# An Elegant and Efficient Implementation of Russian Dolls Search for Variable Weighted CSP

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#### Variable Weighted CSP

#### Definition:

- Additive WCSP where only unary constraints (X<sub>i</sub>=1) are weighted
- maximize  $\Omega = \sum w_i X_i$
- Subject to:  $X_i \in \{0,1\}$  + selection constraints

#### Some VWCSPs:

- knapsack problems,
- «soft scheduling »
- prize-collecting TSP



#### Example: Select & Schedule

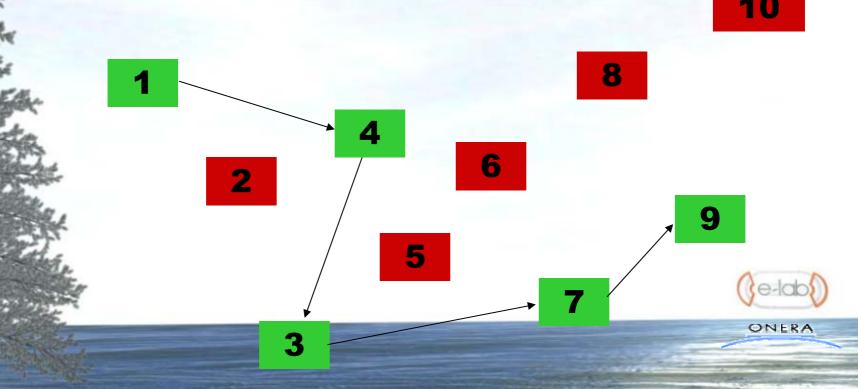
- 10 candidate photos, each with gain  $w_i$
- → Select the subset (selection variables X<sub>i</sub>)
  - $\Box$  of higher gain  $(\Omega = \sum w_i X_i)$
  - $\Box$  that can be scheduled (variables  $T_i$ ) without violating transition times  $t_{i\rightarrow i}$  constraints.



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- Static order on selection variables
- Successive resolutions of nested sub problems

 $Rds_{10}=max(P_{10})$ 

10

1

4

6

5

9

7



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- Static order on selection variables
- Successive resolutions of nested sub problems

 $Rds_9 = max(P_9)$ 

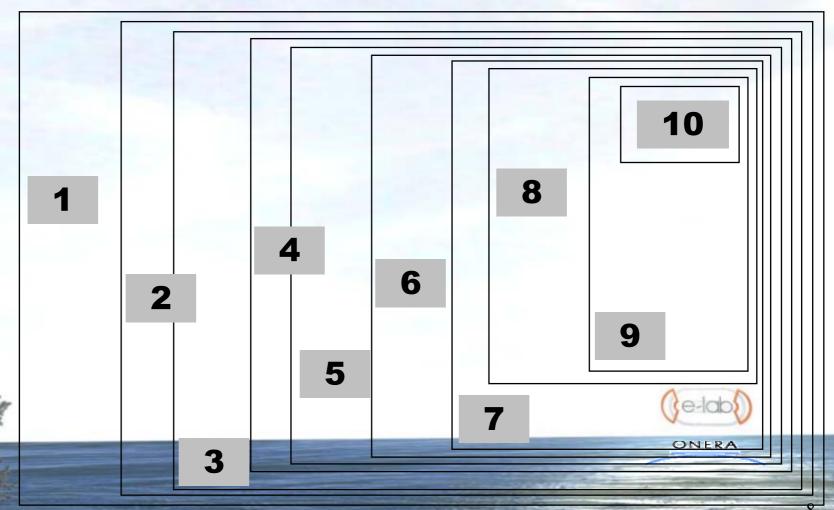


- Static order on selection variables
- Successive resolutions of nested sub problems

and so on... 10

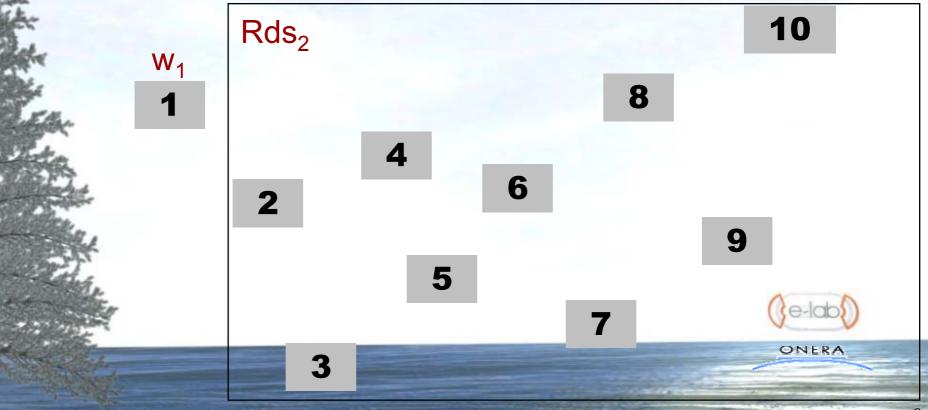
 $Rds_8 = max(P_8)$ 

 $Rds_1=max(P_1)=max(P)$ 

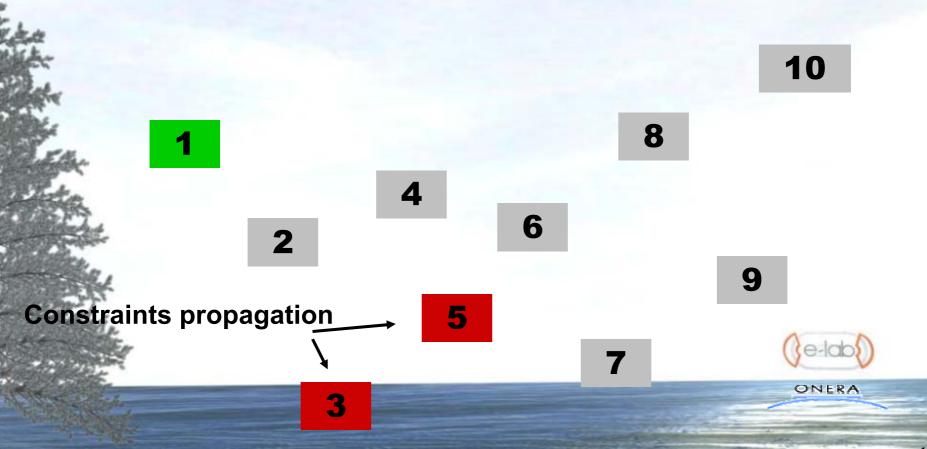


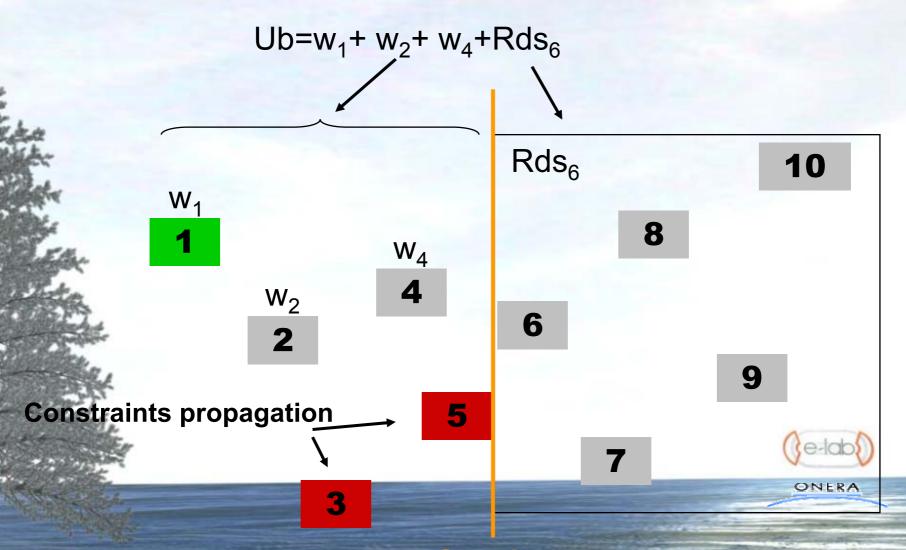
During resolution of the last problem  $(P_1)$  (once  $P_{10}, P_9, ..., P_2$  have been solved)

Initial upper bound:  $\Omega \leq w_1 + Rds_2$ 





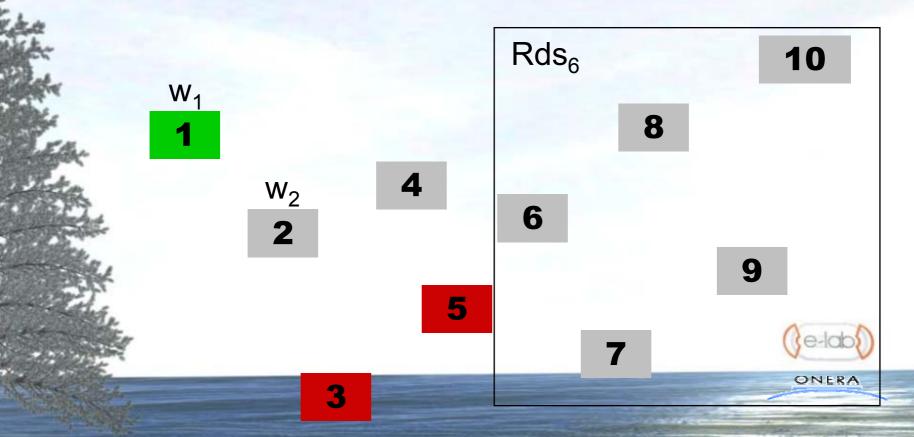




#### RDS cost-based filtering

#### Variable fixing rule:

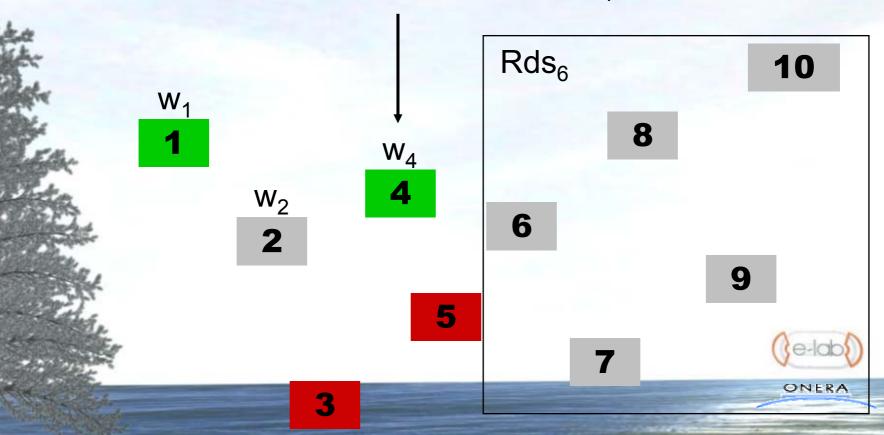
**If** w1+ w2+ Rds6 ≤ currentBest **then** X<sub>4</sub>=1



#### RDS cost-based filtering

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If w1+ w2+ Rds6  $\leq$  currentBest then  $X_4$ =1



#### LightRDS

- Objective function:
  - $\Omega = W_1 X_1 + ... + W_5 X_5 + \Omega_6$
  - $\Omega_6 = w_6 X_6 + ... + w_{10} X_{10}$  with  $\Omega_6 \le Rds_6$
- RDS filtering is naturally performed by these linear constraints

Declarative implementation

$$\Omega = \Omega_1$$

$$\Omega_i = \mathbf{w}_i \mathbf{X}_i + \Omega_{i+1}$$

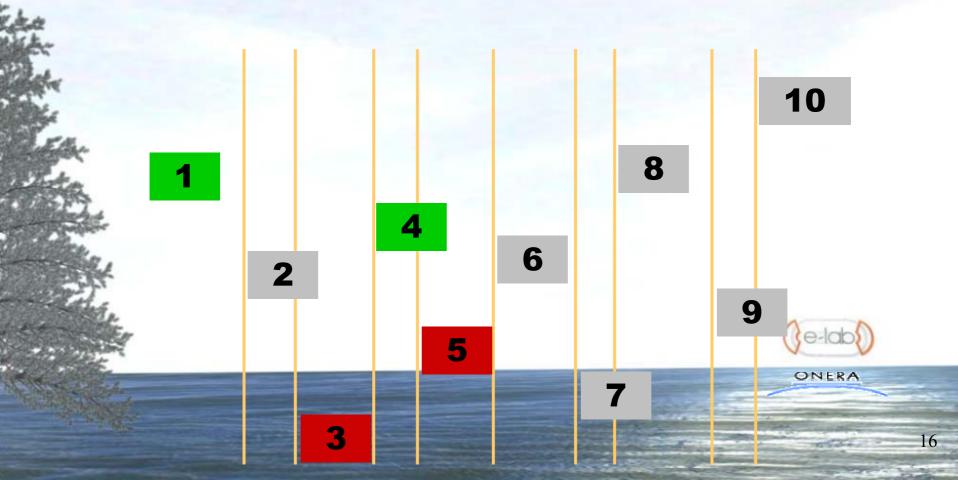
$$\Omega_{10} = \mathbf{W}_{10} \mathbf{X}_{10}$$



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# LightRDS filtering is strictly stronger

This is equivalent to using all frontiers simultaneously

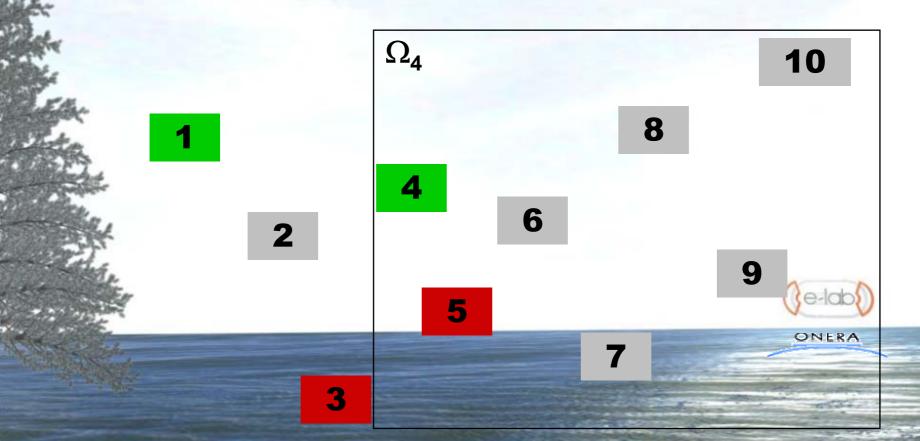


. dhill b

# LightRDS filtering is strictly stronger

If Rds<sub>4</sub><Rds<sub>6</sub>+w<sub>4</sub> then frontier 4 would produce a better bound:

$$w_1 + w_2 + Rds_4 < w_1 + w_2 + w_4 + Rds_6$$



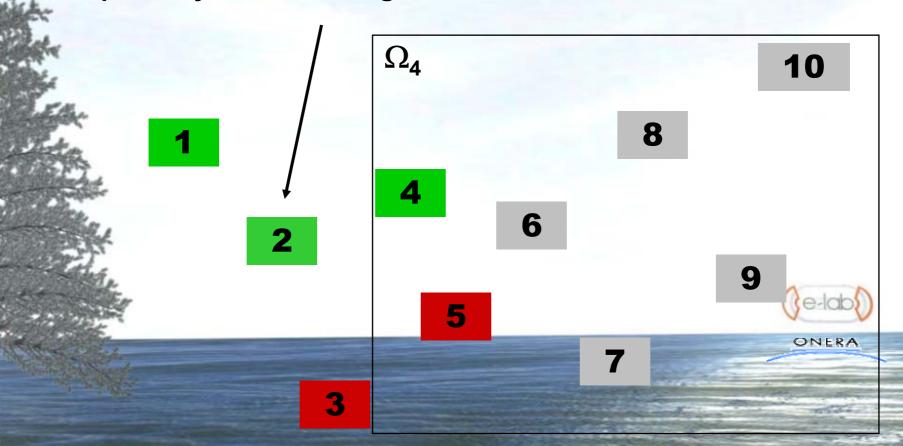
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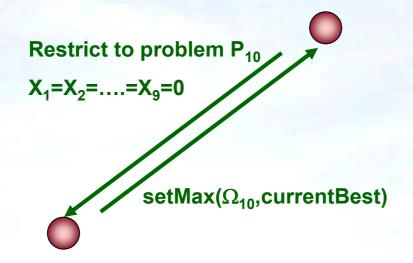
And possibly better filtering:



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#### **Control mechanism**

→ Encapsulated in a special root choice point

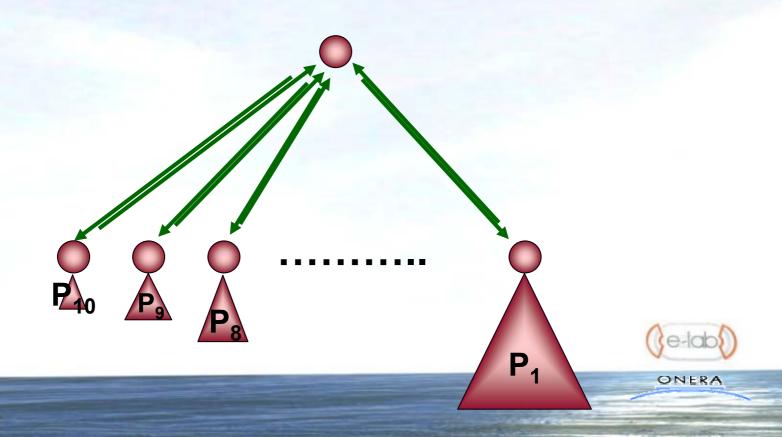




NIRO!

#### **Control mechanism**

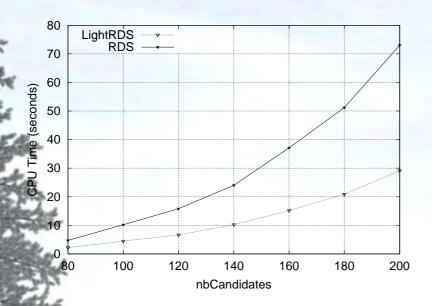
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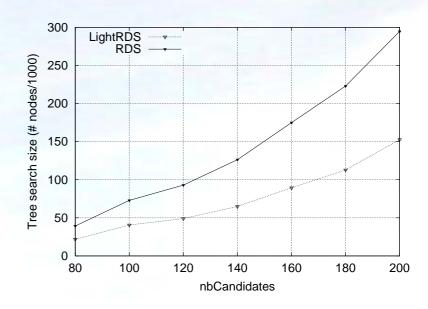


ROS

## **Computational Results**

- Satellite planning problem: ftp://ftp.cert.fr/pub/DCSD/CD/lemaitre/Choco/bep/
- CHOCO model

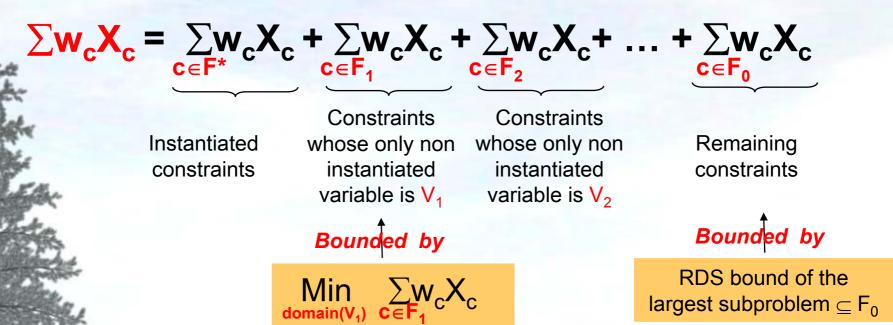




Number of nodes and CPU time are divided by two

#### **NOT** extensible to WCSP

Forward Checking (FC)





dynamic partition of constraints

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-> static reformulation of the objective function seems impossible

#### Conclusion

- LightRDS is more efficient
- LightRDS is simple:
  - Declarative implementation
  - No dedicated filtering to program
  - No frontier to manage
- LightRDS can be tested in a few minutes
  - on Variable Weighted CSPs
  - when the constraint graph has a small bandwidth

