



Using the P-factor to Monitor Concurrent Engineering Projects with a Variable Chance of Rework

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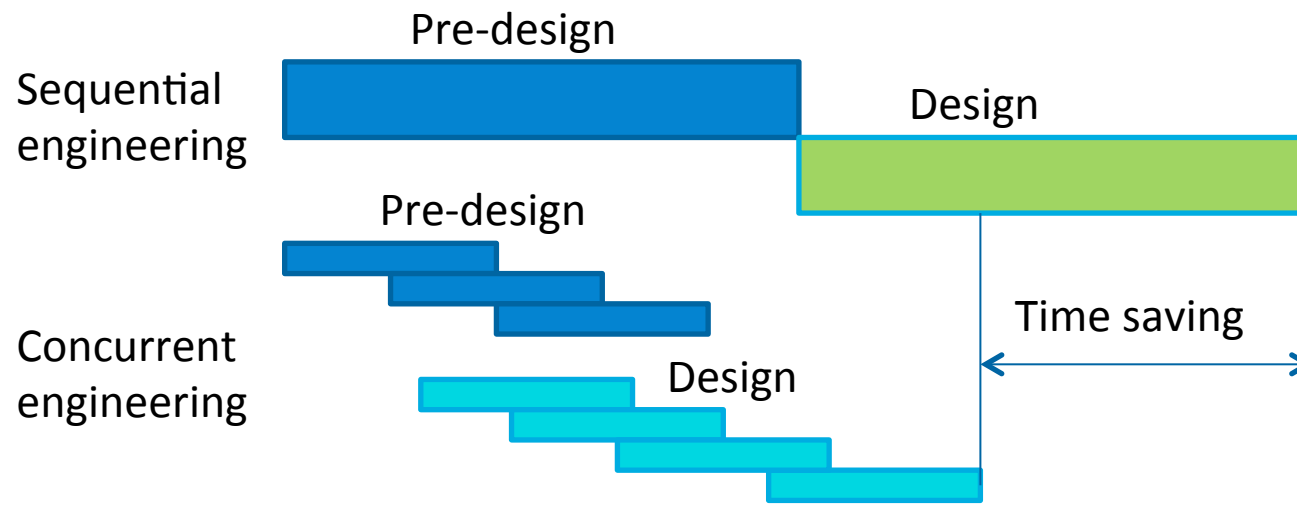


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Basics of concurrent engineering

- Concurrent Engineering?



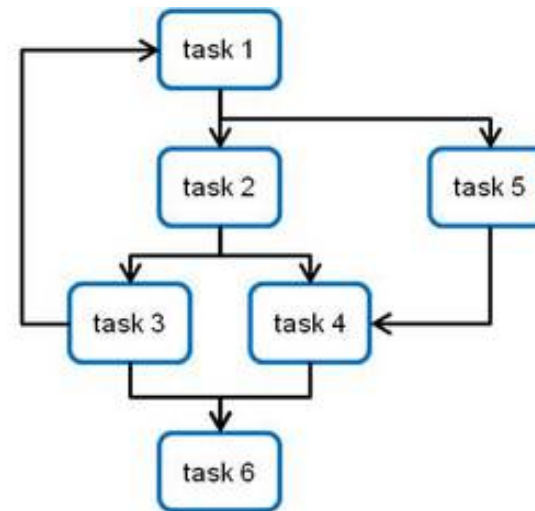
Advantage: time saving

Disadvantage: chance of rework

Basics of concurrent engineering

- Design Structure Matrix (DSM)?

	task 1	task 2	task 3	task 4	task 5	task 6
task 1		X			X	
task 2			X	X		
task 3	X					X
task 4						X
task 5				X		
task 6						

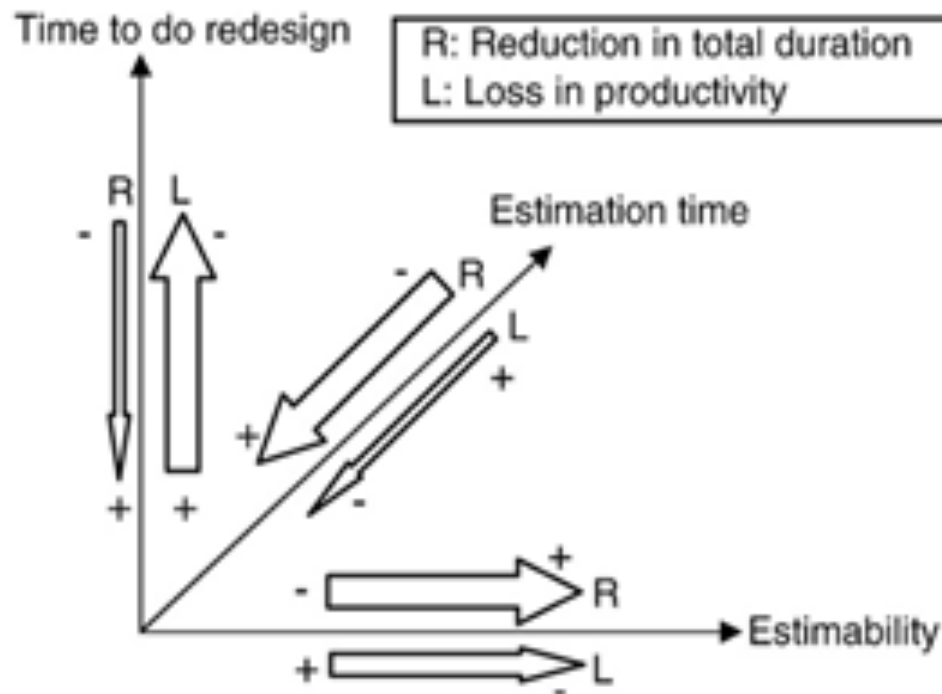


http://www.dsmweb.org/typo3temp/pics/DSM_Tutorial_Basic3_9f5f52512b.jpg

- Binary DSM/Numerical DSM

Existing research

- Chua, D. K. H., & Hossain, M. A. (2011). A simulation model to study the impact of early information on design duration and redesign:



Goal: Study the impact of these three factors on total duration and productivity

Probability of redesign changes for different values, but constant throughout simulation

Fig. 7. Influence of three factors on reduction in total duration and loss in productivity.

Existing research

- Bogus, S., & Diekmann, J. (2011). Simulation of Overlapping Design Activities in Concurrent Engineering:

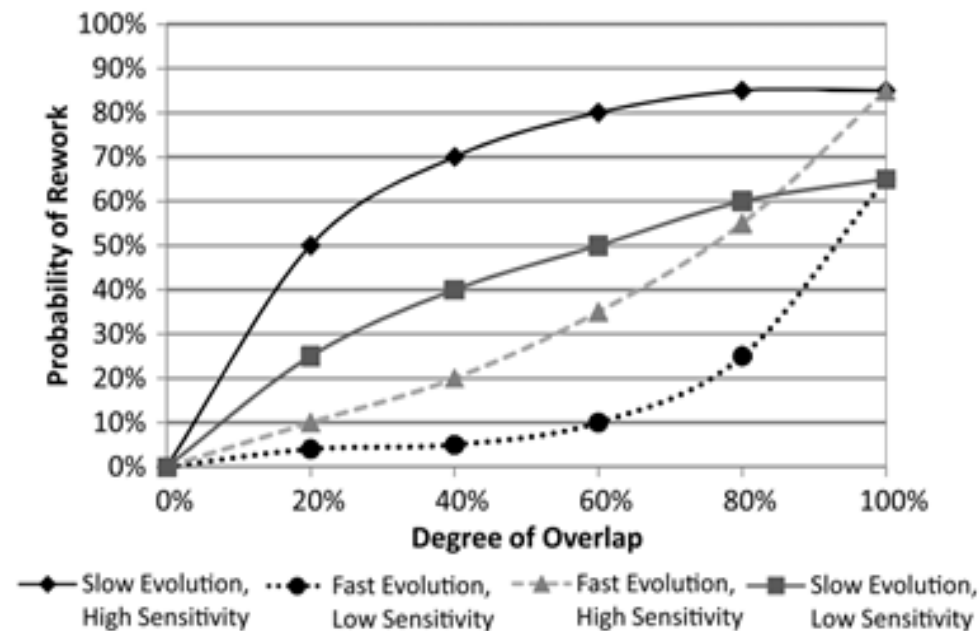


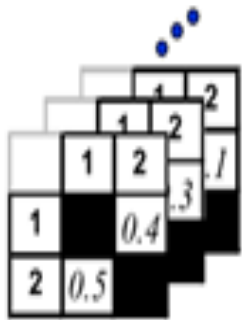
Fig. 3. Theoretical relationship between probability of rework and degree of overlap

Goal: determine optimal pairs of evolution and

probability of rework changes for sensitivity/evolution, but remains constant during simulation

Existing research

- Smith, R., & Morrow, J. (2001). Product development process modeling:



$RP(i, j, r)$



$RI(i, j) \quad (i, j=1,2; r=1,2,...)$

Figure 4. Rework Probability and Impact

Goal: “The model can be used for better project planning and control by identifying leverage points for process improvements and evaluating alternative planning and execution strategies”

Rework probability is fixed



Existing research

Existing simulation models

- Constant chance of rework during simulation run
- Not very realistic!

Amount of rework depends on schedule adherence!



Research questions

Is there a way to calculate schedule adherence?

P-factor =

Portion of earned value accrued in congruence with schedule

Formula:

$$p = \frac{\sum_{i \in N} \min(PV_{i,ES}, EV_{i,AT})}{\sum_{i \in N} PV_{i,ES}}$$

N

Set of activities in a project

$PV_{i,ES}$

Planned Value of activity i at time instance ES

$EV_{i,AT}$

Earned Value of activity i at the current time AT

Lies between 0 and 1

Research questions

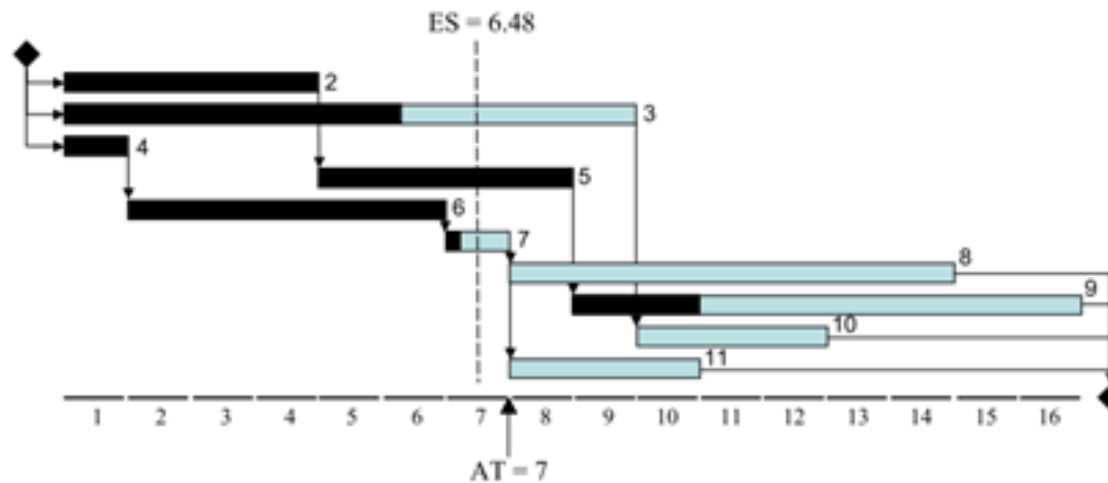


Fig. 2.3 Real life execution of the example project relative to the baseline schedule

Vanhoucke, M. (2009). Measuring time: improving project performance using earned value management (p. 183).

Graphical interpretation: Black bars left of ES-line / total EV



Research questions

Does schedule adherence influence rework?

Decrease of schedule adherence

→ Increased risk of rework

Idea: Use of p-factor to calculate rework

$$R = f(r) \cdot EV(r) = f(r) \cdot (1 - P) \times EV$$

With: $f(r) = 1 - C^n \cdot e^{(-m \cdot (1 - C))}$

C = fraction complete of project (EV/BAC)

e = natural number (base "e")

^ = signifies an exponent follows



Research questions

Problems: No trend information and no forecast total rework possible

Solution: Normalize R to work remaining

$$\text{Schedule Adherence Index} = \text{SAI} = R / (\text{BAC} - \text{EV})$$

Increasing SAI = decreasing schedule adherence

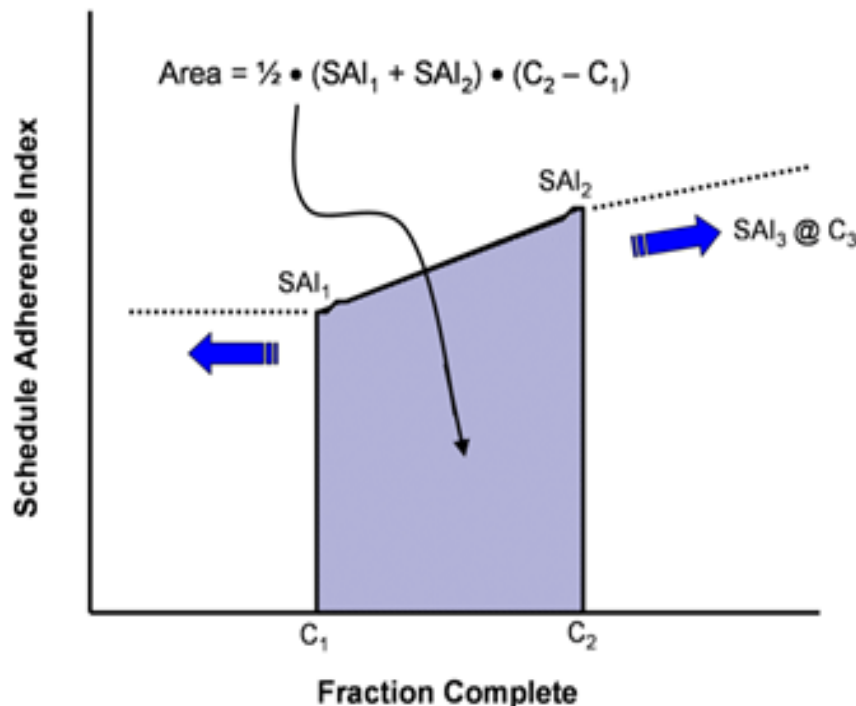
Characteristics:

1. Enables forecasting of total rework
2. Takes cumulative effects of imperfect schedule adherence into account

Research questions

Forecast of total rework:

Compute rework through trapezoidal approximation technique



Lipke, W. (2010). Schedule Adherence and Rework.

$$R_p(n) = BAC \cdot \left[\frac{1}{2} \cdot (SAI_n + SAI_{n-1}) \cdot (C_n - C_{n-1}) \right]$$

n = the performance period of interest

Area multiplied with BAC to obtain rework cost for any period

$$R_{cum} = \sum R_p(n)$$

$$R_{tot} = R_{cum} + SAI \cdot (BAC - EV)$$



Research questions

How to determine chance of rework based upon this forecast?

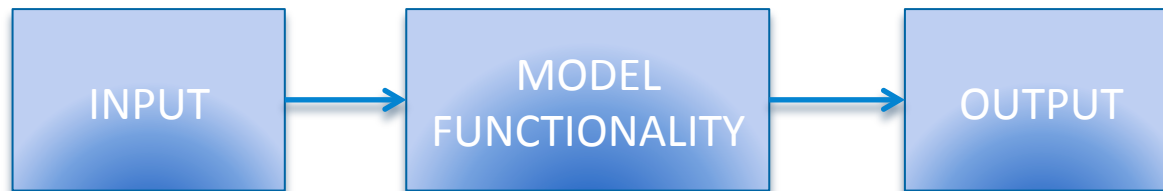
The forecasted rework will be used to compute a chance of rework at each status point of the project

$R_{tot} = \text{Chance of rework} \times \text{rework impact} \times \text{duration}$

→ $\text{Chance of rework} = R_{tot} / (\text{rework impact} \times \text{duration})$



The simulation model





Input levers

Activity
duration

DSM

Rework Impact

Evolution

Sensitivity

Risk-factor



Input levers

Activity
duration

3 estimates for each activity:

1. Worst-case estimate(=P)
2. Most likely estimate(=M)
3. Best-case estimate(=O)

DSM

Rework Impact

This is also known as the PERT method:

Expected value of activity duration= $(O + 4M + P) / 6$

Evolution

Sensitivity

Random values will be selected between worst-case and best-case estimates

Risk-factor

Input levers

Activity
duration

DSM

Rework Impact

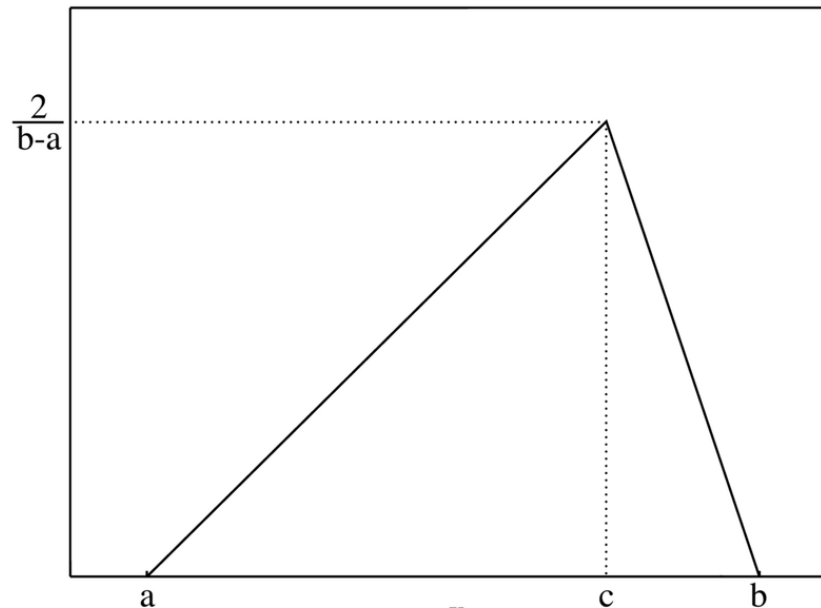
Evolution

Sensitivity

Risk-factor

Assumption: Each activity duration follows a triangular distribution

Probability density function of triangular distribution:



http://upload.wikimedia.org/wikipedia/commons/thumb/4/45/Triangular_distribution_PMF.png/325px-Triangular_distribution_PMF.png



Input levers

Activity
duration

DSM

Rework Impact

Evolution

Sensitivity

Risk-factor

Binary DSM to represent information dependency

Used to:

1. Make schedule:
100% overlapping between independent activities
2. Add rework:
Rework added only to dependent activities



Input levers

Activity
duration

DSM

Rework Impact

Evolution

Sensitivity

Risk-factor

Characteristics:

1. For each activity rework impact will be needed
2. Determines duration of rework when rework occurs
3. Varies from 0% to 100%

Input levers

Activity
duration

DSM

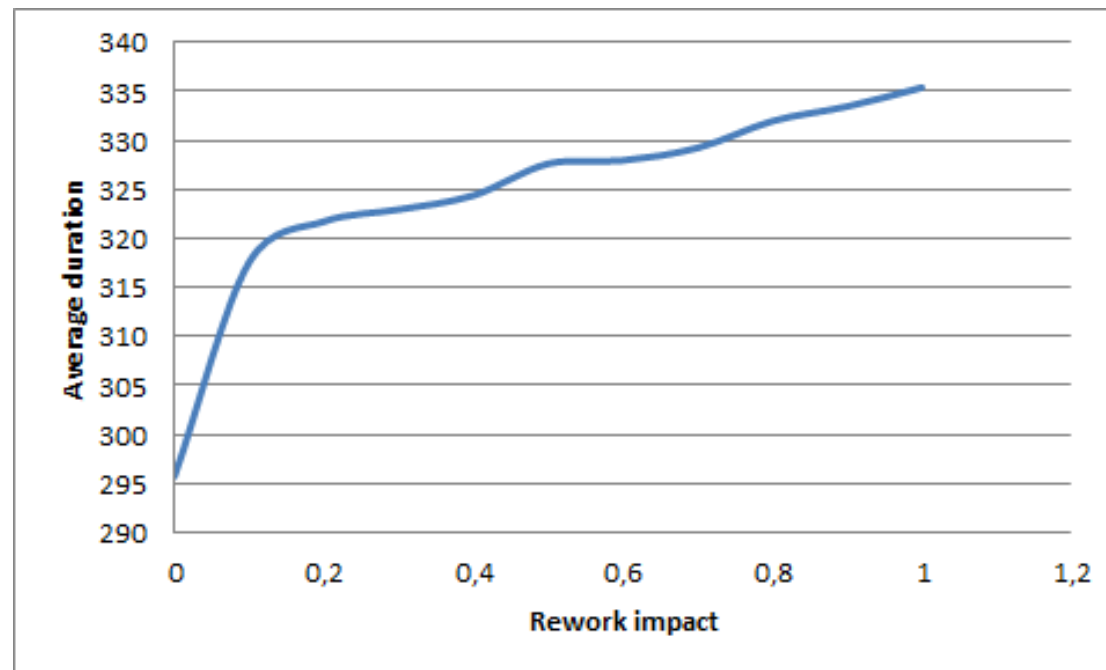
Rework Impact

Evolution

Sensitivity

Risk-factor

Impact on project duration:





Input levers

Activity
duration

DSM

Rework Impact

Evolution

Sensitivity

Risk-factor

Development speed of information and knowledge

4 factors of evolution:

1. Design optimisation
2. Constraint satisfaction
3. External information exchange
4. Standardisation

General rule: More iterations in design is slower evolution



Input levers

Activity
duration

Range of values:
0 = Slow evolution
1 = High evolution

DSM

Rework Impact

Influence on model:

Changing evolution will lead to a change in
overlapping degree

Evolution

Sensitivity

0	10% - 20% overlap
1	40% - 60% overlap

Risk-factor



Input levers

Activity
duration

Sensitivity of downstream activity to change of information coming from an upstream activity

DSM

Factors of sensitivity:

Rework Impact

1. Constraint sensitive
2. Input sensitive
3. Integration sensitive

Evolution

Value range:

Sensitivity

0 = Low sensitivity
1 = High sensitivity

Risk-factor



Input levers

Activity
duration

DSM

Rework Impact

Evolution

Sensitivity

Risk-factor

Influence on model:

Changing sensitivity will also lead to a change in overlapping degree

0 → 20% - 60% overlap

1 → 10% - 40% overlap



Input levers

Activity
duration

Risk-factor: value of 0, 1 or 2

DSM

Determines degree of overlap in accordance with figure 3 of Bogus, S., & Diekmann, J. (2011). Simulation of Overlapping Design Activities in Concurrent Engineering.

Rework Impact

Evolution

Values for overlapping degree (for different combinations of evolution/sensitivity):

Sensitivity

0 → 0% overlap

1 → 10% (0,1); 20% (0,0); 40% (1,1); 60% (1,0)

Risk-factor

2 → 20% (0,1); 40% (0,0); 60% (1,1); 80% (1,0)

Input levers

Activity
duration

DSM

Rework Impact

Evolution

Sensitivity

Risk-factor

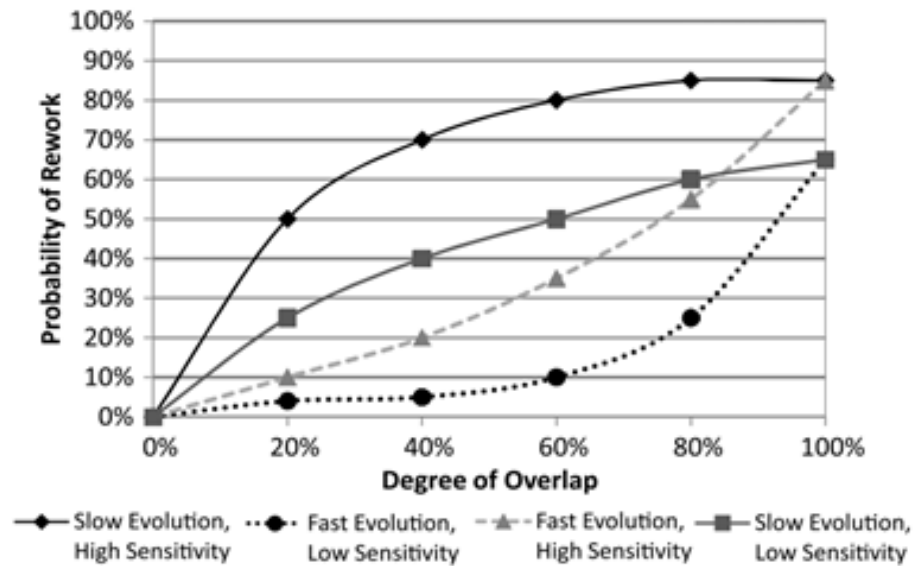


Fig. 3. Theoretical relationship between probability of rework and degree of overlap

Bogus, S., & Diekmann, J. (2011). Simulation of Overlapping Design Activities in Concurrent Engineering.



Input levers

Activity
duration

DSM

Rework Impact

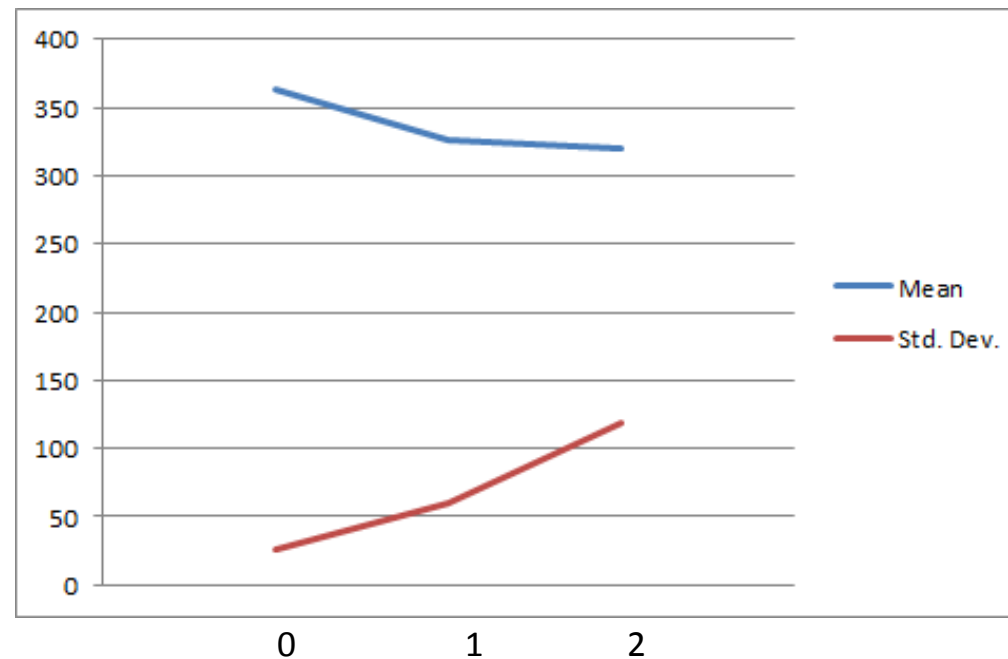
Evolution

Sensitivity

Risk-factor

Impact of increasing risk:

1. Mean project duration decreases
2. Variance/ standard deviation increases
3. Chance of rework increases





Model Functionality

Simulation language: C++

Steps of simulation:

1. Read input files
2. Determine degree of overlap
3. Create baseline schedule
4. Randomize durations
5. Determine real schedule
6. Calculate SAI and forecast total rework
7. Calculate chance of rework for each activity
8. Add rework
9. Advance to next status point



Outputs

Two outputs will be generated:

1. Average SAI
2. The actual project duration

This should allow to find link between SAI and project duration for different values of input variables



Exemplary problem

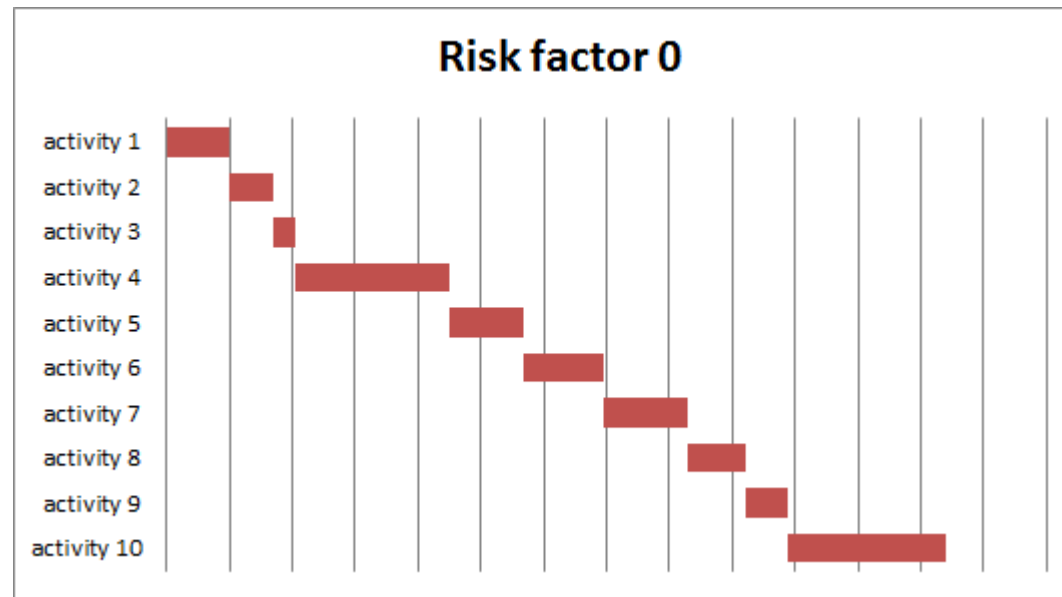
Project characteristics:

- 10 activities
- Start 01/01/2013
- Rework impact 0,5
- Activities use information from all previous activities

10.000 simulation runs for each risk factor

Exemplary problem

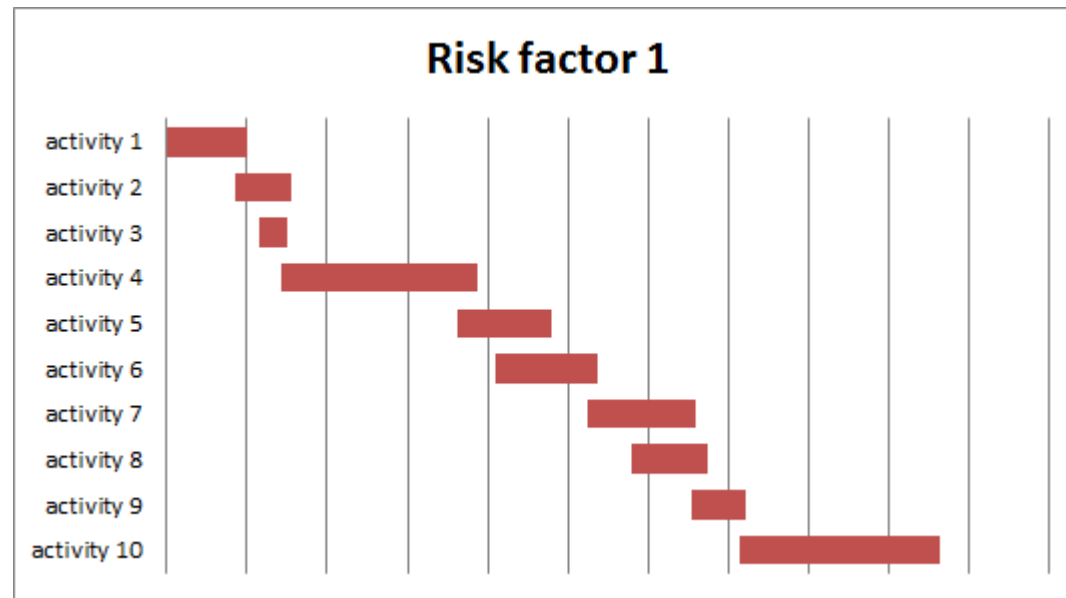
Baseline schedule(risk factor 0):



Expected duration: 371,6667 days

Exemplary problem

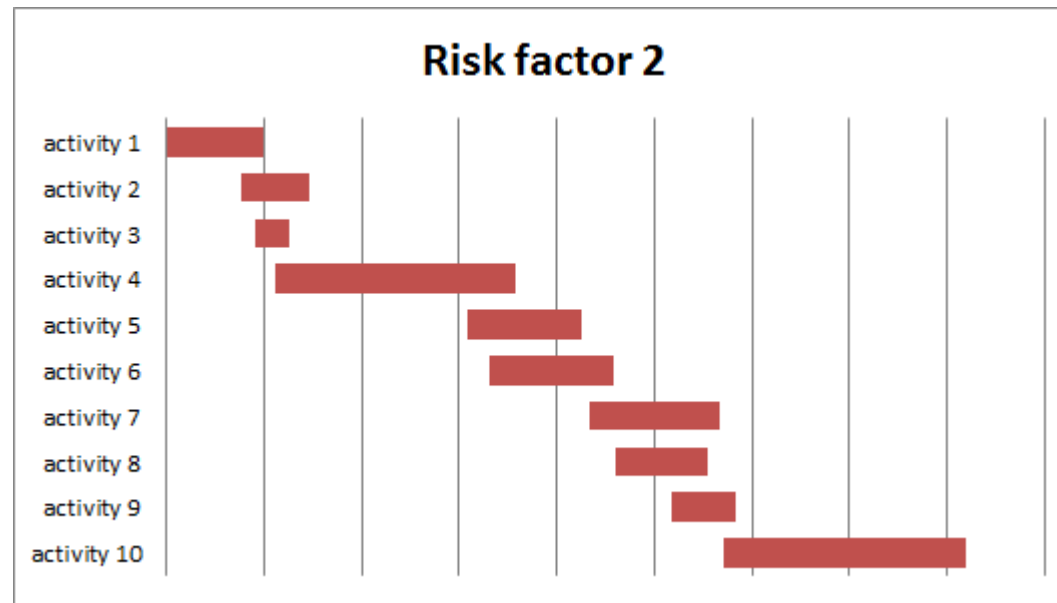
Baseline schedule(risk factor 1):



Expected duration: 289,16 days

Exemplary problem

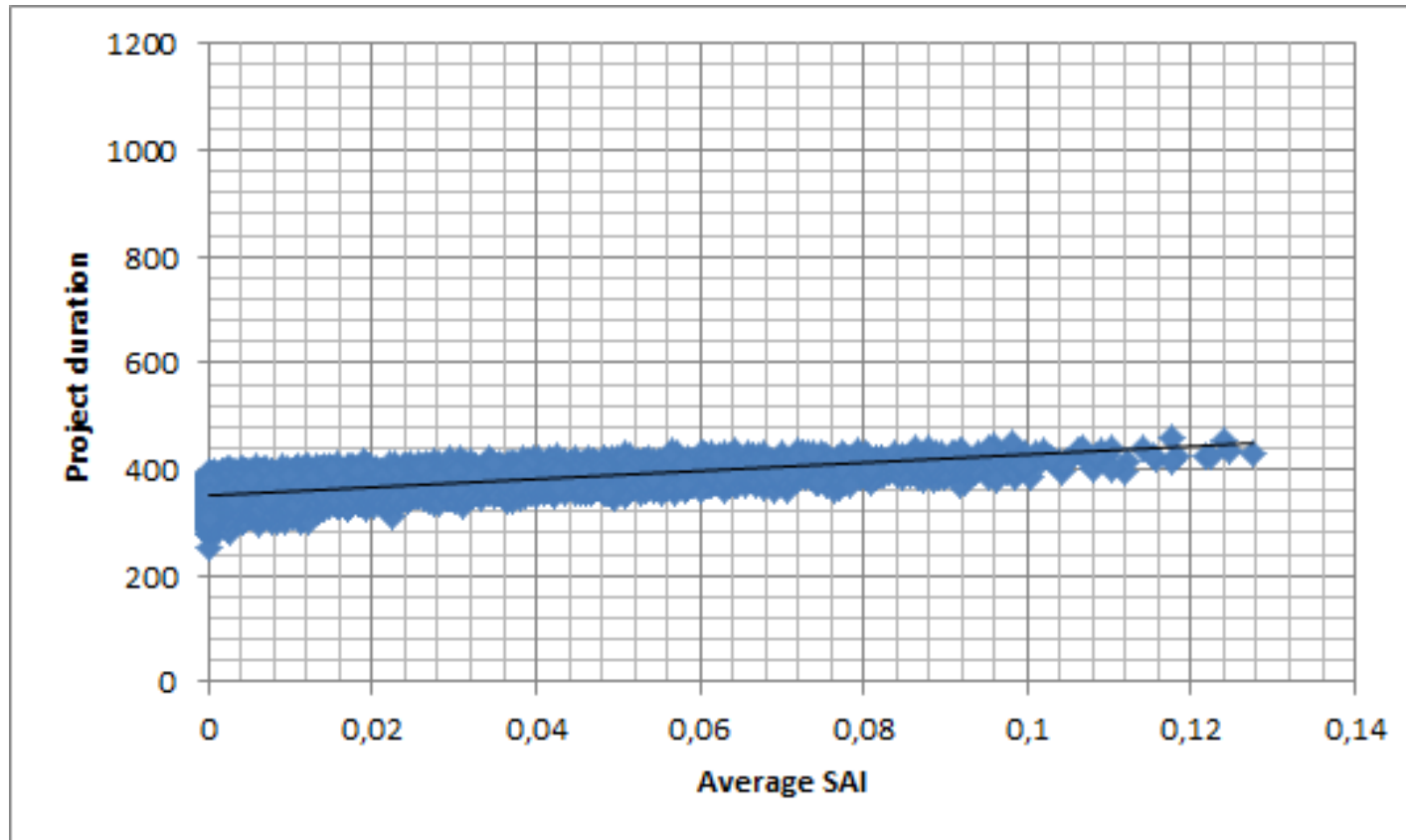
Baseline schedule(risk factor 2):



Expected duration: 246 days

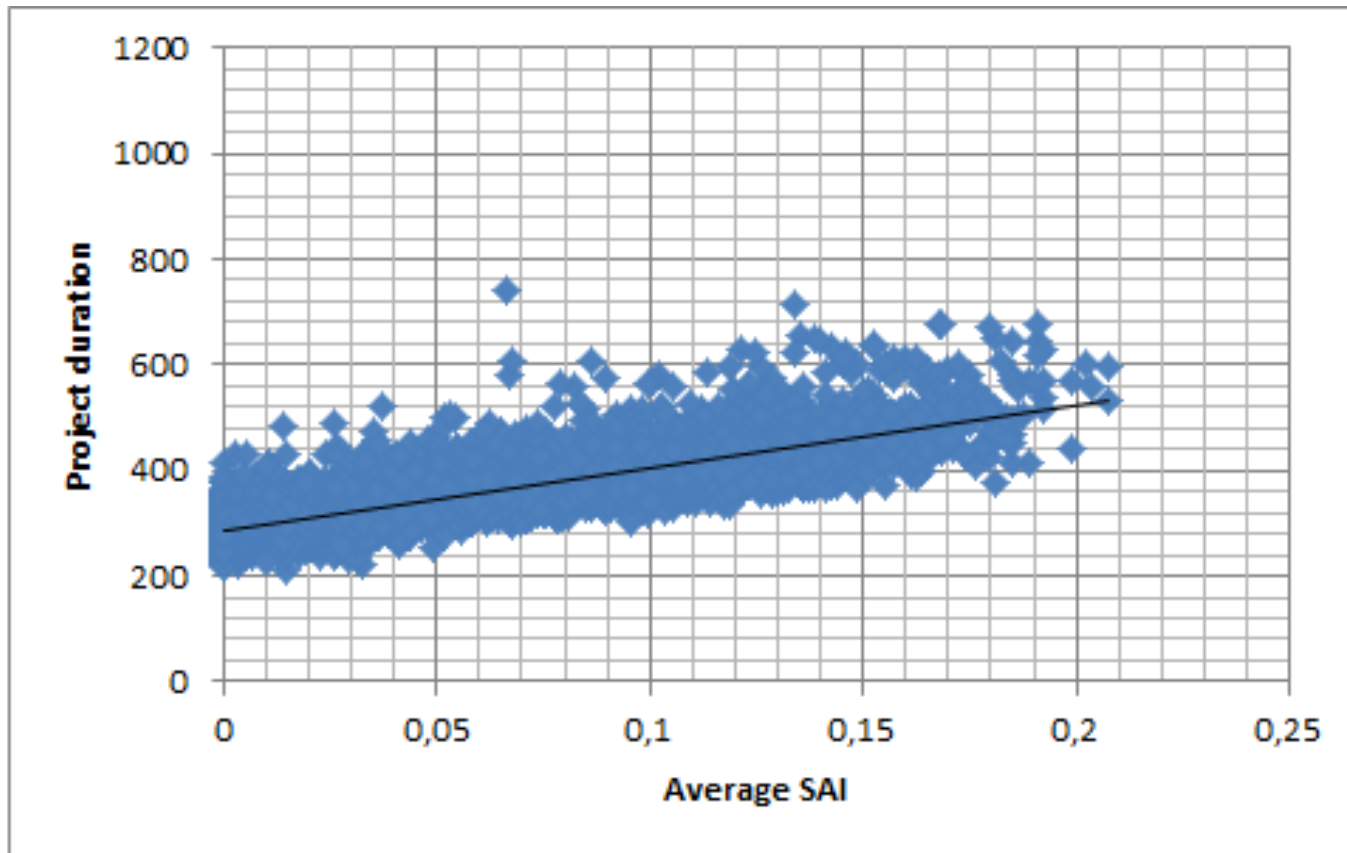
Exemplary problem

Risk-factor 0:



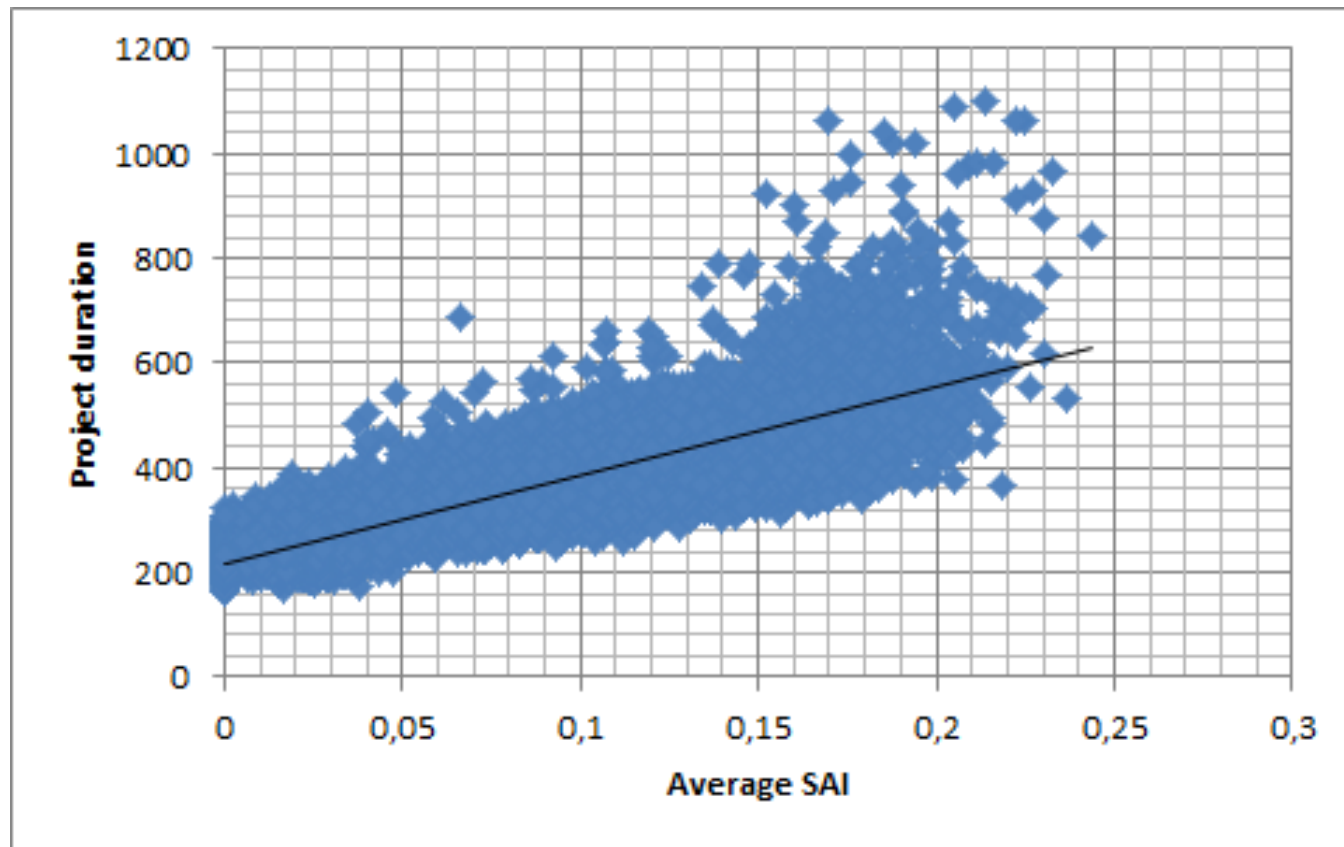
Exemplary problem

Risk-factor 1:



Exemplary problem

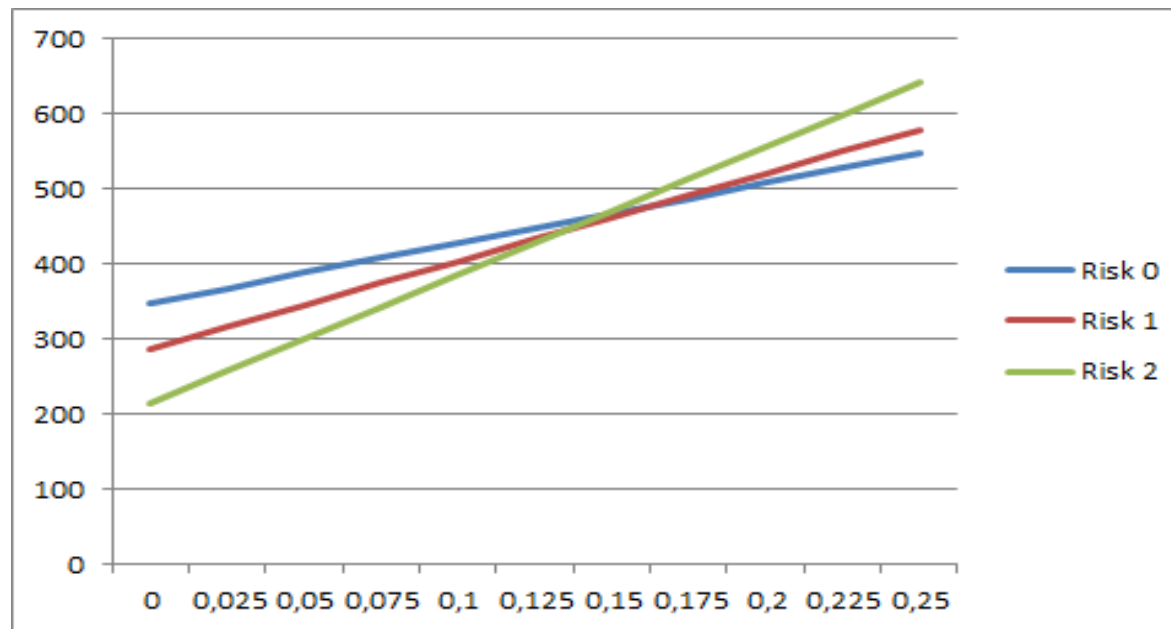
Risk-factor 2:



Exemplary problem

Regression results:

1. $DUR_0 = 348.722156499855 + 795.1174 * SAI$
2. $DUR_1 = 286.9742 + 1169.713 * SAI$
3. $DUR_2 = 215.6778 + 1703.153 * SAI$





Exemplary problem

For what values of SAI is risk not appropriate?

Look at intersections:

1. Risk 0 and Risk 1 at 0,165153 SAI
2. Risk 0 and Risk 2 at 0,146519 SAI

On average from SAI of 0,165 scheduling sequential is better than scheduling concurrent for the example project



Conclusion

The example shows us that concurrent scheduling isn't always beneficial

Simulation model allows managers to identify the degree of schedule adherence when scheduling sequential becomes more beneficial for a certain project

This can be used to, for example, modify the baseline schedule when schedule adherence deteriorates to much during project execution



Questions?

Thank you very much for your attention!



References

- Chua, D. K. H., & Hossain, M. A. (2011). A simulation model to study the impact of early information on design duration and redesign.
- Bogus, S., & Diekmann, J. (2011). Simulation of Overlapping Design Activities in Concurrent Engineering.
- Lipke, W. (2010). Schedule Adherence and Rework.
- Vanhoucke, M. (2009). Measuring time: improving project performance using earned value management (p. 183).
- Bogus, S. M., Molenaar, K. R., & Diekmann, J. E. (2006). Strategies for overlapping dependent design activities.
- Smith, R., & Morrow, J. (2001). Product development process modeling.
- Bogus, S. (2005). Concurrent engineering approach to reducing design delivery time.