

Robust Supply Chain Costs Minimization Considering Operational Risks

Invited Talk: NexTech 2020

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Invited Talk



- 2001: Diploma Degree at Saarland University
- 2001-2003: Scientific Assistant at the German Research Center for Artificial Intelligence (DFKI)
- 2003-2006: Scientist at the German Meteorological Service
- 2006-2010: PhD Student at the Distance University of Hagen
- 2011-2015: Postdoc at Goethe University Frankfurt am Main
- 2015-now: Research Associate at Lucerne University of Applied Science and Arts
- 2019-now: Lecturer at FFHS (Fernfachhochschule der Schweiz)

Supply Chain

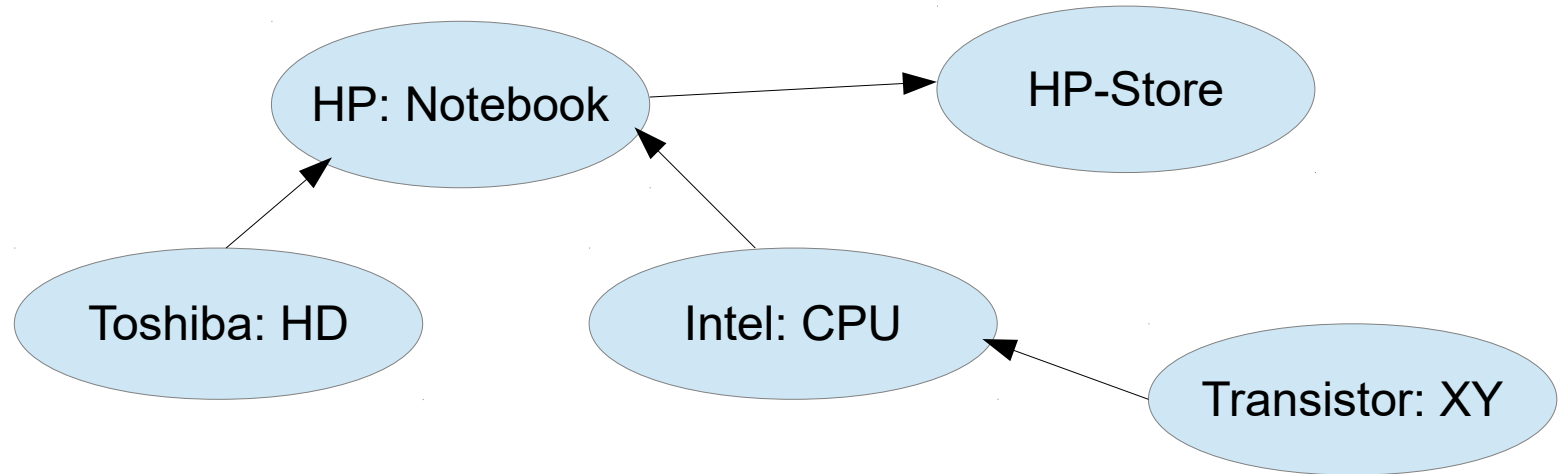
- Consider the production of a notebook
- To produce a notebook one needs
 - Hard drive
 - LCD
 - Keyboard
 - Touchpad ...

To produce an LCD one need:

- Rare Earth Materials
-

Supply Chain

- A supply chain can be represented by a graph
- Nodes: locations / items
- Edges: transportation between locations



Costs minimization

Costs minimization can be accomplished using an optimization model:

$$\text{Minimize } \sum_{iz} c_{iz} P_{iz} + \sum_{ijz} t_{ijz} Q_{ijz} + \sum_{iz} d_{iz} I_{iz}$$

- $D_{iz} \leq \sum_j Q_{j,i,z}$

- $Q_{j,i,z} \leq P_{jz} + I_{ji}$

.....

i, j : locations, z : item, c_{iz} d_{iz} t_{ijz} : production/inventory/transportation costs

- D_{iz} : demand of item z at location i

- P_{jz} : Number of items z produced at location j

- I_{ji} : number of items z contained in inventory at j

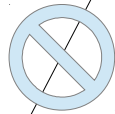
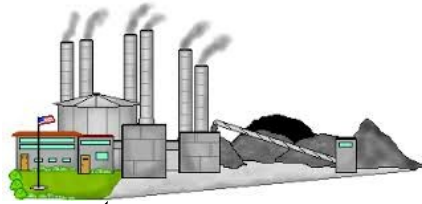
- $Q_{j,i,z}$: number of items z transported from i to j

Risks

A supply chain can be affected by several risks

- Disruption risks
- Price escalation risks
- Inventory and scheduling risks
- Technology access risk
- Quality risks

Disruption risks



- Possible disruption risks:
- Natural disaster
 - Fire, earth quake, lightning strike, volcano eruption
 - Political risks
 - Labor strike, Brexit
 - Sabotage
 - Cyber-attack, burglars

Group Risks

A certain risk event can cause a simultaneous breakdown of several suppliers

-Examples

- Strike: can affect several countries in the same country and industry branch
- Natural disaster: can cause a breakdown of all suppliers in a certain region

Topological Risk Measures

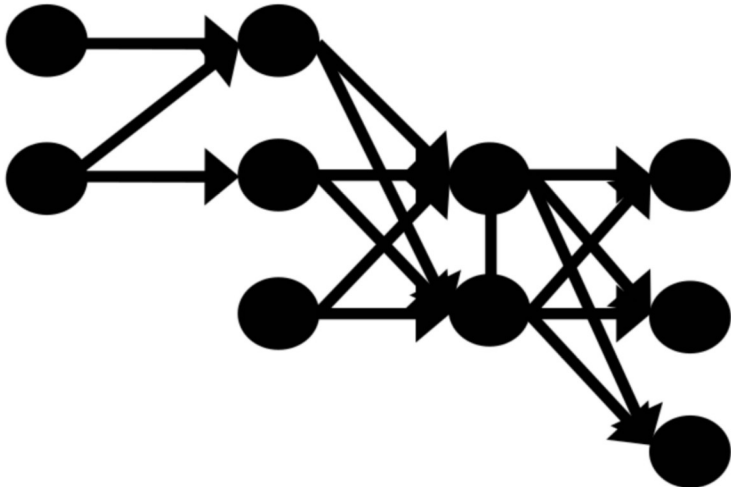
Principle:

Get rough overview over potential supply chain threats by looking at the supply chain structure or geographical map [2,3]

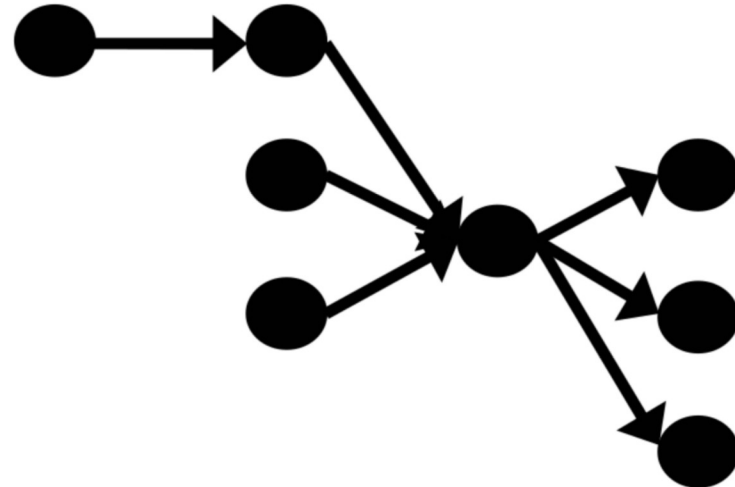
Topological Risk Measures

Critical nodes[2,3]: Nodes that are essential for the proper functioning of the total supply chain

Low supply chain node criticality



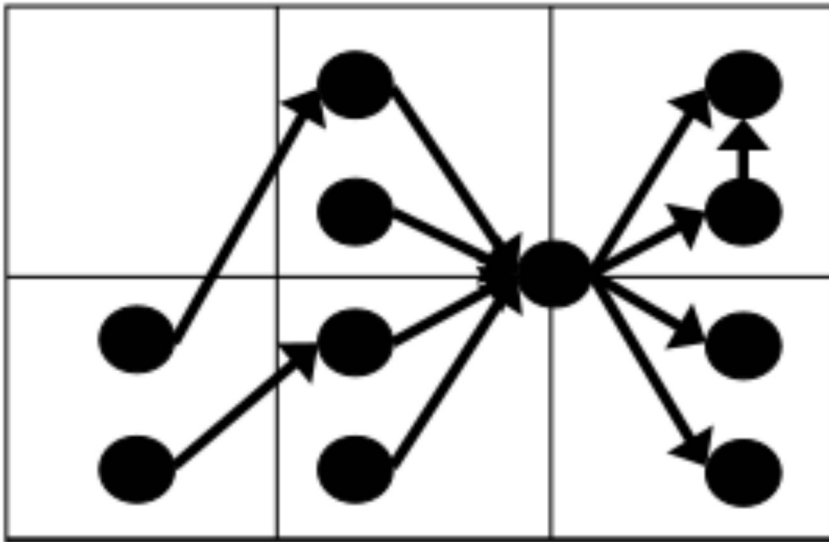
High supply chain node criticality



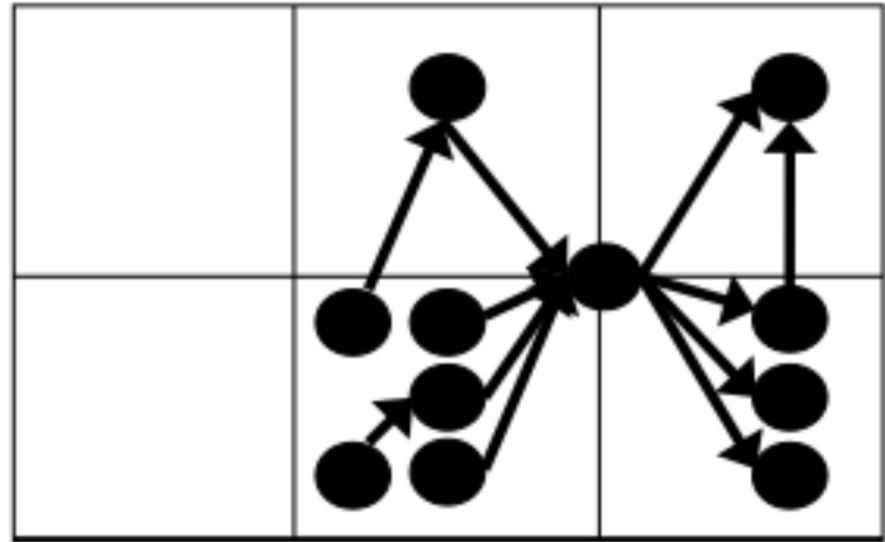
Topological Risk Measures

Density: Supply chain containing a geographical area with high number of suppliers might be vulnerable to group risks

Low supply chain density



High supply chain density



Disruption risks

Topological risk measures only provide a rough overview over potential risks. Is there a more precise method?

Yes → using stochastic optimization

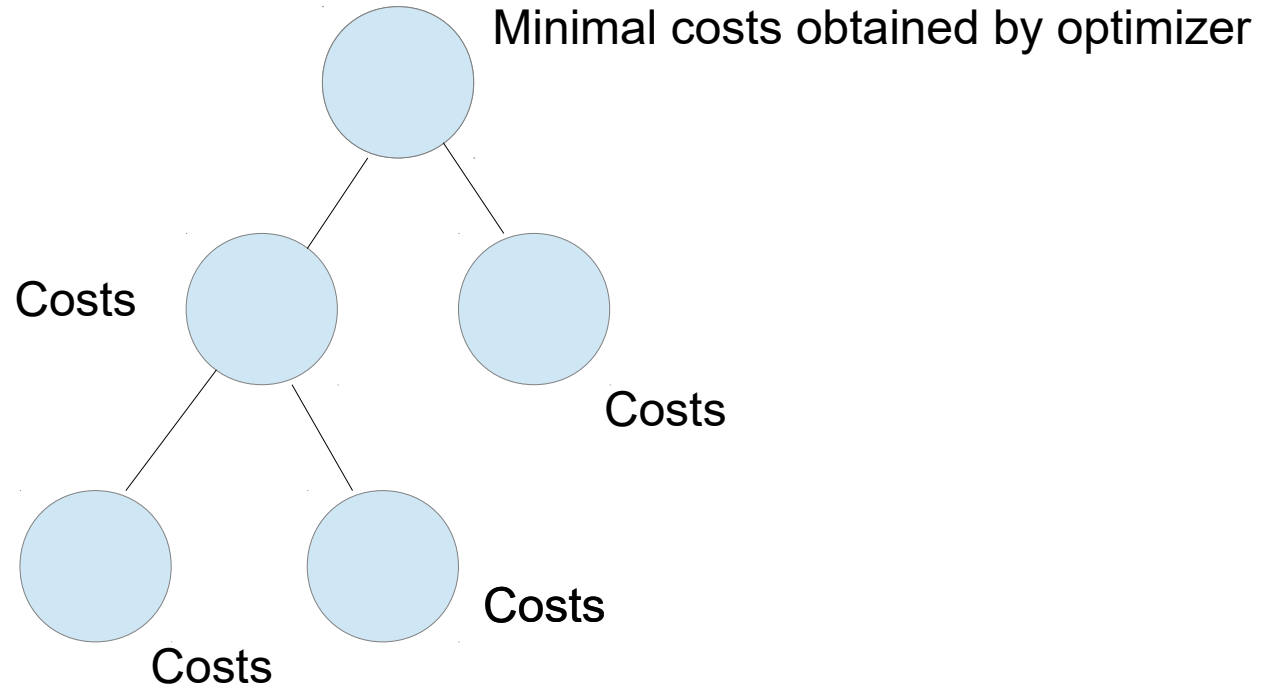
- Two possibilities to extend optimization model to consider disruption risks:
 - Iterate first over risks and then over locations
 - Iterate first over locations and then over risks

Research Questions

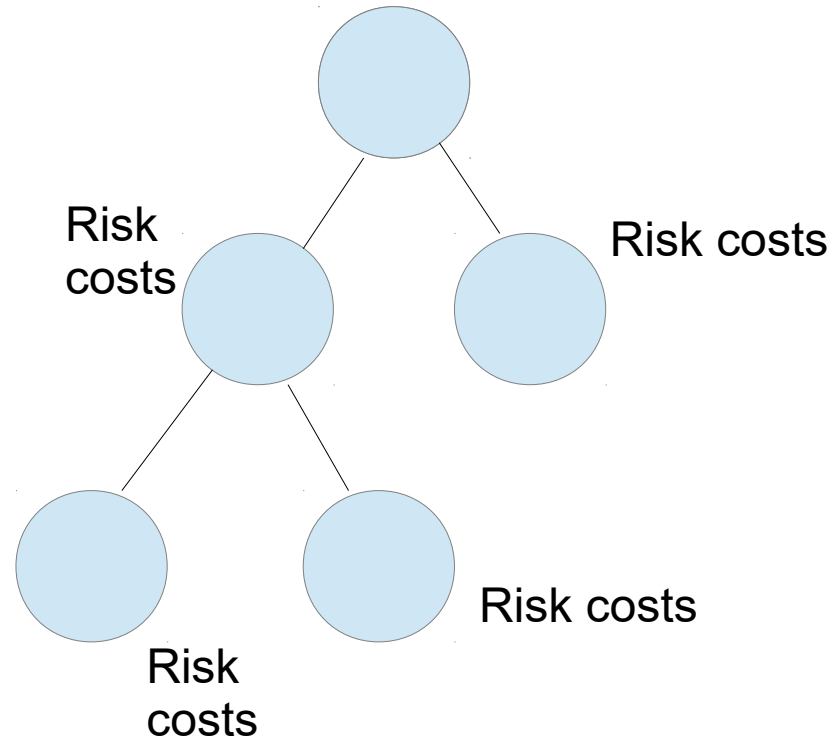
- What are the advantages and drawbacks of the two methods?
- What are advantages and drawbacks of automatic vs manual scenario generation?
- How large is the conducted error by assuming the scenarios are disjoint, while in reality they are stochastically independent?

Stochastic optimization

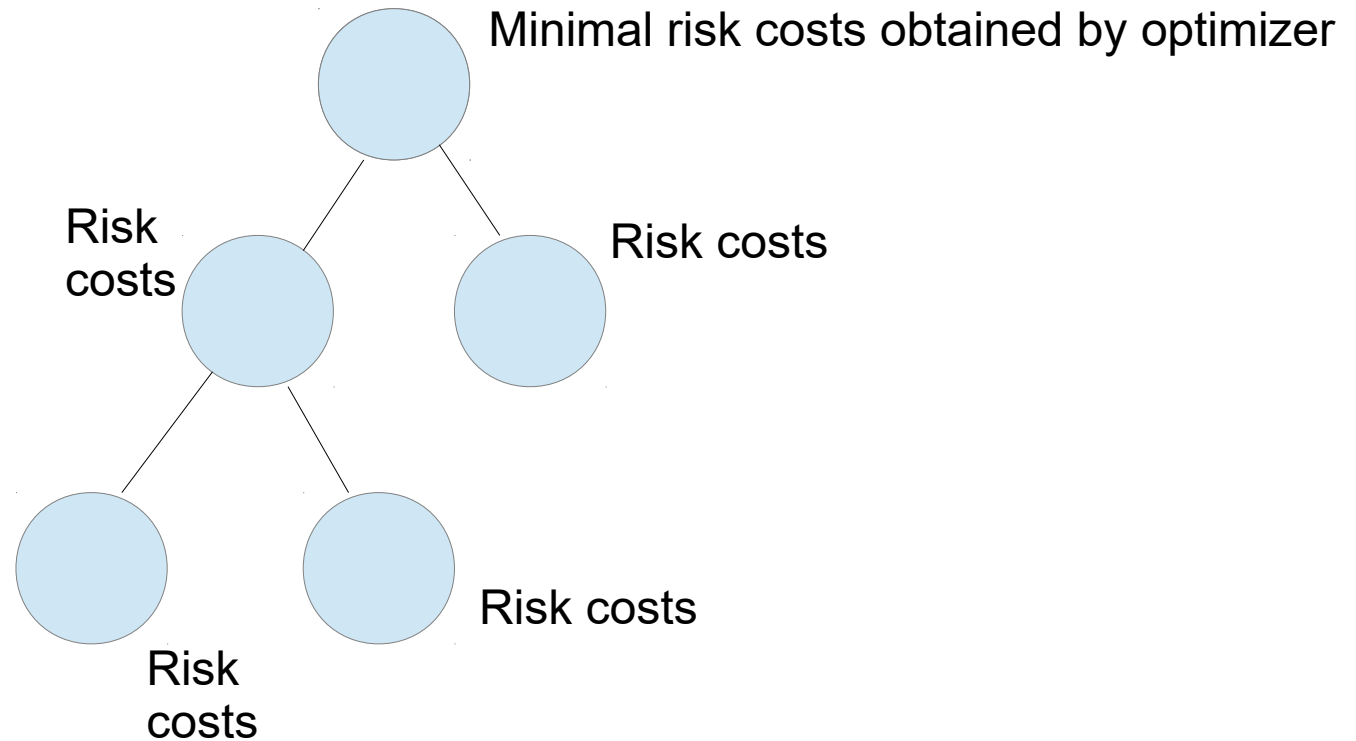
Possibility 1: Iterate first over risks and then over locations



Stochastic optimization



Stochastic optimization



Stochastic optimization

- A risk scenario is always associated to exactly one node in the supply chain
- A local estimation of risk costs is required

Stochastic optimization



- Fast, only one optimization run with only few decision variables
- As result: one single optimal supply chain flow



- Unclear, how to assess single and multiple source risks
- Unclear, how to deal with dependencies between risks and group risks

Stochastic optimization

So far:

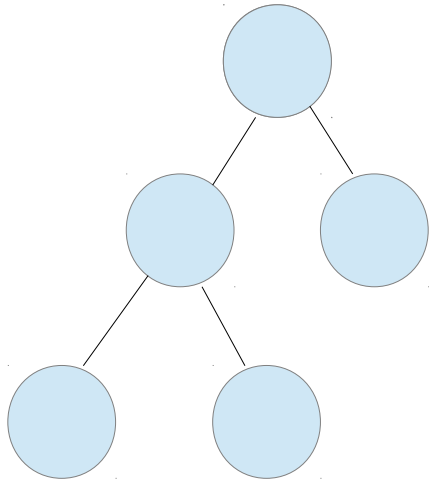
- Aggregate costs
- Iterate over scenarios

Alternative model (Babazadeh and Razmi):

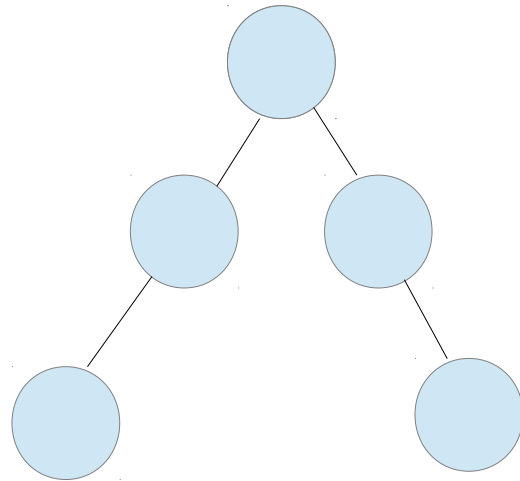
- Iterate over scenarios
- Aggregate costs

Robust optimization [1]

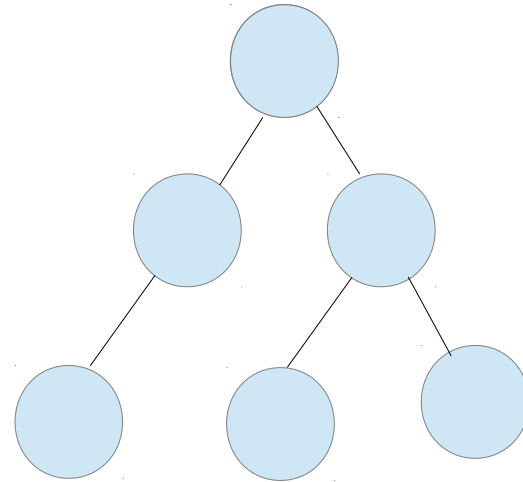
Risk scenario 1



Risk scenario 2

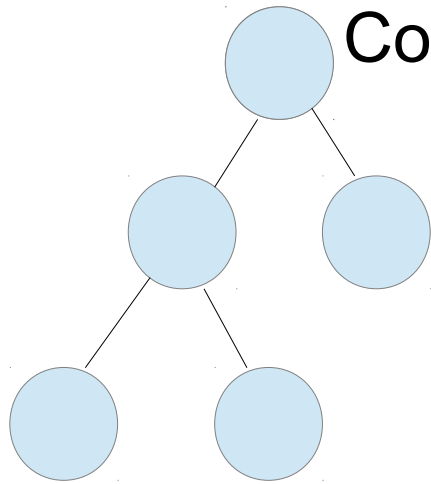


Risk scenario 3



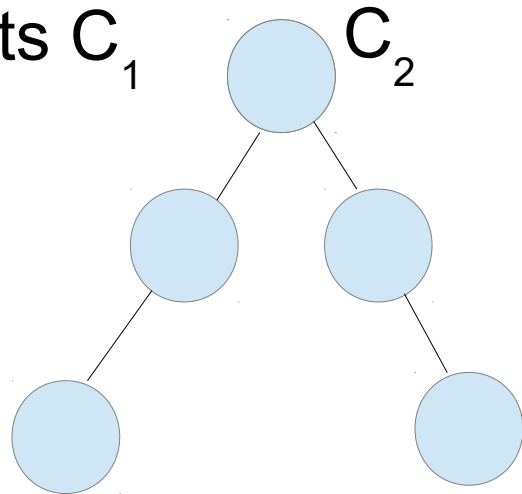
Robust optimization

Risk scenario 1



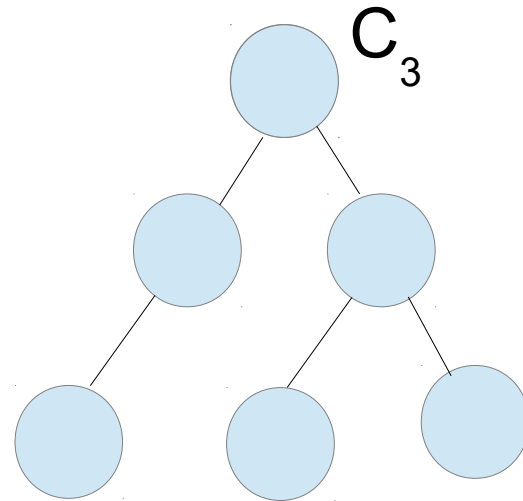
Costs C_1

Risk scenario 2



Costs C_2

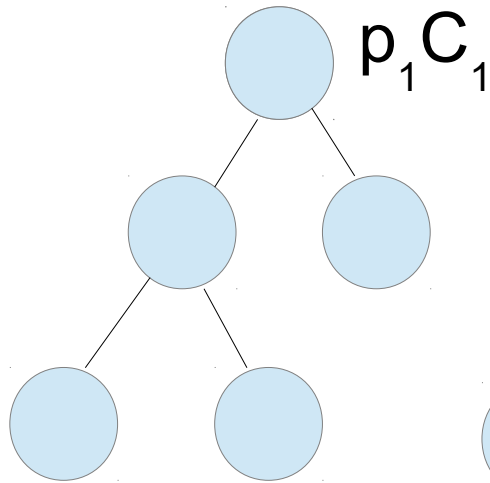
Risk scenario 3



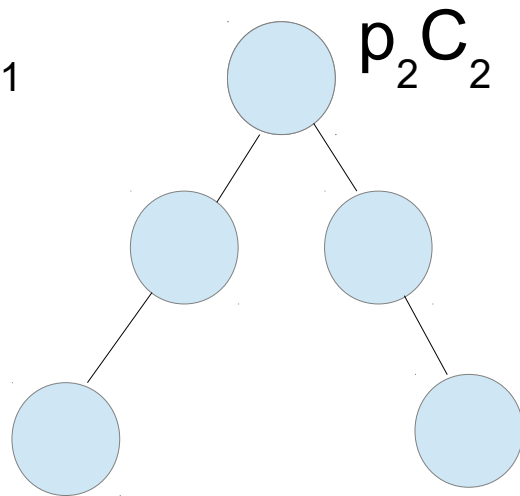
Costs C_3

Robust optimization

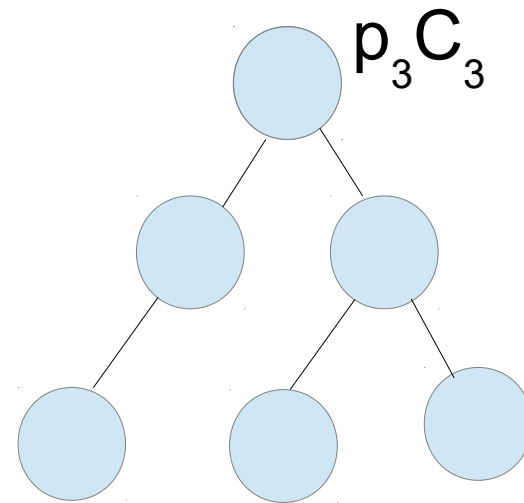
Risk scenario 1



Risk scenario 2

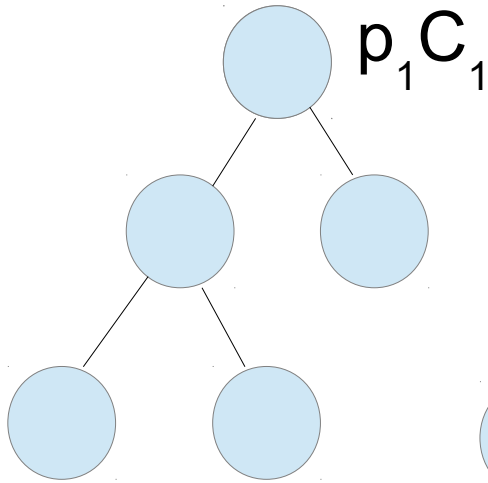


Risk scenario 3

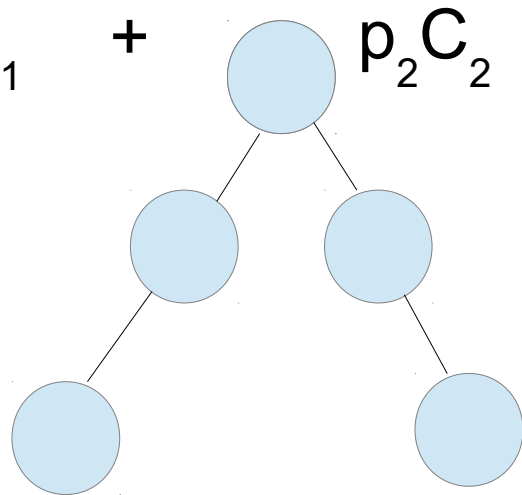


Robust optimization

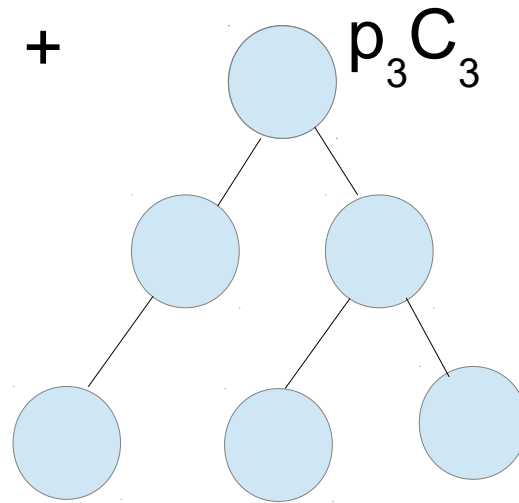
Risk scenario 1



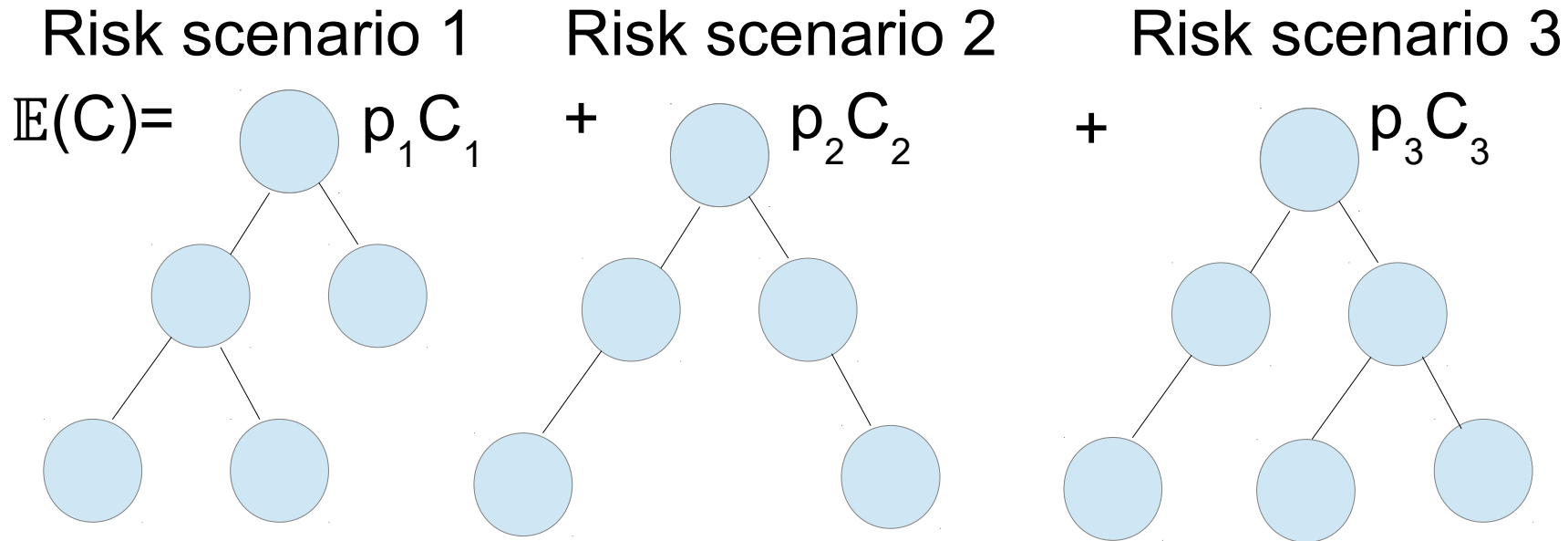
Risk scenario 2



Risk scenario 3



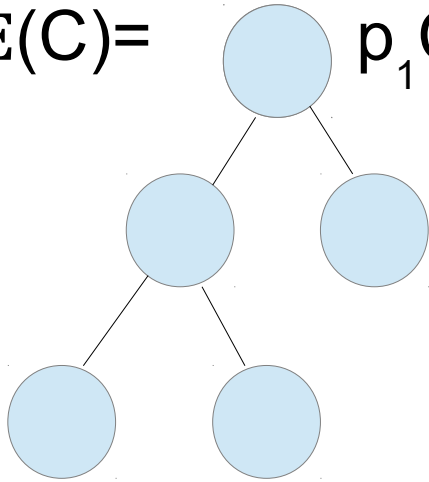
Robust optimization



Robust optimization

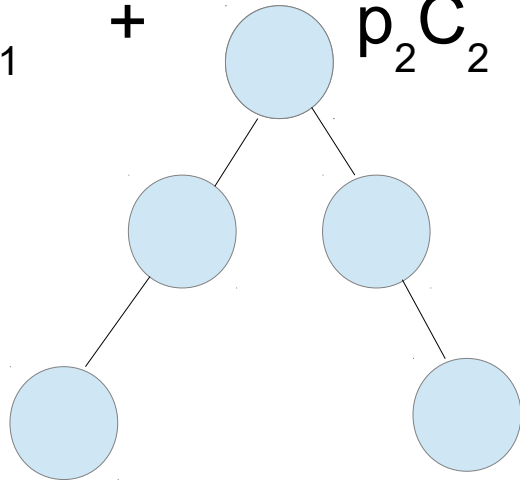
Risk scenario 1

$$\mathbb{E}(C) = p_1 C_1$$



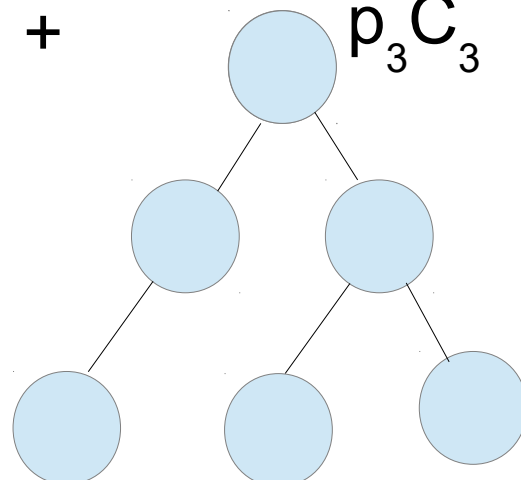
Risk scenario 2

$$+ p_2 C_2$$



Risk scenario 3

$$+ p_3 C_3$$

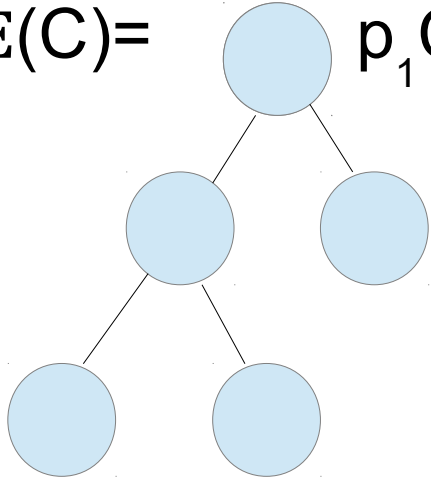


$$\mathbb{V}(C) = \sum_s (p_s (\mathbb{E}(C) - C_s)^2)$$

Robust optimization

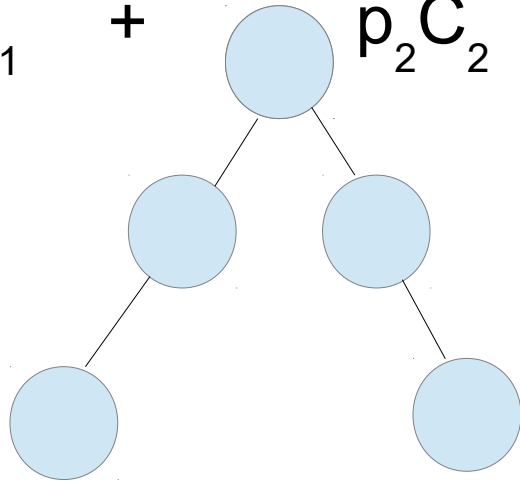
Risk scenario 1

$$\mathbb{E}(C) = p_1 C_1$$



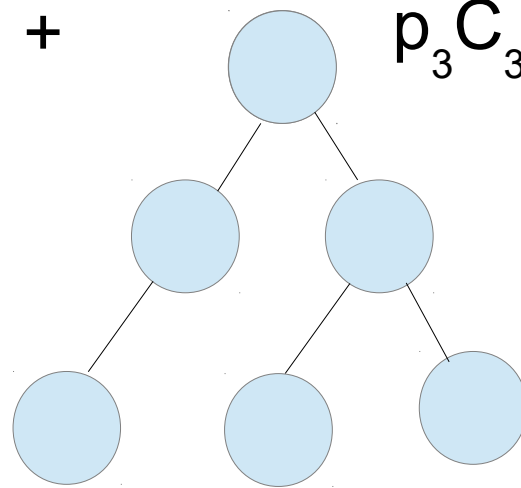
Risk scenario 2

$$+ p_2 C_2$$



Risk scenario 3

$$+ p_3 C_3$$

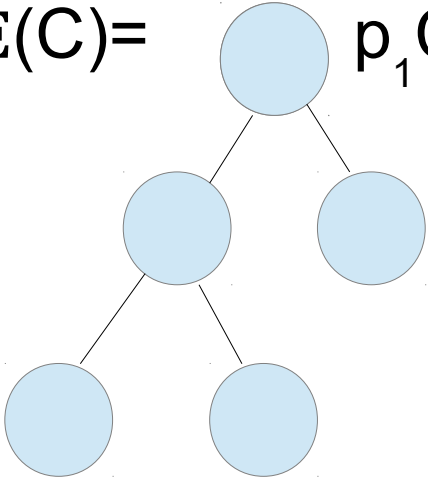


$$\mathbb{V}_{\text{abs}}(C) = \sum_s p_s |(\mathbb{E}(C) - C_s)|$$

Robust optimization

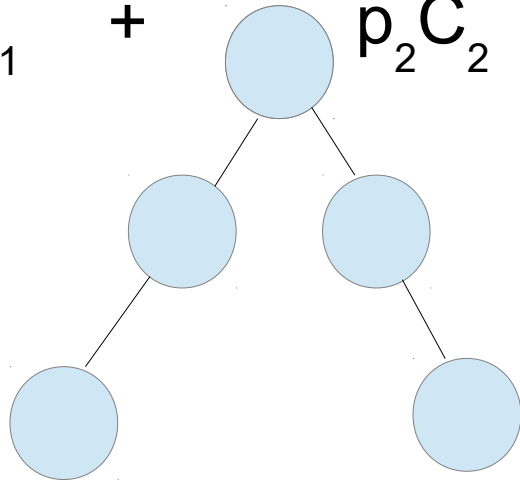
Risk scenario 1

$$\mathbb{E}(C) = p_1 C_1$$



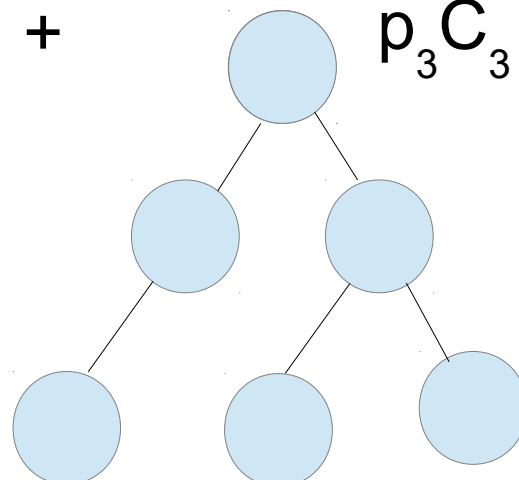
Risk scenario 2

$$+ p_2 C_2$$



Risk scenario 3

$$+ p_3 C_3$$



$$\mathbb{V}_{\text{abs}}(C) = \sum_s p_s |(\mathbb{E}(C) - C_s)|$$

Robust optimization

Objective function:

$$\text{obj: Min } \mathbb{E}(C) + \gamma \mathbb{V}_{\text{abs}}(C) + \omega \sum_s p_s \alpha \delta_{ks}$$

δ_{ks} not satisfied demand

$$\sum_j \sum_n Q_{jkns} + \delta_{ks} \geq d_{ks}$$

d_{ks} : demand of customer zone k in scenario s

Q_{jkns} : quantity shipped from j to k by mode n in scenario s

Robust optimization

- For each risk scenario, a separate minimal flow is determined by the optimizer
- Optimizer is only applied once

What is not provided by this method:

- An optimal overall flow

Terminology

- Without considering risks: Deterministic optimization
- Considering risks but no variance: Stochastic optimization
- Considering risk and costs variance: Robust optimization

Questions

Do single source situations have a higher expected costs than dual source situations?

Yes, since single source situations lead more often to unsatisfied demands, which are penalized.

Pro / Cons



- Unsatisfied demand / group risks / single and multiple sourcing are treated properly
- Determines optimal flow for each risk scenario



- Rather slow caused by large amount of decision variables
- Determines no global overall optimal flow

Manager-defined scenarios

Supply chain expert specifies list of risk scenarios where each scenario contains of:

- probability of occurrence
- impact in form of supplier or transport disruption (example: production capacity is reduced by 80% due to this risk)

Automated scenario generation

- Supply chain expert specifies list of risk scenarios templates where each scenario template contains
- probability of occurrence
 - impact in form of supplier or transport disruption degree distribution (here Weibull) typically given by variance and expected value
 - Scenarios are then sampled using this distribution

Research questions

- How many automatically generated scenarios are needed for a precise estimate?
- What is the critical point (point of convergence) using automatically generated scenarios?
- What is the convergence rate?
- (Future) What is the gain to use the full distribution in contrast to just use the expected value? For that we need a non-linear model.

Setup

We are using a real-world supply chain with:

- 40 Customer

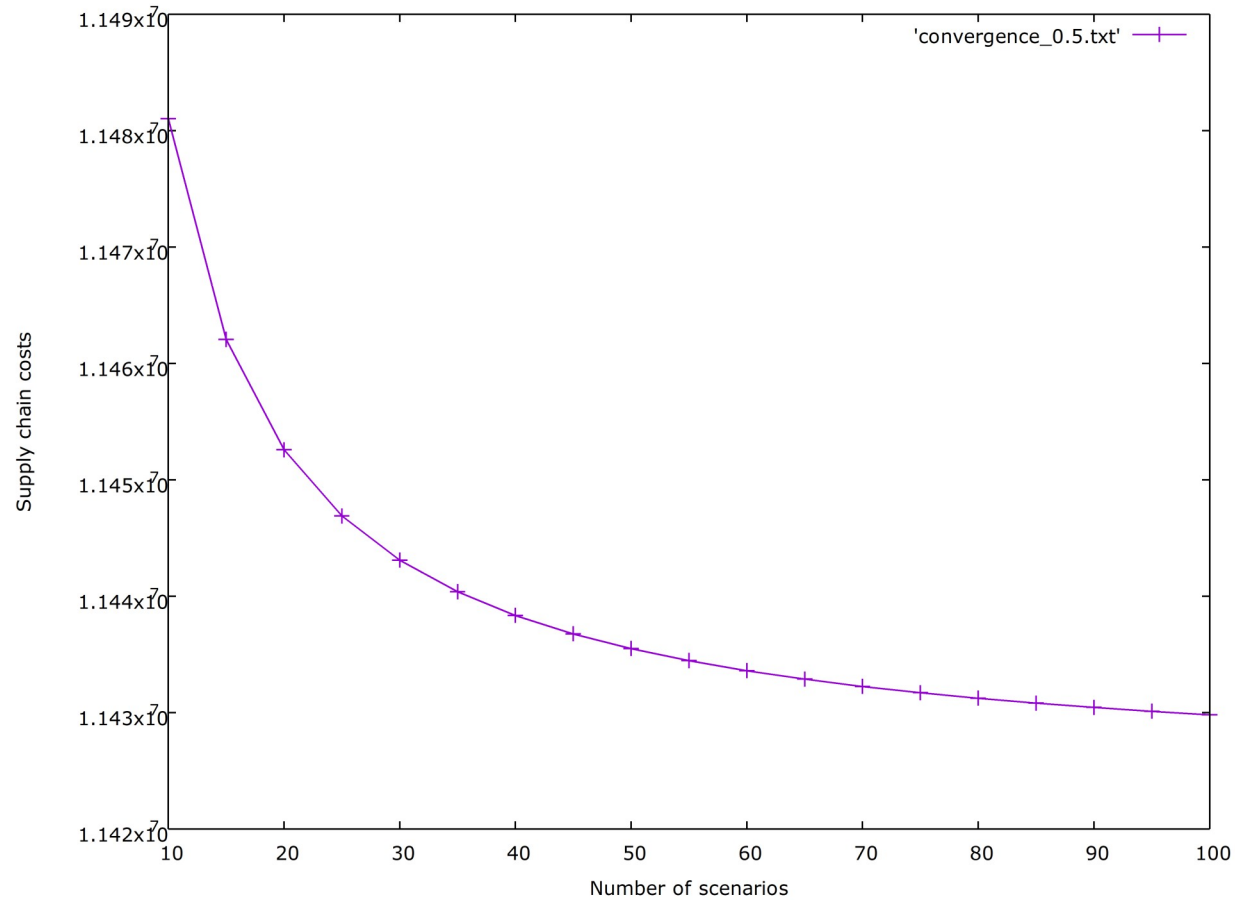
- 100 suppliers

- 230 Items

- 200 transport links

So far we employed only stochastic and not robust optimization for the automated scenario generation approach.

Costs convergence



Findings

Findings:

- Manager defined approach can be seen as special case of automated scenario generation approach with scenarios $\rightarrow \infty$ and variance $\rightarrow 0$
- so, costs obtained for manager-defined scenarios be seen as approximately the convergence point for the automated approach with small variance (here 0.05)
- with 100 scenarios, we still have an estimated error of 0.05 %
- converge rate: only sublinear, i.e. rather slow convergence

Error Estimation

- For the robust/stochastic optimization approach it is assumed that the risk scenarios are all disjoint:
- However in practice they are more likely to be independent.
- This can cause the sum of the probabilities to be too large, in extreme cases even exceed one.
- How much can the sum of probabilities deviate from the correct value?
- We can estimate this error using the formula of Sylvester.

Error Estimation

$$\begin{aligned}\text{Error} &= \left(\sum_{i=1}^n P(S_i) - \sum_{k=1}^n (-1)^{k-1} \sum_{1 \leq i_1 < \dots < i_k \leq n} P(S_{i_1} \cap \dots \cap S_{i_k}) \right) \\ &= \left(\sum_{i=1}^n P(S_i) - \sum_{k=1}^n (-1)^k \sum_{1 \leq i_1 < \dots < i_k \leq n} P(S_{i_1}) \dots P(S_{i_k}) \right)\end{aligned}$$

S_i : risk scenario i

$P(S_i)$: probability of risk scenario i

In our setup:

Absolute error: 0.0001

Relative error: 0.0067

Conclusion

- Robust optimization is a powerful approach for dealing with supply chain risks
- It allows for considering partial breakdowns as well as group risks
- The number of automatically generated scenarios required for reliable cost estimates in our example setup is >100 and the costs converges only sublinear
- Performance of single period robust optimization model is high for our example setup with 8 scenarios (only several seconds)

Future Work

- Use automatic scenario generation together with nonlinear models (for instance: models containing integer constraints) and robust optimization (currently only stochastic optimization)
- Use of a multiperiod robust optimization model
- Combining topological risk models with a robust optimization model

Bibliography

- [1] Reza Babazadeh and Jafar Razmi. A robust stochastic programming approach for agile and responsive logistics under operational and disruption risks. *International Journal of Logistics Systems and Management*, 2012.
- [2] Christopher W. Craighead, Jennifer Blackhurst, and M. Johnny Rungtusanatham. The severity of supply chain disruptions: Design characteristics and mitigation capabilities. *Decision Sciences*, 38(1), 2007.
- [3] Mauro Falasca, Christopher W. Zobel, and Deborah Cook. A decision support framework to assess supply chain resilience. In *Proceedings of the 5th International ISCRAM Conference*, Washington, DC, 2008.
- [4] Mark Treleven, Sharon Bergman, and Schweikhart. A risk/benefit analysis of sourcing strategies: Single vs. multiple sourcing. *Journal of Operations Management*, 7(4), 1988.