

Global Chance-Constraints: an Application to Stochastic Inventory Control

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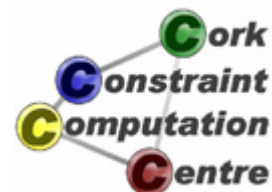
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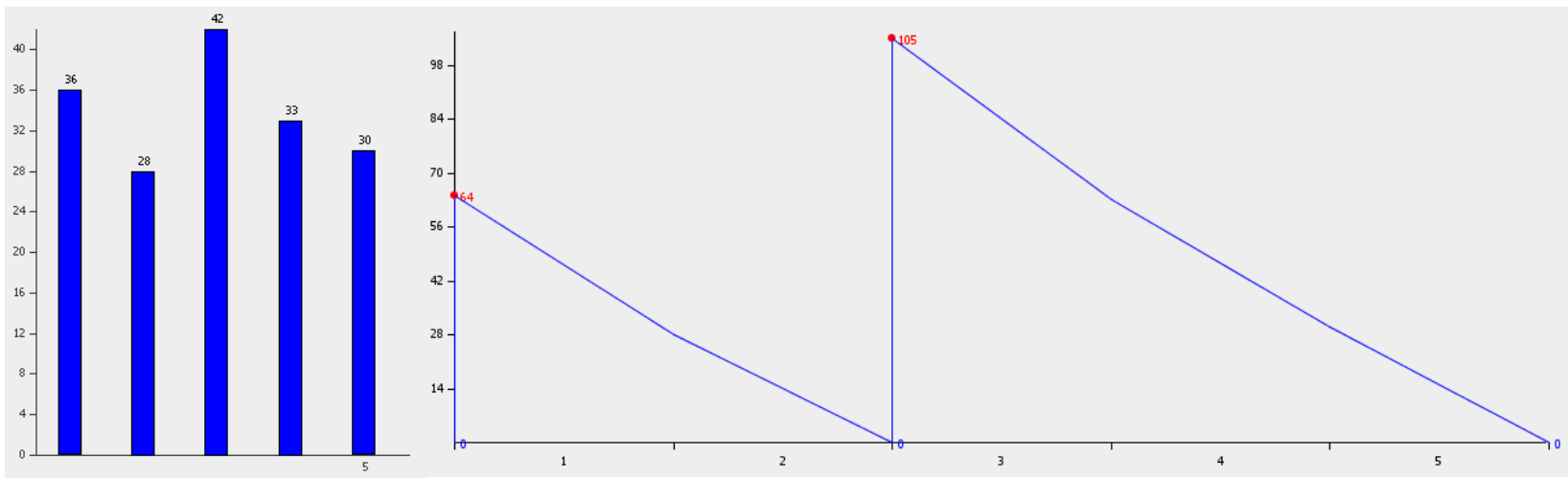
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Uncapacitated Deterministic Inventory Control

- Finding an optimal order policy to meet the demand over a given planning horizon minimizing costs
 - Holding cost: 1
 - Fixed ordering cost: 100 (the unit variable cost can be ignored!)
- Can be solved in Polynomial time using Wagner and Whitin (Shortest Path) algorithm [Wagner et al. - 58]



Capacitated Deterministic Inventory Control

- NP-Hard [Florian et al. - 80]: reduction from Subset Sum
- Their proof can be extended to the uncapacitated case under particular cost functions
 - A convex objective function with concave holding cost and fixed ordering cost can be interpreted as a capacity constraint

Summary of Complexity Results

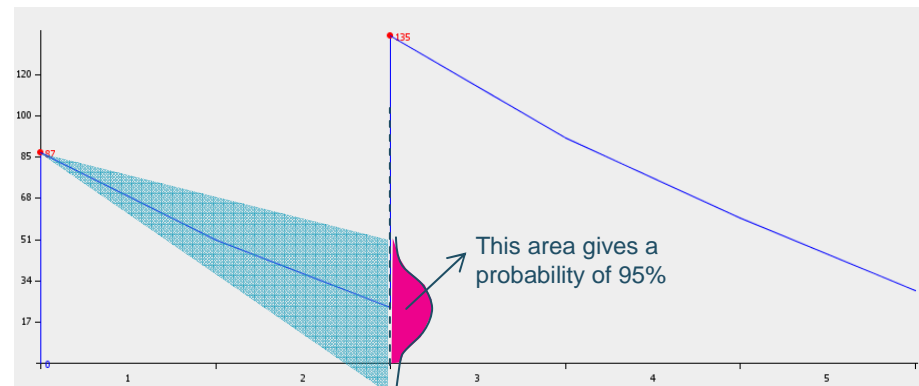
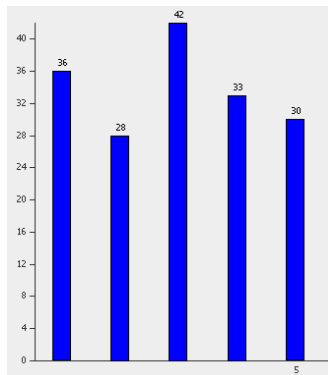
arbitrary cost functions		capacity limits			convex cost functions
		infinite	equal	arbitrary	
set-up costs	zero	†	†	†	*
	equal	†	†	†	†
	arbitrary	†	†	†	†
concave cost functions		*	*	†	

* solvable in polynomial time

† solvable in pseudopolynomial time, and *NP*-hard even for equal demands and zero storage costs
(*x*) see §*x*.

Stochastic Inventory Control under Service Level Constraint

- PSPACE – Stochastic Constraint Programming [Walsh - 02]
- Reduction from TQBF $\exists x_1 \forall x_2 \exists x_3 \forall x_4 : (x_1 \vee \neg x_3 \vee x_4) \wedge (\neg x_2 \vee x_3 \vee \neg x_4)$
- Finding an optimal order policy to meet the demand over a given planning horizon minimizing costs and meeting the required service level chance constraint
 - Holding cost: 1
 - Fixed ordering cost: 100 (the unit variable cost cannot be ignored in general!)
 - Service level: 95%
 - Coeff of variation: 0.3



Stochastic Inventory Control under Service Level Constraint

- Chance-constrained model:

$$\begin{aligned} & \text{minimise } E\{\text{TC}\} \\ & = \int_{d_1} \int_{d_2} \cdots \int_{d_N} \sum_{t=1}^N (a\delta_t + hI_t + vX_t) \\ & \quad \times g_1(d_1)g_2(d_2)\dots g_N(d_N)d(d_1)d(d_2)\dots d(d_N) \quad (1) \end{aligned}$$

subject to

$$\delta_t = \begin{cases} 1 & \text{if } X_t > 0, \\ 0 & \text{otherwise,} \end{cases} \quad t = 1, \dots, N, \quad (2)$$

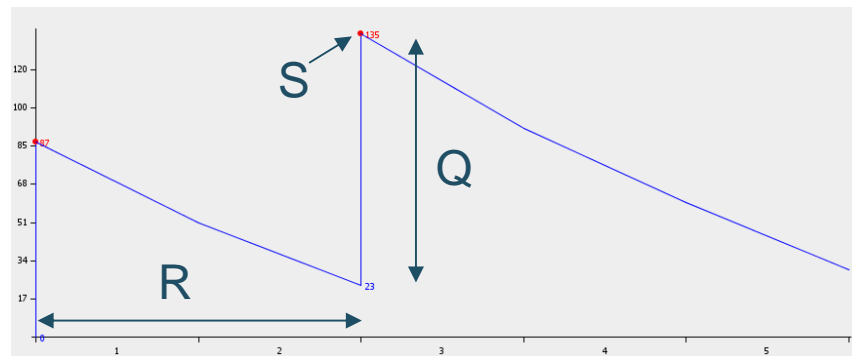
$$I_t = I_0 \sum_{i=1}^t (X_i - d_i), \quad t = 1, \dots, N, \quad (3)$$

$$\Pr\{I_t \geq 0\} \geq \alpha, \quad t = 1, \dots, N, \quad (4)$$

$$X_t, I_t \geq 0, \quad t = 1, \dots, N,$$

Stochastic Inventory Control under Service Level Constraint

- Different strategies (policies) to model multistage stochastic problems:
 - Observation first
 - Decision first
 - Hybrid strategies
- (R^n, S^n) policy is hybrid: **order-up-to-level** as deterministic decisions



Stochastic Inventory Control under Service Level Constraint

- First complete method for (R^n, S^n) : CP approach [Tarim et al. 05]
 - certainty equivalent model (NP-Hardness due to buffer constraints)
 - negative order quantities classified as rare events and therefore ignored

$$\min E[TC] = \sum_{t=1}^N (a\delta_t + h\tilde{I}_t)$$

$$\text{s.t. } t = 1, \dots, N$$

$$\tilde{I}_t + \tilde{d}_t - \tilde{I}_{t-1} \geq 0$$

$$\tilde{I}_t + \tilde{d}_t - \tilde{I}_{t-1} > 0 \Rightarrow \delta_t = 1$$

$$\tilde{I}_t \geq \Phi[t, \max_{j \in [1..t]} j\delta_j]$$

$$\tilde{I}_t \in \mathbf{Z}^+ \cup \{0\} \quad \delta_t \in \{0,1\}$$

Cumulative distribution function

$$\Phi[i, j] = G_{d_j + d_{j+1} + \dots + d_i}^{-1}(\alpha) - \sum_{k=j}^i \tilde{d}_k$$

Stochastic Inventory Control under Service Level Constraint

- First complete method
 - certainty equivalent model
 - negative order quantities ignored

Identifies the correct **deterministic buffer stock level** needed to satisfy the required service level for a given replenishment cycle

$$\min E[TC] = \sum_{t=1}^N (a\delta_t + b\tilde{I}_t)$$

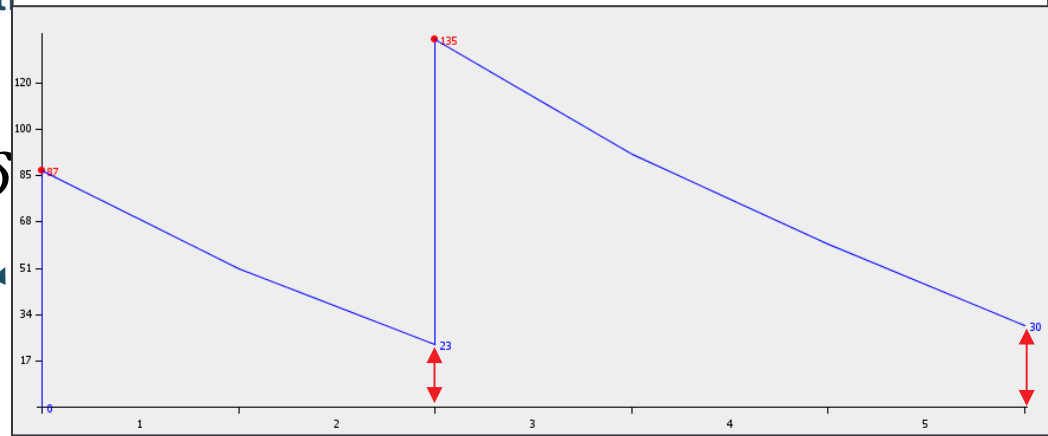
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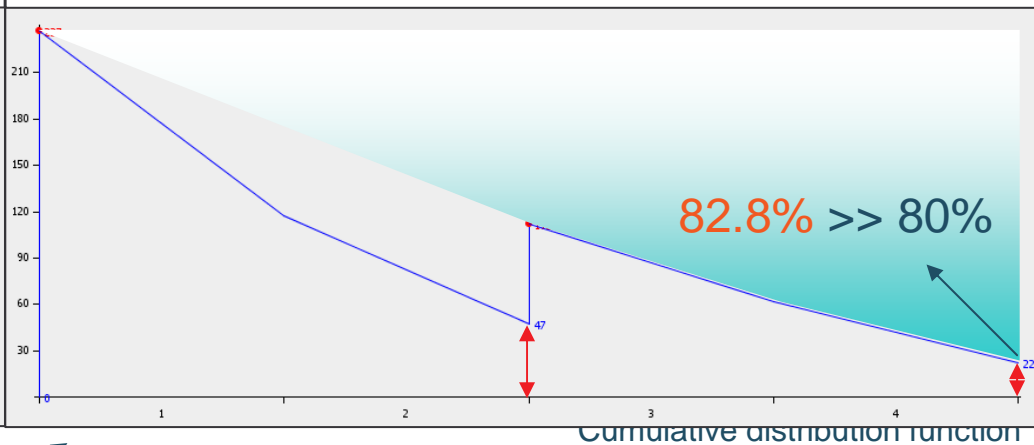
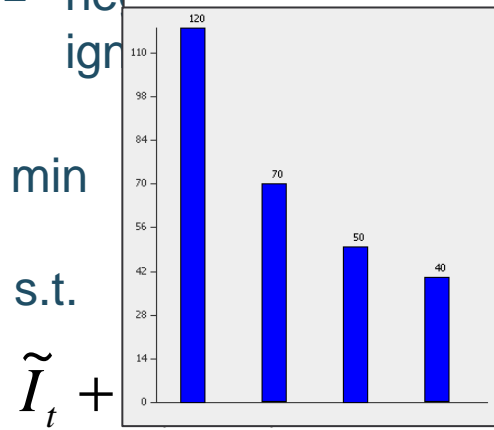


$$\Phi[i, j] = G_{d_j + d_{j+1} + \dots + d_i}^{-1}(\alpha) - \sum_{k=j}^i \tilde{d}_k$$

Stochastic Inventory Control under Service Level Constraint

- First order policy
 - cost: $O_c = 150$, $H_c = 1$
 - negative inventory: $\text{Service} = 80\%$
 - ignore: $C_v = 0.4$

Ignores the effect of *negative order quantities*:
 Effect on the *cost*
 Effect on the *service level* of following periods



$$\tilde{I}_t + \tilde{d}_t - \tilde{I}_{t-1} > 0 \Rightarrow \delta_t = 1$$

$$\tilde{I}_t \geq \Phi[t, \max_{j \in [1..t]} j \delta_j]$$

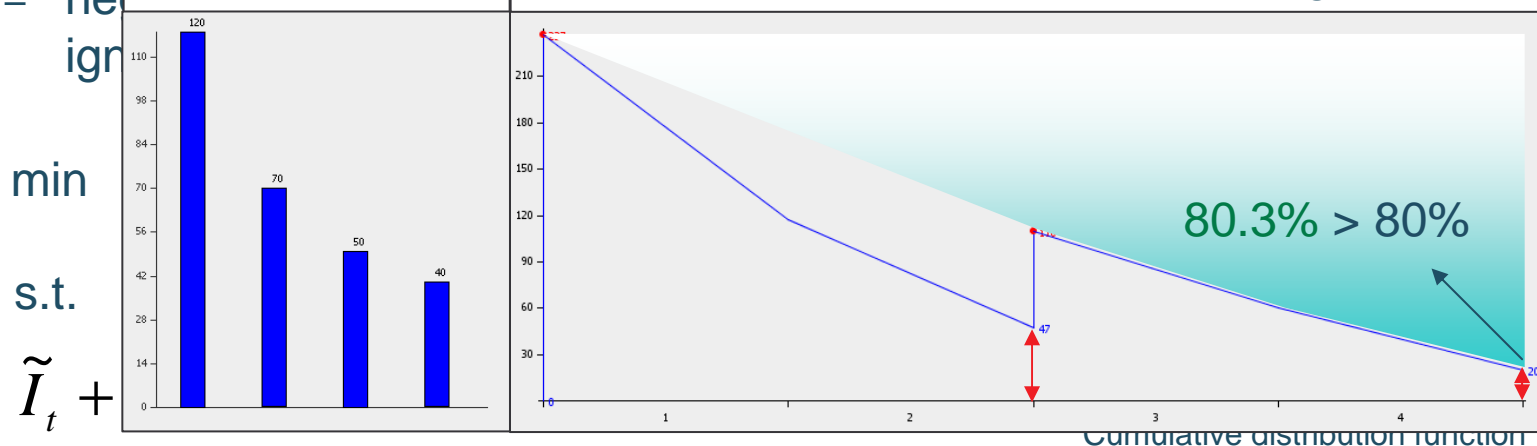
$$\tilde{I}_t \in \mathbb{Z}^+ \cup \{0\} \quad \delta_t \in \{0,1\}$$

$$\Phi[i, j] = G_{d_j + d_{j+1} + \dots + d_i}^{-1}(\alpha) - \sum_{k=j}^i \tilde{d}_k$$

Stochastic Inventory Control under Service Level Constraint

- First order quantity
 - cost $O_c = 150$
 - holding $H_c = 1$
 - negative $\text{Service} = 80\%$
 - ignored $C_v = 0.4$

Ignores the effect of *negative order quantities*:
 Effect on the *cost*
 Effect on the *service level* of following periods

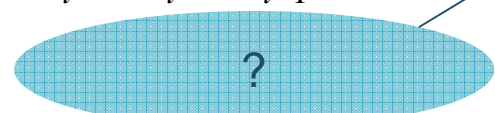


min

s.t.

$$\tilde{I}_t +$$

$$\tilde{I}_t + \tilde{d}_t - \tilde{I}_{t-1} > 0 \Rightarrow \delta_t = 1$$

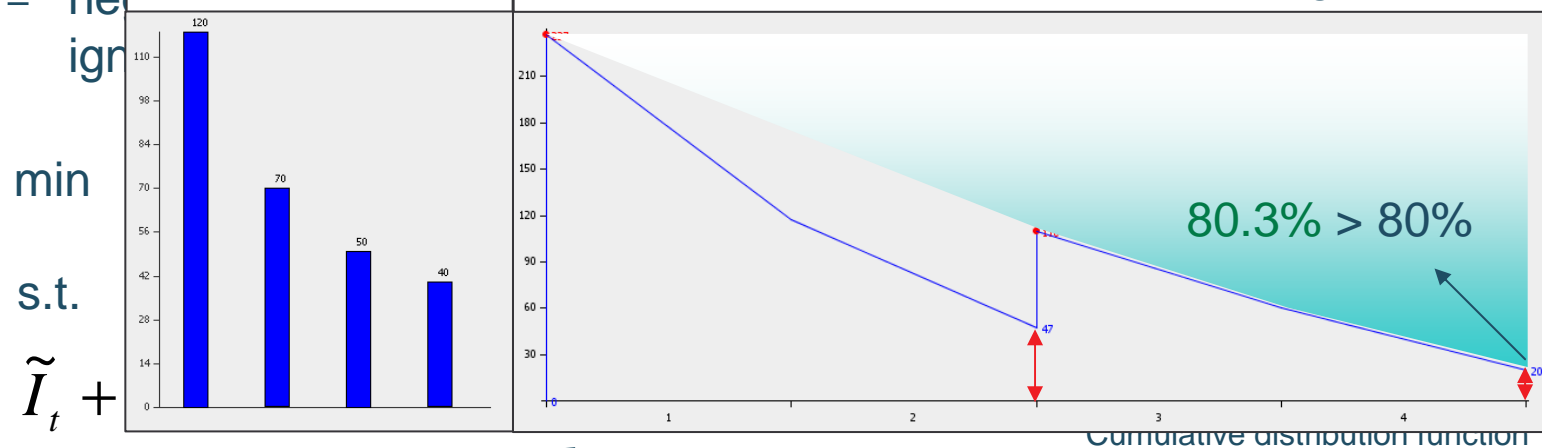


$$\tilde{I}_t \in \mathbb{Z}^+ \cup \{0\} \quad \delta_t \in \{0,1\}$$

$$\Phi[i, j] = G_{d_j + d_{j+1} + \dots + d_i}^{-1}(\alpha) - \sum_{k=j}^i \tilde{d}_k$$

Stochastic Inventory Control: Global Chance-Constraint

- First order
 - ce $O_c = 150$
 - ne $H_c = 1$
 - ign $Service = 80\%$
 - $C_v = 0.4$
- Ignores the effect of *negative order quantities*:
Effect on the *cost*
Effect on the *service level* of following periods



min

s.t.

$\tilde{I}_t +$

$$\tilde{I}_t + \tilde{d}_t - \tilde{I}_{t-1} > 0 \Rightarrow \delta_t = 1$$

serviceLevel($\tilde{I}, \delta, d, \alpha$)

$$\tilde{I}_t \in \mathbb{Z}^+ \cup \{0\} \quad \delta_t \in \{0,1\}$$

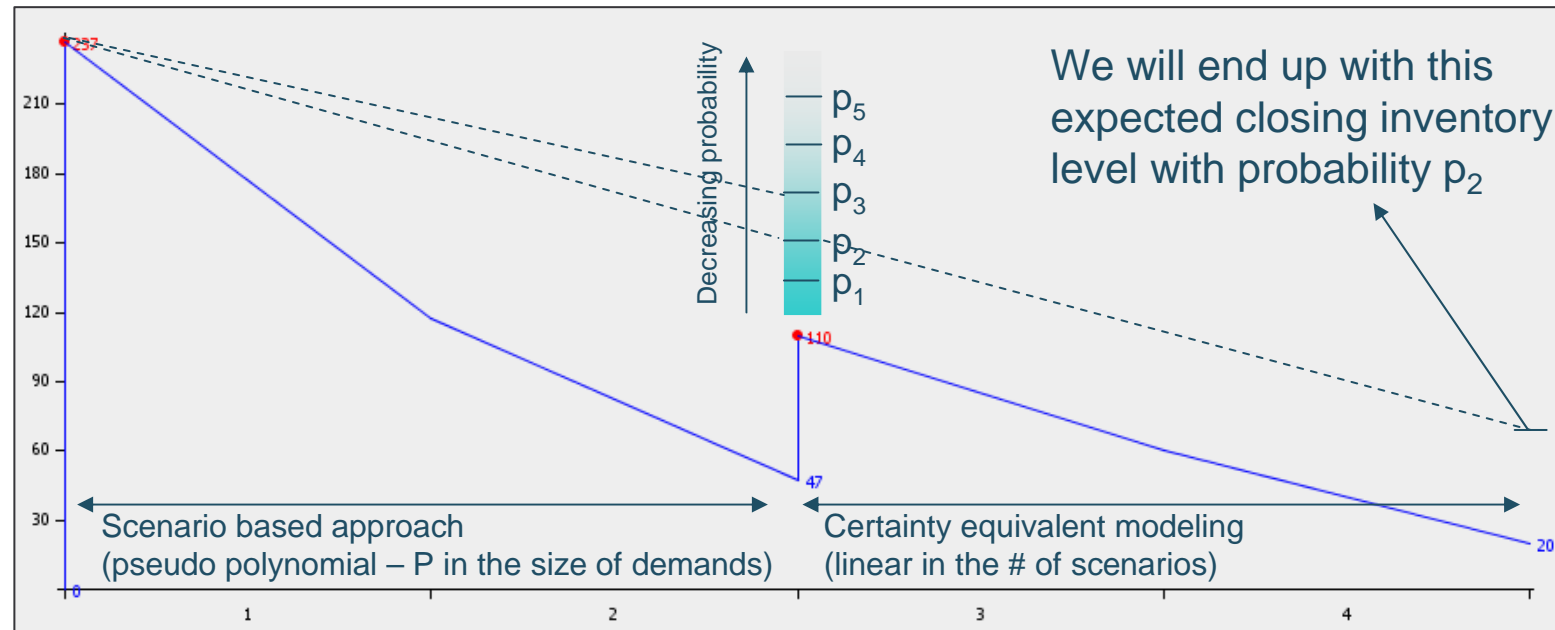
$$\Phi[i, j] = G_{d_j + d_{j+1} + \dots + d_i}^{-1}(\alpha) - \sum_{k=j}^i \tilde{d}_k$$

Stochastic Inventory Control: Global Chance-Constraint

- $serviceLevel(\tilde{I}, \delta, d, \alpha)$ assures that for each replenishment period, at the end of each and every time period, the probability the net inventory will not be negative is at least alpha
 - \tilde{I}, δ are arrays of deterministic decision variables
 - d is an array of random variables with p.d.f. $g_i(d_i)$
 - α is the required threshold probability
 - It is therefore semantically equivalent to the original set of chance constraints $\Pr\{I_t \geq 0\} \geq \alpha, t \in \{1, \dots, N\}$
- The interaction between consecutive replenishment cycles is considered, we are therefore solving the original problem formulation in PSPACE and not a relaxed NP-Hard problem.

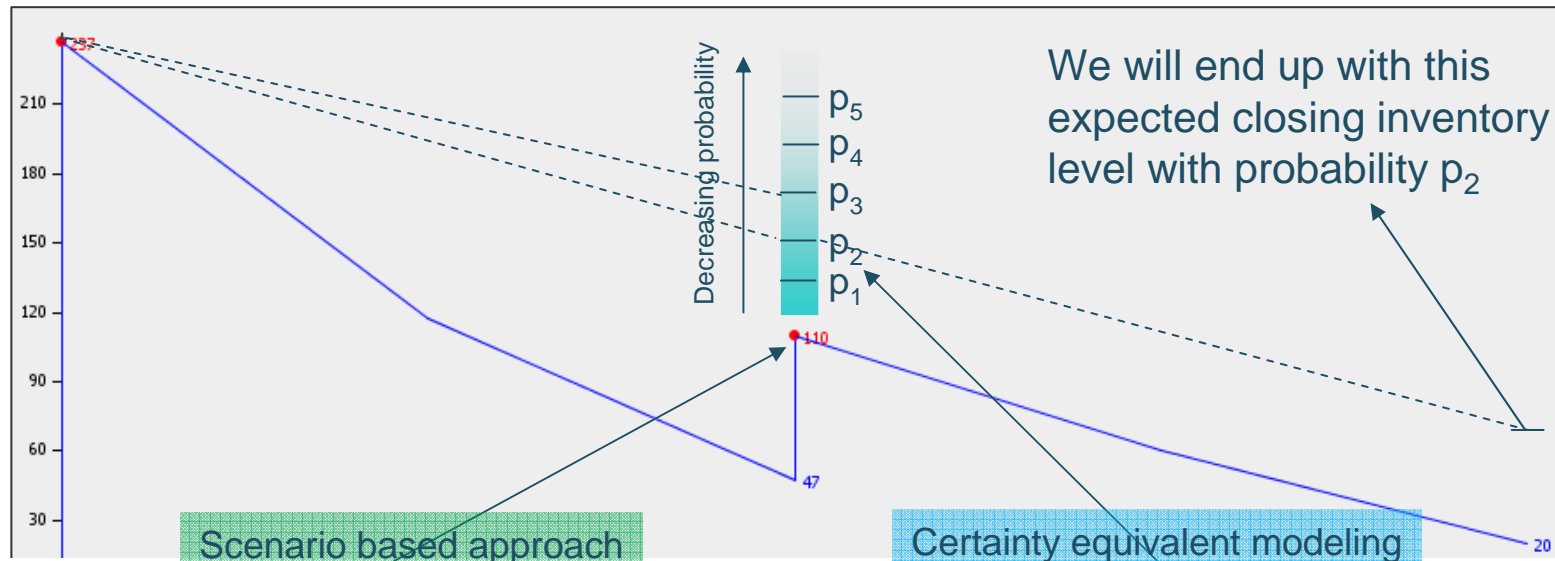
Stochastic Inventory Control: Global Chance-Constraint

- $serviceLevel(\tilde{I}, \delta, d, \alpha)$ computes at each node of the search tree a **convolution integral** that considers joint probabilities of different scenarios which contribute to the overall service level...



Stochastic Inventory Control: Global Chance-Constraint

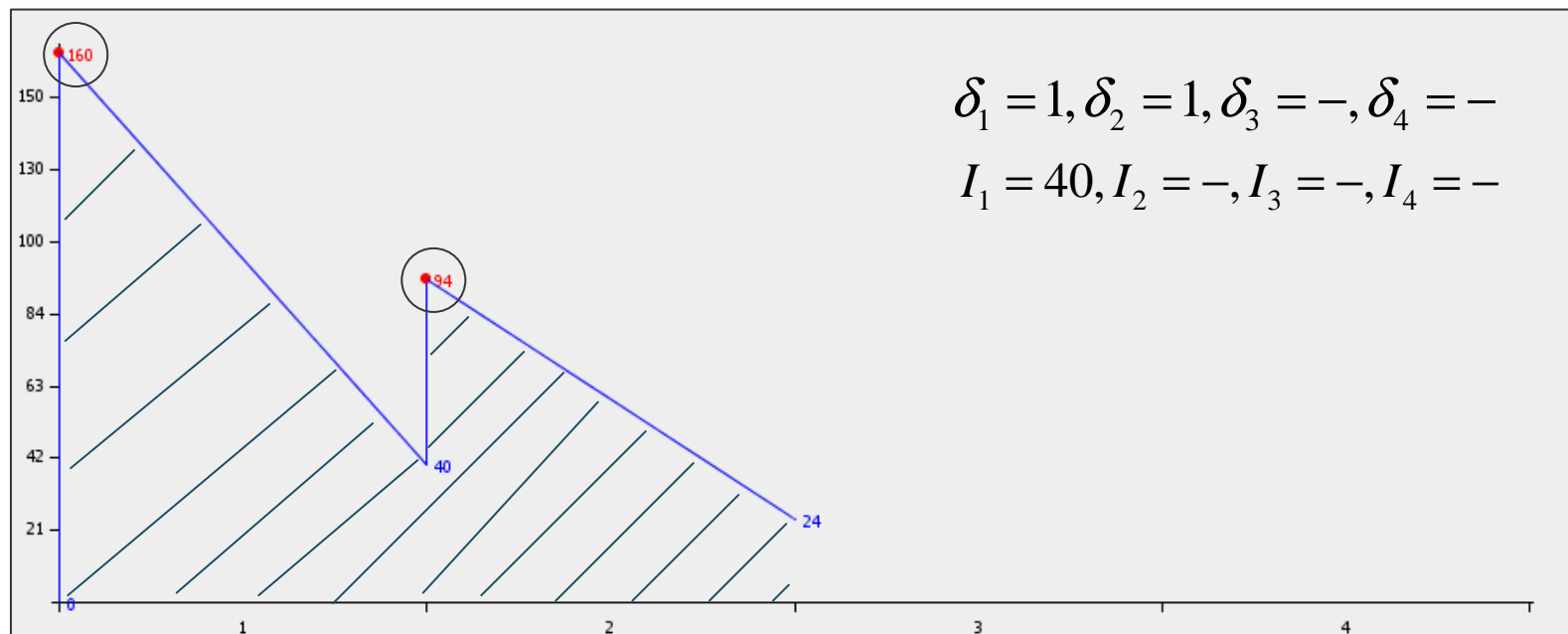
- $serviceLevel(\tilde{I}, \delta, d, \alpha)$...is therefore able to check if a given configuration for the \tilde{I}, δ variables is feasible w.r.t. the required service level constraint and to enforce the condition during search



$$\Pr\{d_1 + d_2 > S_1 - S_2\} \square G_{d_3+d_4}(S_2) + \sum_{i=0}^{S_1-S_2} \left(\Pr\{d_1 + d_2 = i\} \square G_{d_3+d_4}(S_1 - i) \right) \geq \alpha$$

Stochastic Inventory Control: Global Chance-Constraint

- $serviceLevel(\tilde{I}, \delta, d, \alpha)$...is able to detect infeasible partial assignments for \tilde{I}, δ variables and to filter domains.



Global Chance-Constraint: Future Extensions

- $serviceLevel(\tilde{I}, \delta, d, \alpha)$
- More effective domain filtering methods and forward checking strategies can be developed
 - **forward checking based on scenario probabilities**: fails when, observing current assignments, the remaining decision variable assignments are not able to add up to the required threshold: “Stochastic Constraint Programming”, [Walsh – 02]
 - **linear dependency between closing inventory levels for non-replenishment periods**: not all the former closing inventory levels have to be ground. For each replenishment cycle we require only that at least one \tilde{I}_t variable is ground
 - **cost based filtering by relaxation**: the relaxation can be a simple problem (**P**) or again a difficult problem (**NP-hard**), which is easier than our original problem (**PSPACE**). Intuitively the more difficult is the relaxation the tighter will be the bound obtained

Conclusions

- We described an application of Stochastic Constraint Programming in Inventory Control
- We developed global chance-constraints to enforce multiple chance-constraints, when these are mutually dependent in multistage problems
- We showed how to express these complex relations by merging certainty equivalent modeling and scenario based approach during the search
- We suggested strategies to improve the efficiency of this approach in our Inventory Control application