sequential jobs Framework Results No prediction

Optimal period T (1/3)

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W minimized for  $T_{opt} = \sqrt{2C\mu}$  (Young's approximation)

$$W_{min} = \sqrt{\frac{2C}{\mu}}$$

FrameworkSequential jobsParallel jobsResultsNo predictionOptimal period T (2/3)

$$W = \frac{C}{T} + \frac{\frac{T}{2} + R + D}{\mu}$$
$$W_{min} = \frac{R + D}{\mu} + \sqrt{\frac{2C}{\mu}}$$

Different from Daly: target = steady-state operation of platform target  $\neq$  minimizing expected duration of a given job

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FrameworkSequential jobsParallel jobsResultsNo predictionOptimal period T (3/3)

$$W_{min} = \frac{R+D}{\mu} + \sqrt{\frac{2C}{\mu}}$$
(2)

 $W_{min}$  larger than 1 for very small  $\mu$ (likely to happen with jobs requiring many processors)

 $W_{min} \leq 1$  iff  $\mu \geq 1/
u_b^2$ , where $u_b = rac{-\sqrt{2C} + \sqrt{2C + 4(R+D)}}{2(R+D)}$ 

 $W_{min}^* = \min(W_{min}, 1)$ 

 Framework
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 Platform throughput

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# Sequential jobs

$$\rho = (1 - W^*_{min})N$$

#### **Parallel jobs**

$$\rho = \sum_{j=0}^{Z} (1 - W_{\min}^*(j)) 2^j \beta_j$$

use  $\mu_j$  instead of  $\mu$  in (2) to derive  $W^*_{min}(j)$ 

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Framework

# Numerical results: yield $\rho/N$ for scenario "2015"

	$\mu=1$ month			
N	per. chkpt.	prev. chkpt.	prev. mig.	
2 <sup>8</sup>	96.04%	99.81%	98.99%	
211	88.23%	98.50%	98.04%	
214	62.28%	88.75%	86.41%	
2 <sup>17</sup>	10.66%	40.04%	27.73%	
2 <sup>20</sup>	1.33%	5.01%	3.47%	

	$\mu=1$ year			
N	per. chkpt.	prev. chkpt.	prev. mig.	
2 <sup>8</sup>	98.89%	99.98%	99.59%	
2 <sup>11</sup>	96.80%	99.88%	99.75%	
2 <sup>14</sup>	90.59%	99.01%	98.79%	
2 <sup>17</sup>	70.46%	92.41%	90.84%	
2 <sup>20</sup>	15.96%	54.77%	45.46%	

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Framework	Sequential jobs	Parallel jobs	Results	No prediction
l imiting id	ob size			

- MTTF  $\mu = 1$  year
- Exponentially distributed failures
- Scenario "2015"
- Tightly coupled parallel job with 2<sup>20</sup> nodes (whole platform)
- Experiences a failure every 0.5 minutes!
- Throughput close to 0 for both fault tolerance and fault avoidance 🙁

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Framework

# Yield $\rho/N$ for scenario "2015" and capped job sizes

	${\it N}=2^{20}$ , $\mu=1$ month			
max job size	per. chkpt.	prev. chkpt.	prev. mig.	
2 <sup>20</sup>	1.33%	5.01%	3.47%	
2 <sup>19</sup>	2.67%	10.01%	6.93%	
2 <sup>18</sup>	5.33%	20.02%	13.87%	
2 <sup>17</sup>	10.66%	40.04%	27.73%	
2 <sup>16</sup>	21.32%	63.07%	55.46%	
2 <sup>15</sup>	42.64%	79.04%	74.72%	

	$\mu=1$ year				
max job size	per. chkpt.	prev. chkpt.	prev. mig.		
2 <sup>20</sup>	15.96%	54.77%	45.65%		
2 <sup>19</sup>	31.92%	73.57%	68.13%		
2 <sup>18</sup>	55.59%	85.54%	82.56%		
2 <sup>17</sup>	70.46%	92.41%	90.84%		
2 <sup>16</sup>	80.05%	96.11%	95.30%		
2 <sup>15</sup>	86.36%	98.03%	97.62%		

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Framework	Sequential jobs	Parallel jobs	Results	No prediction
Conclusion				

- Short term: prefer preventive migration to preventive checkpointing
- Longer term: not so clear, but may prefer preventive checkpointing
- Long-term scenarios and very large scale platforms:
  - Poor scaling of non-prediction-based traditional fault tolerance
  - Even with perfect prediction, fault avoidance not much better
  - Necessary to cap job size to achieve reasonable throughput
- Simulator: http://navet.ics.hawaii.edu/~casanova/ software/resilience.tgz



- Software/hardware techniques to reduce checkpoint, recovery, migration times and to improve failure prediction
- "Self-fault-tolerant" algorithms (e.g. asynchronous iterative)

Ahum, don't you see it coming? ...
 ... a nice little scheduling problem! <sup>(2)</sup>
 multi-criteria throughput/energy/reliability
 add replication



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- Need combine all three approaches!



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