

Project Scheduling with Alternative Technologies and Stochastic Activity Durations

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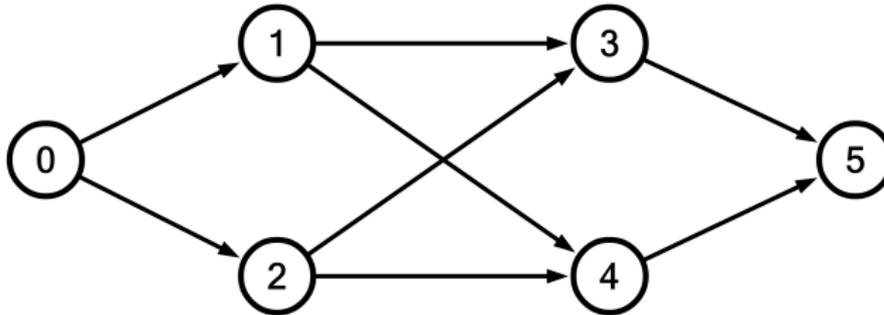
Introduction: Module Networks

Our goal is to maximize the NPV of projects in which:

- activities can fail,
 - activities that pursue the same result may be grouped in “modules”,
 - each module needs to be successful for the project to succeed,
 - a module is successful if at least one of its activities succeeds
- ⇒ not all activities in the network have to be started in order for the project to be successful,
- ⇒ upon failure of all activities in the module, the module fails, resulting in overall project failure.

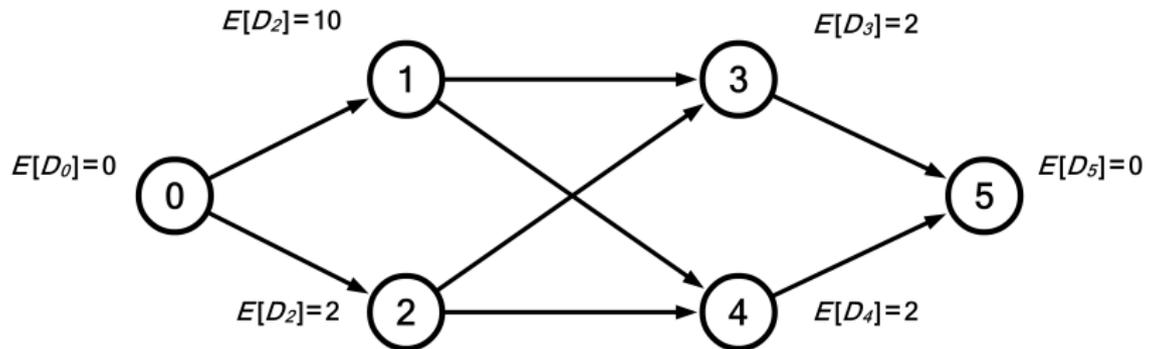
This is common in R&D (especially in NPD) but also in other sectors: pharmaceuticals, software development, fundraising ...

Example: Definitions



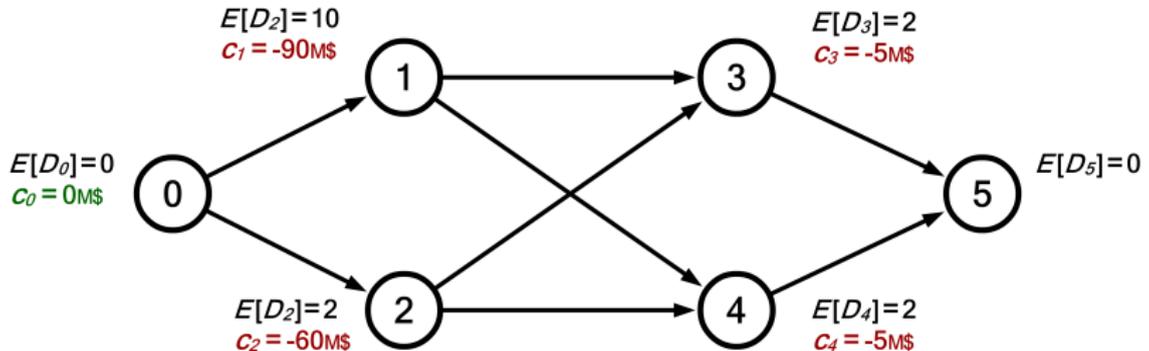
- (AON) project network with n activities

Example: Definitions



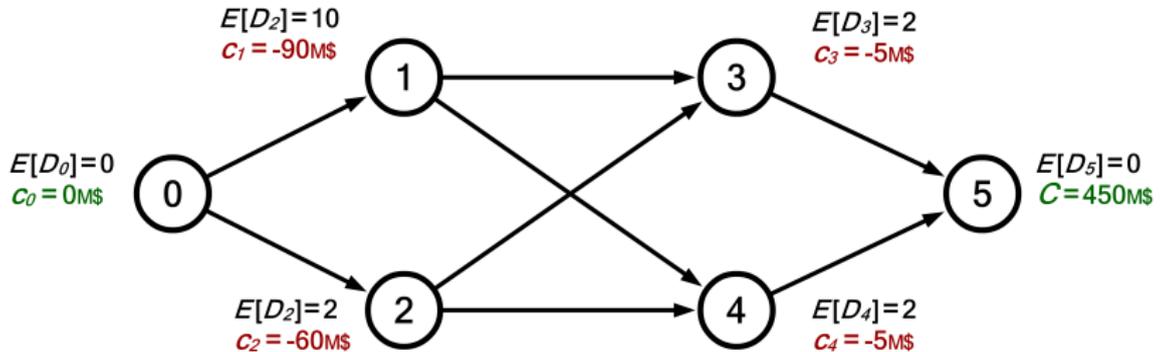
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 - Stochastic activity durations: expected duration $E[D_j]$ of activity j

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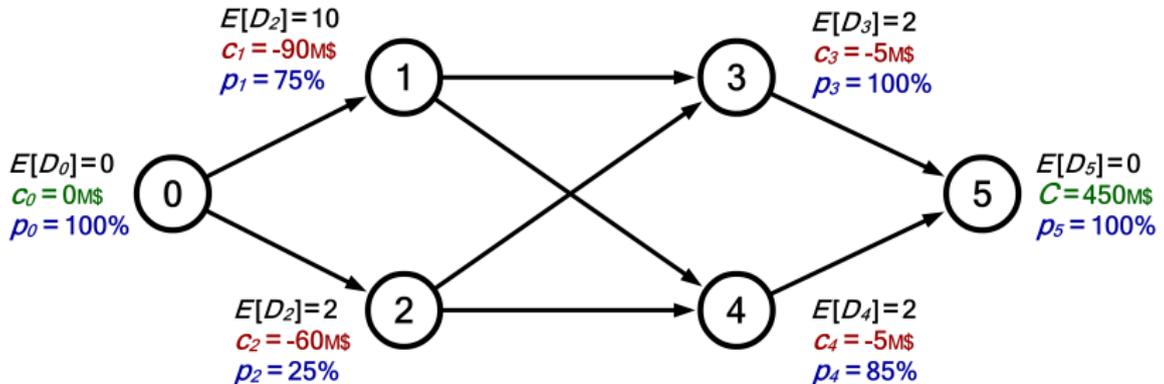
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- Expected NPV-objective: cash flow c_j is incurred at the start of activity j

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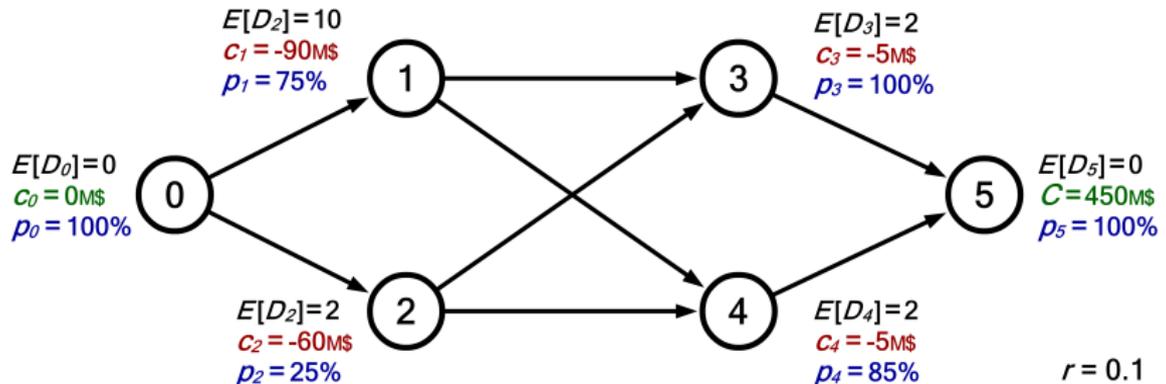
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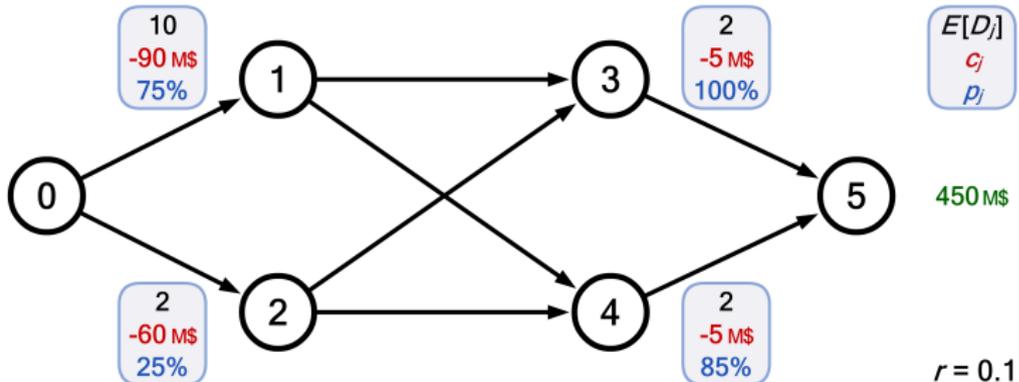
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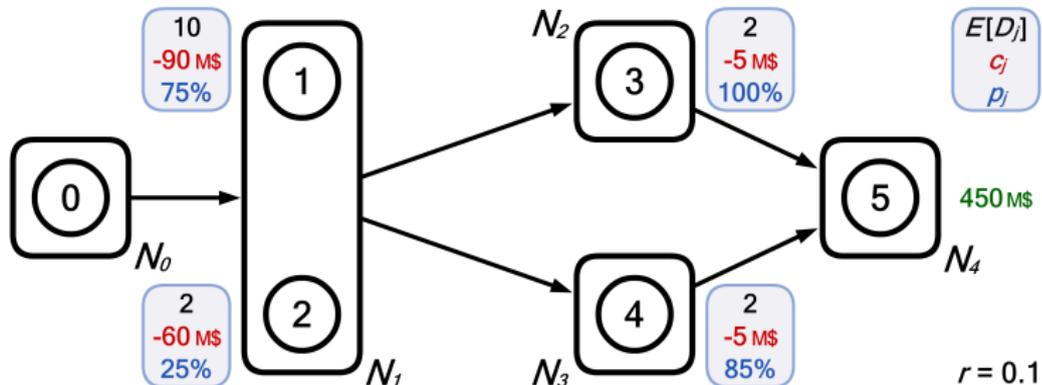
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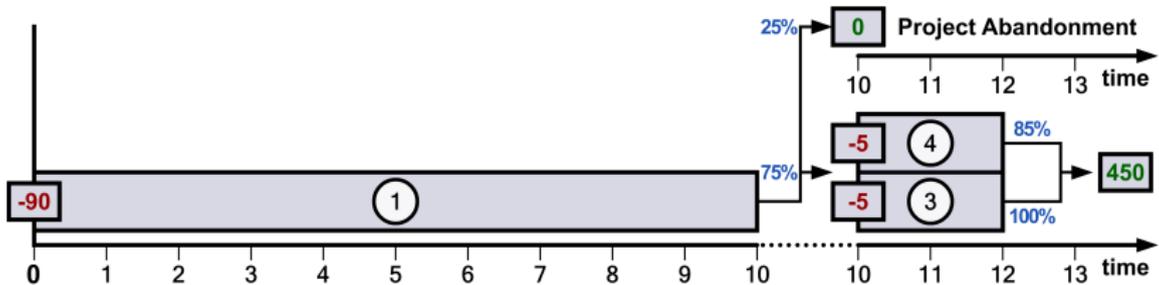


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- Time value of money \Rightarrow discount rate r
- m modules N_i

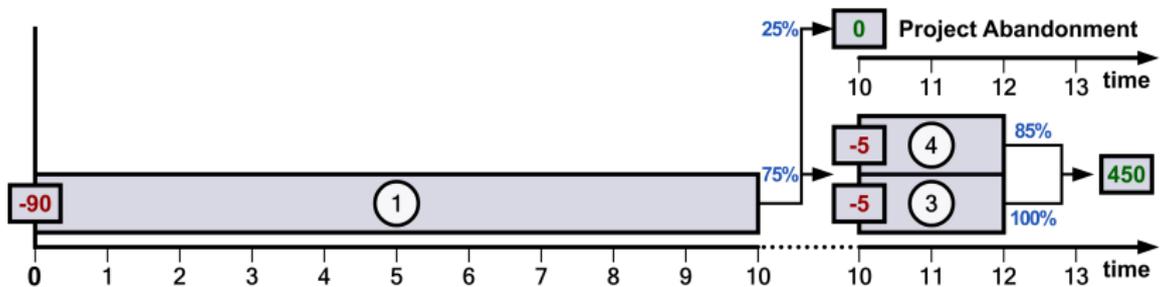
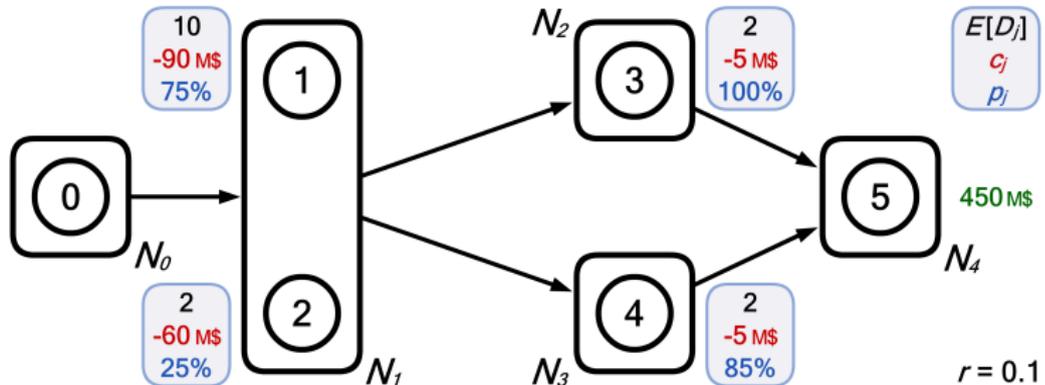
Example: Policy Π_1

A solution is not a schedule but rather a scheduling policy (even with deterministic durations)

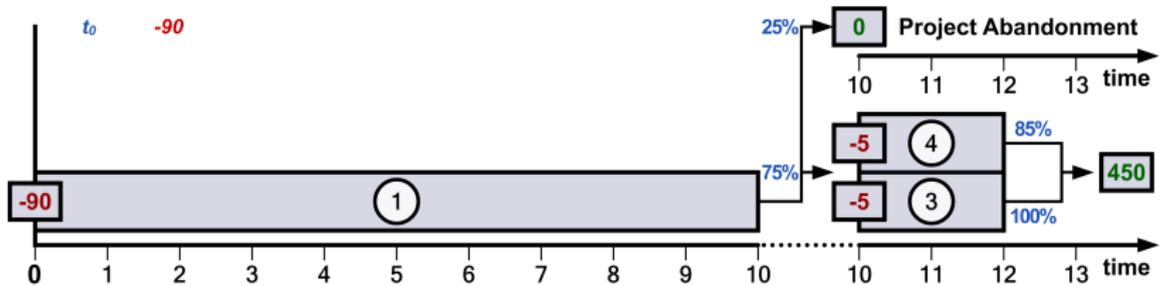
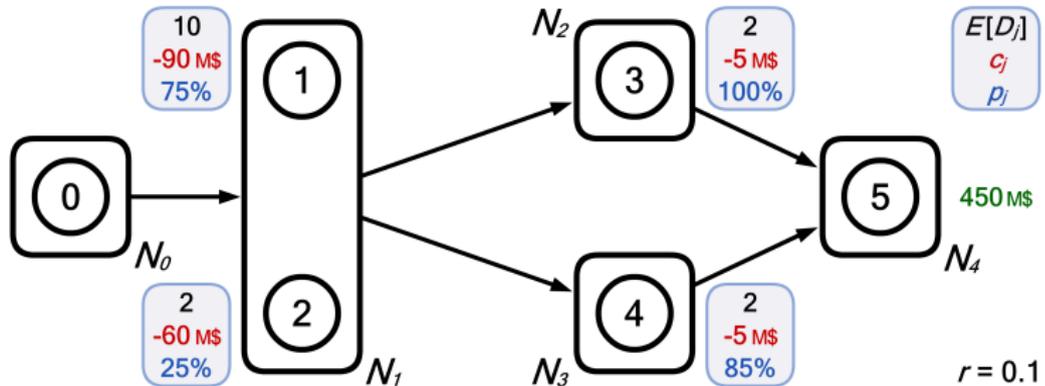
Policy Π_1 results in a NPV of **-6.35M\$** if activity durations are deterministic



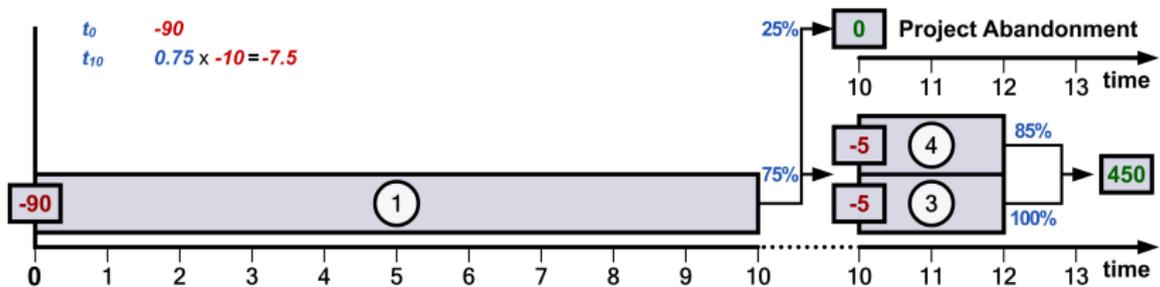
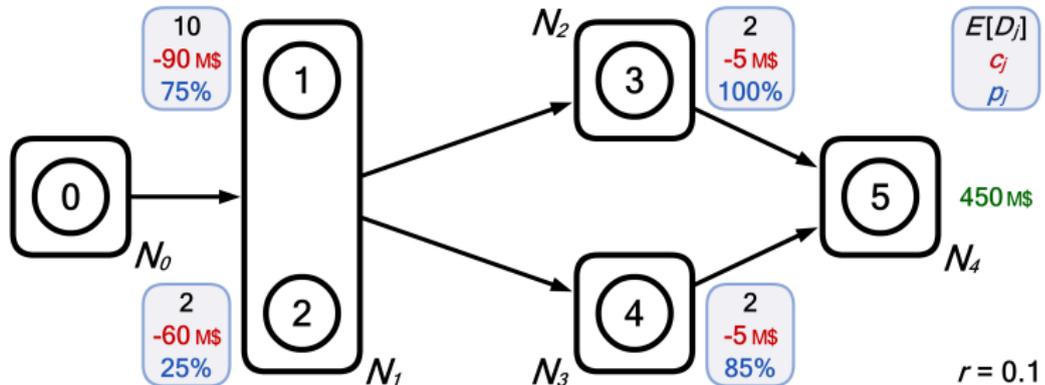
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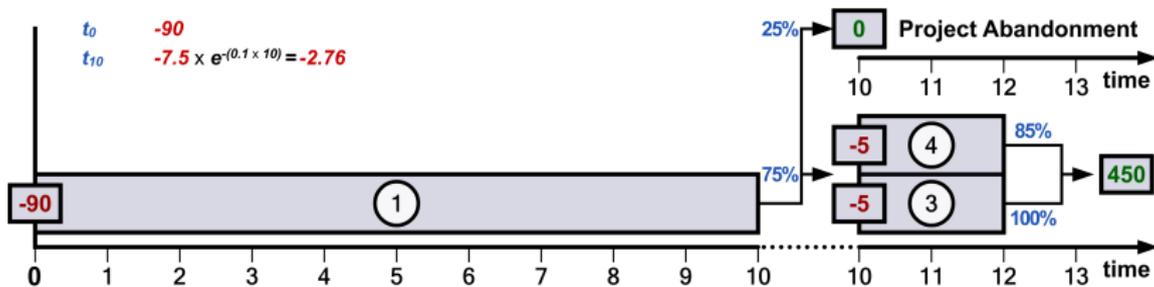
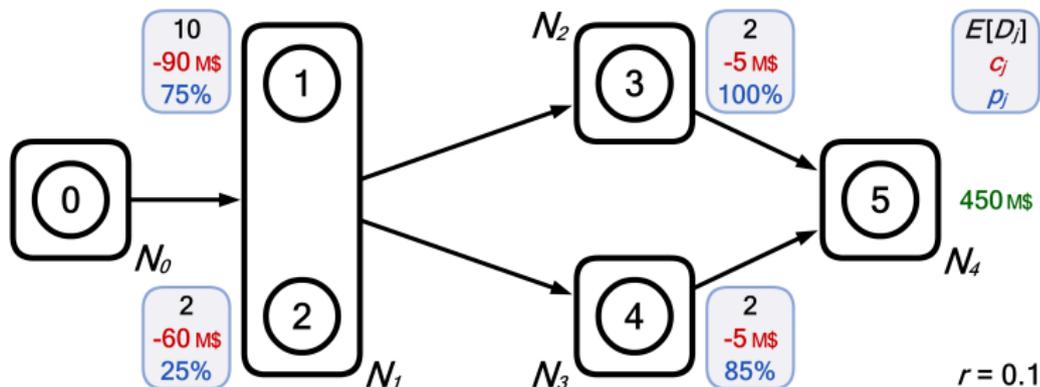
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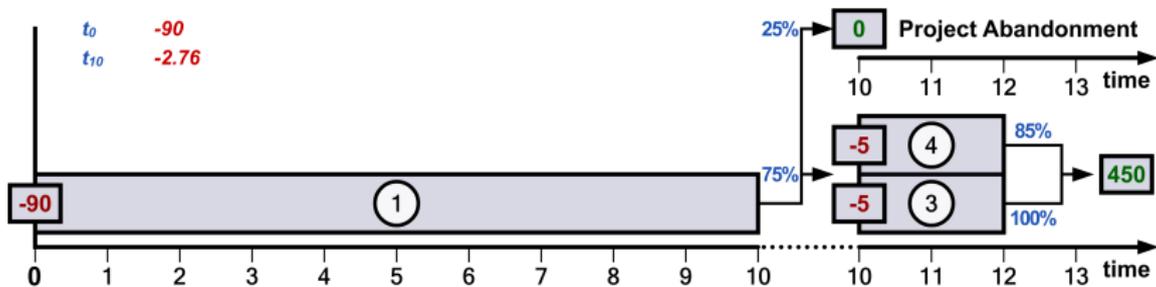
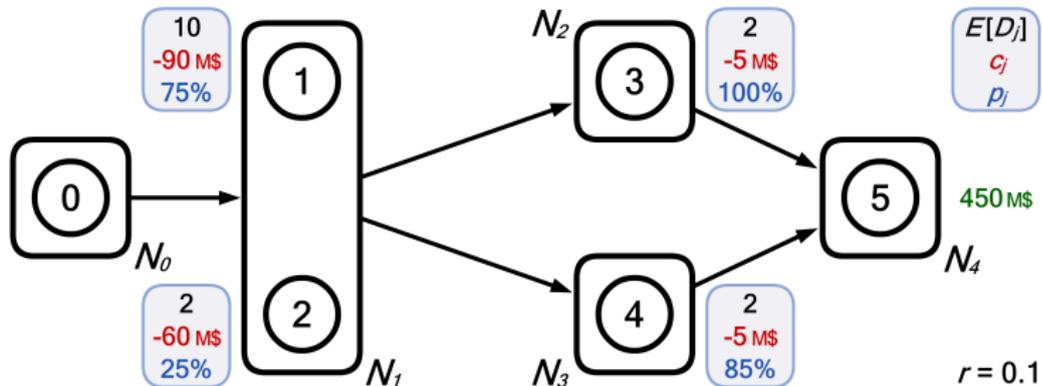
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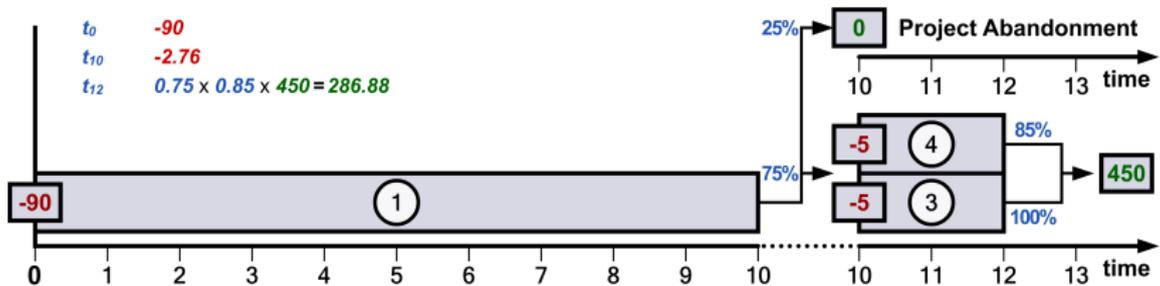
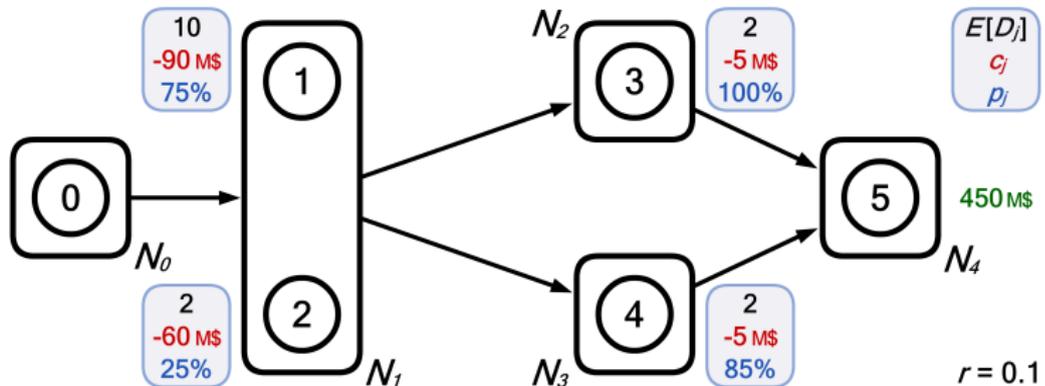
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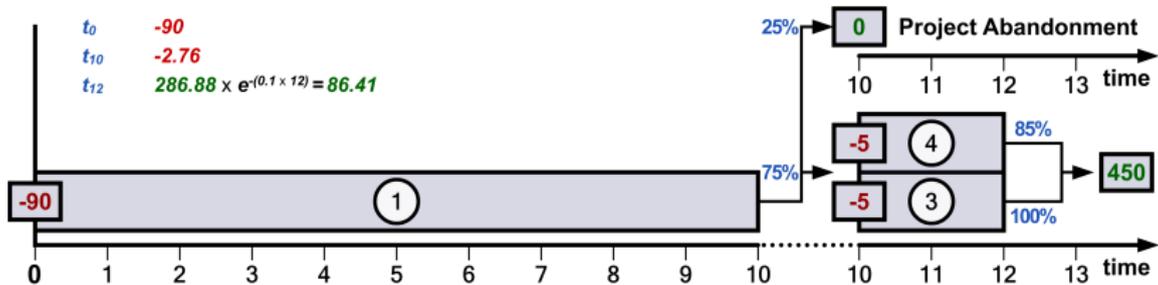
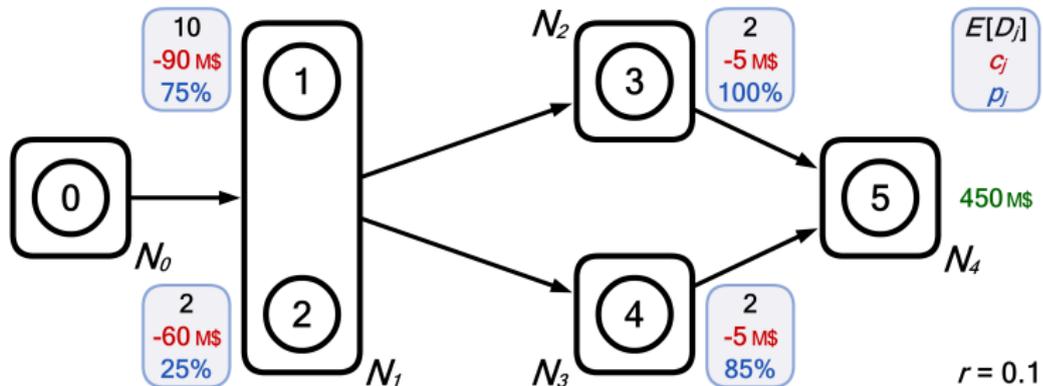
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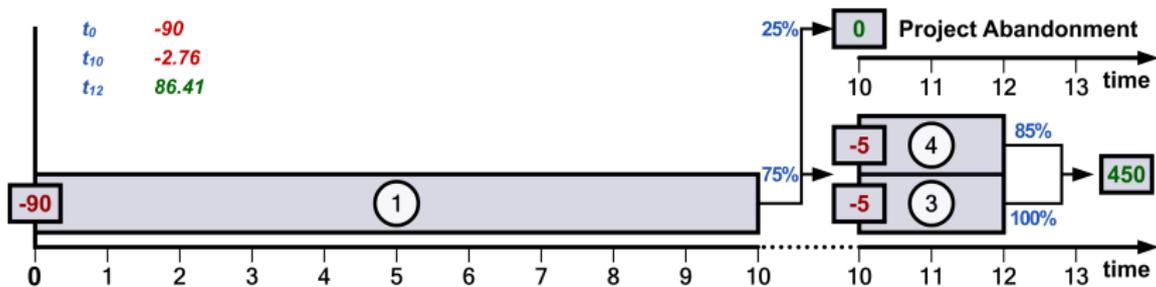
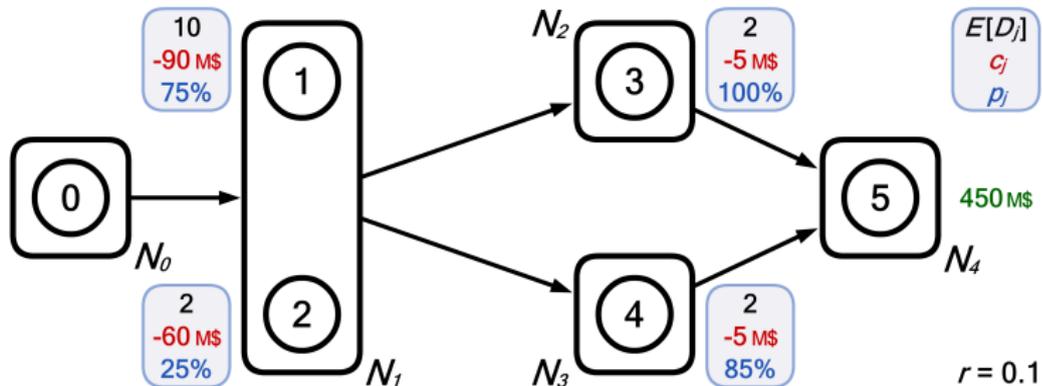
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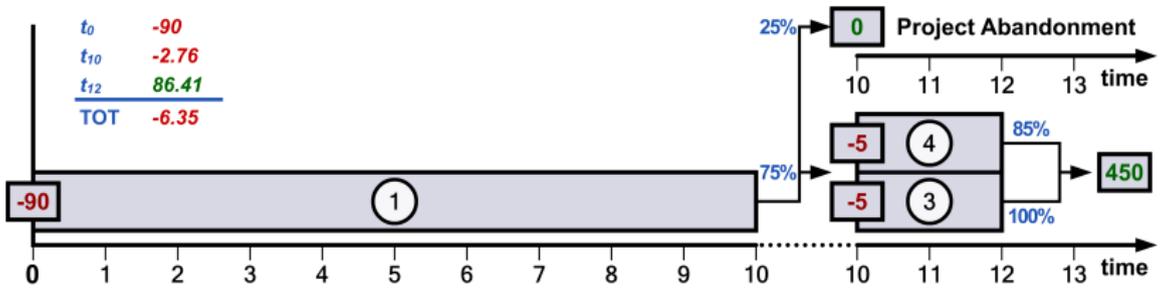
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A solution is not a schedule but rather a scheduling policy (even with deterministic durations)

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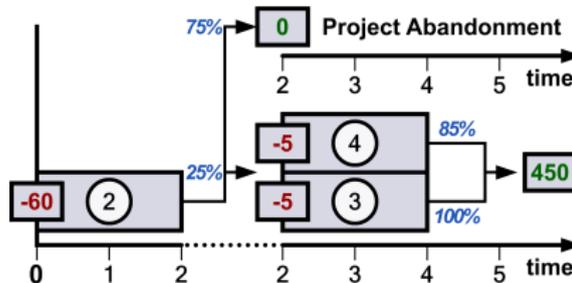
Example: Policy Π_2

A solution is not a schedule but rather a scheduling policy (even with deterministic durations)

Policy Π_1 results in a NPV of **-6.35M\$** if activity durations are deterministic

Policy Π_2 is optimal for deterministic durations and yields a NPV of **2.05M\$**

| | |
|-------|-------|
| t_0 | -60 |
| t_2 | -2.04 |
| t_4 | 64.10 |
| TOT | 2.05 |



Backward SDP-recursion: concepts & definitions

Exponentially distributed activity durations \Rightarrow use of a Continuous-Time Markov Chain (CTMC) to model the statespace.

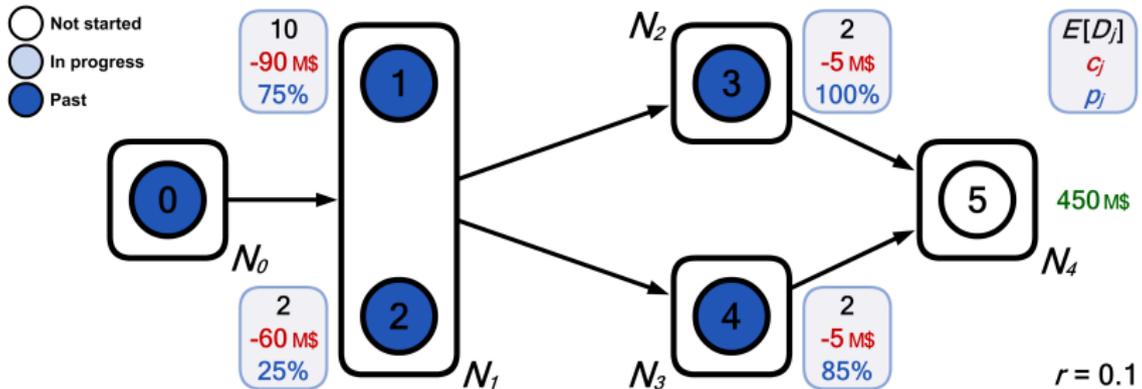
The state of an activity j at time t can be:

- $\Omega_j(t) = 0$: not started,
- $\Omega_j(t) = 1$: in progress,
- $\Omega_j(t) = 2$: past (successfully finished, failed or considered redundant because its module is completed).

The state of the system at a time instance t is given by vector $\mathbf{\Omega}(t) = \{\Omega_0(t), \dots, \Omega_n(t)\}$.

The size of the statespace has upper bound 3^n . Most states do not satisfy precedence constraints \Rightarrow a strict definition of the statespace is required and provided in Creemers et al. (2008).

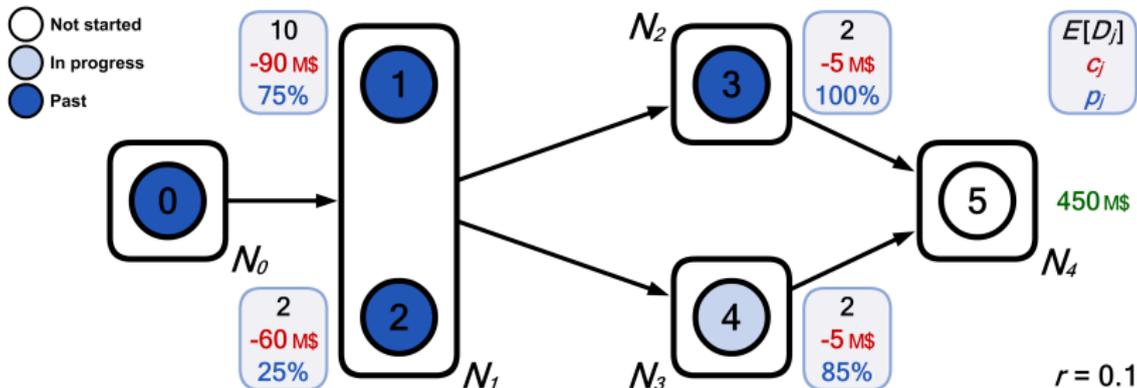
Example: Stochastic Durations



(2,2,2,2,2,0) [450M\$]

Project value upon entry of the final state = project payoff

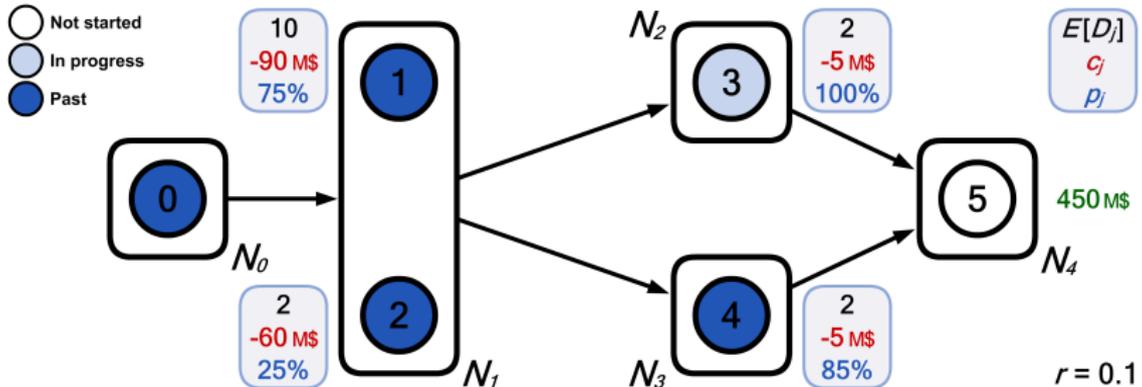
Example: Stochastic Durations



$(2,2,2,2,2,0)$ [450M\$]
 ↳ $(2,2,2,2,1,0)$ [318.75M\$]

Discount factor: $(1/D_j) \times (r + (1/D_j))^{-1}$
 $D_4 = 2 \Rightarrow$ discount factor = 0.83
 Discounted value upon state entry = 375
 $p_4 = 0.85 \Rightarrow$ NPV upon state entry = 318.75

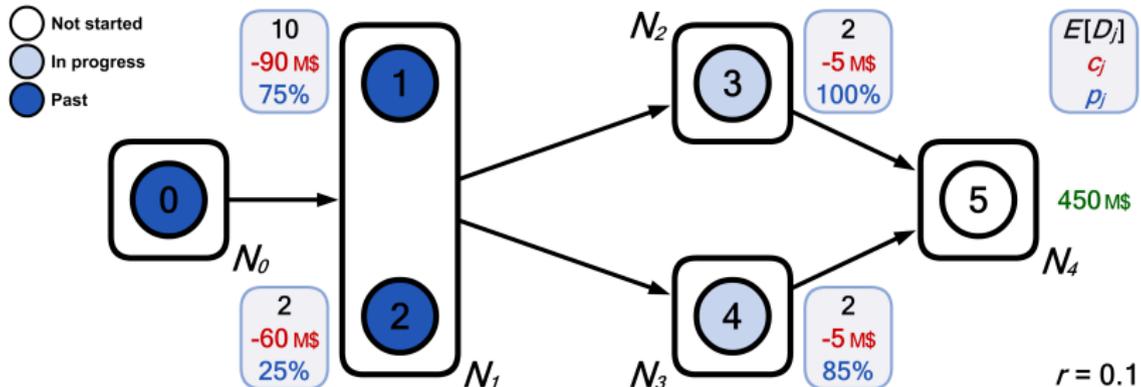
Example: Stochastic Durations



- (2,2,2,2,2,0) [450M\$]
- └─ (2,2,2,2,1,0) [318.75M\$]
- └─ (2,2,2,1,2,0) [375M\$]

Discount factor: $(1/D_j) \times (r + (1/D_j))^{-1}$
 $D_3 = 2 \Rightarrow$ discount factor = 0.83
 Discounted value upon state entry = 375
 $p_3 = 1.00 \Rightarrow$ NPV upon state entry = 375

Example: Stochastic Durations



(2,2,2,2,2,0) [450M\$]

└ (2,2,2,2,1,0) [318.75M\$]

└ (2,2,2,1,2,0) [375M\$]

└└ (2,2,2,1,1,0) [289.77M\$]

Discount factor = 0.91

Probability of finishing activity j first : $(1/D_j) \times (\text{SUM}(1/D_j))^{-1}$

=> Probability 3 finishes first = 50% & $p_3 = 100\%$

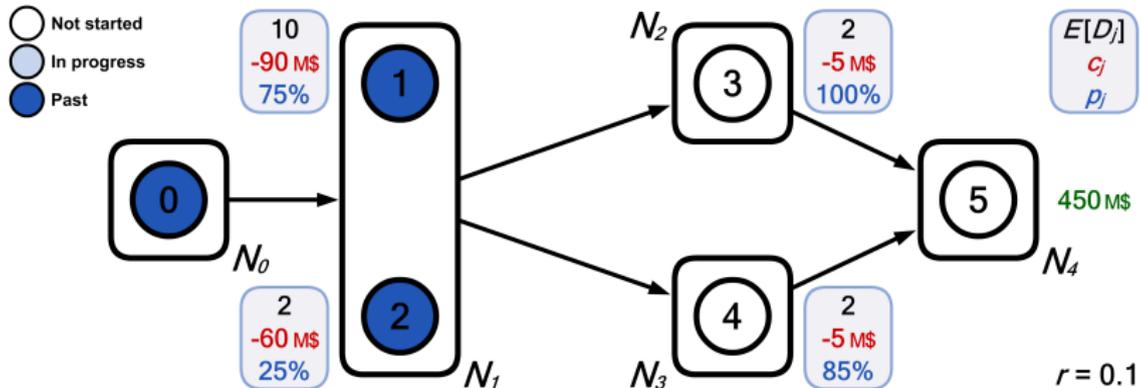
$0.5 \times 0.91 \times 1.00 \times 318.75 = 144.89$

=> Probability 4 finishes first = 50% & $p_4 = 0.85\%$

$0.5 \times 0.91 \times 0.85 \times 375 = 144.89$

=> NPV upon state entry = 289.77

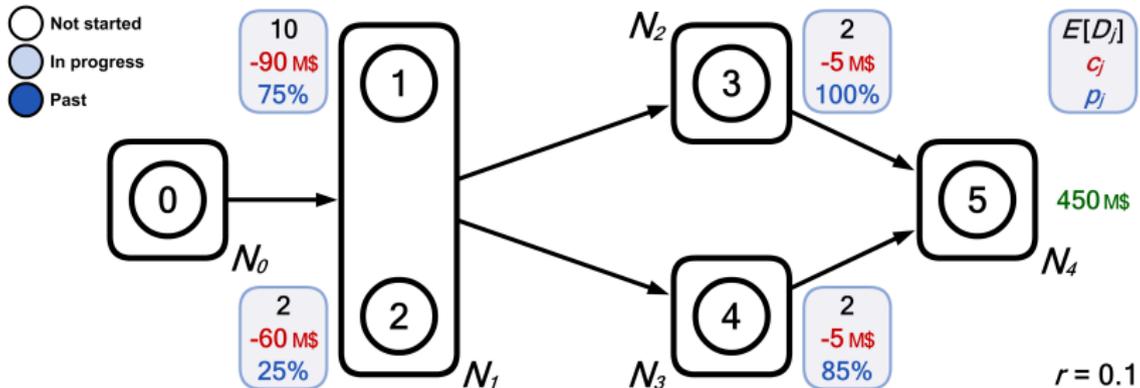
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- (2,2,2,2,2,0) [450M\$]
- └─ (2,2,2,2,1,0) [318.75M\$]
- └─ (2,2,2,1,2,0) [375M\$]
- └─ (2,2,2,1,1,0) [289.77M\$]
- └─ (2,2,2,0,0,0) [279.77M\$]

- 3 possible decisions (pick the optimal one):
- Start activity 3 => incur cost $c_3 = -5M\$$
=> end up in (2,2,1,0,0)
 - Start activity 4 => incur cost $c_4 = -5M\$$
=> end up in (2,2,0,1,0)
 - Start activity 3 & 4 => incur cost $c_3 + c_4 = -10M\$$
=> end up in (2,2,1,1,0)[289.77M\$]

Example: Stochastic Durations



- (2,2,2,2,2,0) [450M\$]
- └ (2,2,2,2,1,0) [318.75M\$]
- └ (2,2,2,1,2,0) [375M\$]
- └ (2,2,2,1,1,0) [289.77M\$]
- └ (2,2,2,0,0,0) [279.77M\$]
- └ (...)
- └ (0,0,0,0,0,0) [14.70M\$]

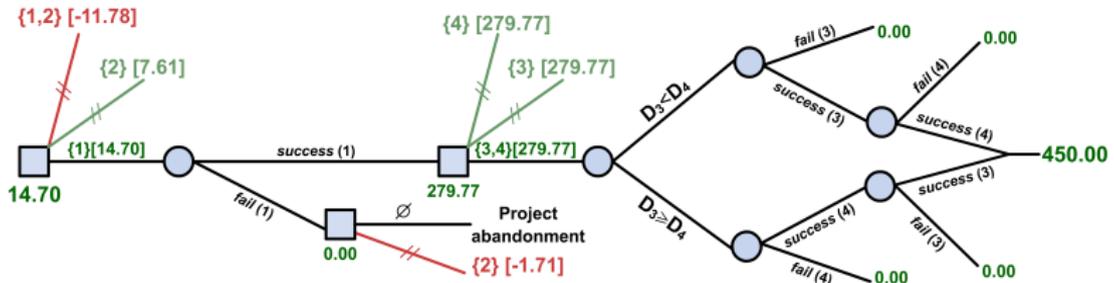
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Policy Π_1 results in a NPV of **-6.35M\$** if activity durations are deterministic

Policy Π_2 is optimal for deterministic durations and yields a NPV of **2.05M\$**

For stochastic durations, policy Π_1 is optimal with a NPV of **14.70M\$**



Results & Future Research

Computational results:

- 100 project networks were generated varying in size from 75 activities up to 120 activities. Out of these project networks, 75 have been solved to optimality.
- Computation times vary from less than a second to a maximum of 81,593 seconds. The average computation time for those networks solved amounts to 4,808 seconds.
- The main determinant of the computation time is the density of the network.

Future research:

- Using the model to generate insights in the use of modules
- General activity durations using Phase-Type distributions
- Resources

Time for questions

