

Dynamic uncapacitated lot sizing with random demand under a fillrate constraint

Horst Tempelmeier and Sascha Herpers

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Agenda

- 1 Introduction
 - The Problem
 - Solution Approaches
- 2 Optimization Model
 - Formulation
- 3 Solution approaches
 - Exact solution
 - Heuristic solution
- 4 Numerical Results
 - Experiment 1
 - Experiment 2
- 5 Conclusion

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Planning situation

Dynamic and Random Demand

- Demand (forecasted averages and variations)

Period t	25/2009	26/2009	...	35/2009
μ_t
σ_t

- Holding costs
- Setup costs
- Service level

Alternatives

- **Common sense approach (MRP, APS)**
Compute safety stocks and add to forecasted demand
- **(s_t, q_t) -policy, (r_t, S_t) -policy**
Use a **stationary** inventory policy with dynamic adjustment of parameters
- **"Static-dynamic uncertainty" strategy**
Fix replenishment **periods** in advance, adjust production quantity
- **"Static uncertainty" strategy**
Fix replenishment **periods** and **quantities** in advance

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Model Formulation I

Model **SSIULSP** $_{\beta_c}^q$:

$$\text{Minimize } Z = \sum_{t=1}^T (s \cdot \gamma_t + h \cdot E \{ [I_t]^+ \}) \quad (1)$$

s. t.

$$I_{t-1} + q_t - D_t = I_t \quad t = 1, 2, \dots, T \quad (2)$$

$$q_t - M \cdot \gamma_t \leq 0 \quad t = 1, 2, \dots, T \quad (3)$$

$$I_t^{f,\text{prod}} = - [I_{t-1} + q_t]^- \quad t = 1, 2, \dots, T \quad (4)$$

$$I_t^{f,\text{end}} = - [I_t]^- \quad t = 1, 2, \dots, T \quad (5)$$

$$F_t = I_t^{f,\text{end}} - I_t^{f,\text{prod}} \quad t = 1, 2, \dots, T \quad (6)$$

Model Formulation II

$$l_t = (l_{t-1} + 1) \cdot (1 - \gamma_t) \quad t = 1, 2, \dots, T \quad (7)$$

$$l_0 = -1 \quad (8)$$

$$\omega_t = \gamma_{t+1} \quad t = 1, 2, \dots, T - 1 \quad (9)$$

$$\omega_T = 1 \quad (10)$$

$$1 - \frac{E \left\{ \sum_{j=t-l_t}^t F_j \right\}}{E \left\{ \sum_{j=t-l_t}^t D_j \right\}} \geq \beta_c^* \quad t \in \{t \mid \omega_t = 1\} \quad (11)$$

Symbols used I

β_c^*	target fillrate
D_t	demand in period t (random variable)
F_t	backorder in period t (random variable)
γ_t	binary setup indicator in period t
h	inventory holding cost
l_t	net inventory at the end of period t (random variable)
$l_t^{f,\text{end}}$	backlog at the end of period t (random variable)
$l_t^{f,\text{prod}}$	backlog immediately after production in period t (random variable)
l_t	number of periods since the last setup prior to period t
M	large number

Symbols used II

ω_t	indicator variable: $\omega_t = 1$, if production takes place in period $t + 1$; $\omega_t = 0$, otherwise
q_t	production quantity in period t
s	setup cost
T	length of planning horizon
$[x]^+$	$= \max\{0, x\}$
$[x]^-$	$= \min\{0, x\}$

Expected Inventory

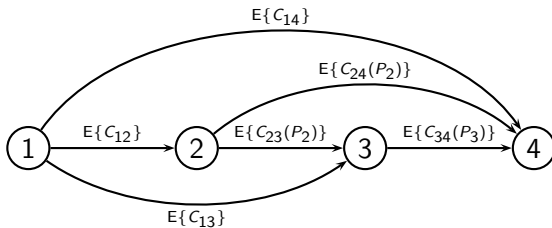
$$\begin{aligned}
 E\{I_t^p\} &= \int_0^{Q^{(t)}} (Q^{(t)} - y) \cdot f_{Y^{(t)}}(y) \cdot dy \\
 &= Q^{(t)} - E\{Y^{(t)}\} + G_{Y^{(t)}}^1(Q^{(t)}) \quad t = 1, 2, \dots \quad (14)
 \end{aligned}$$

$Q^{(t)}$ – cumulated production quantity from period 0 to t

$Y^{(t)}$ – cumulated demand from period 0 to t

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Shortest-Path Network



$$E\{C_{\tau t}\} = s + h \cdot \sum_{\ell=\tau}^{t-1} E \left\{ \left[I_{\tau-1}(P_{\tau}) + q_{\tau t}^* - \sum_{i=\tau}^{\ell} D_i \right]^+ \right\} \quad (15)$$

Shortest-Path Network

	Setup in period					On hand inventory
1	2	3	4	5		$E\{I_5^P\}$
1	0	0	0	0		79.61
1	1	0	0	0		81.14
1	1	1	0	0		87.01
1	1	1	1	0		103.89
1	1	1	1	1		118.95
1	0	1	0	0		86.95
⋮	⋮	⋮	⋮	⋮		⋮

Table: Expected on-hand inventory at the end of period 5 as a function of the setup pattern

Solution Procedure

```

1:  $\mathcal{M} := \mathcal{U} := \{2, 3, \dots, T\}$ 
2:  $\mathcal{E} := \{(\tau, t) | \tau = 1, 2, \dots, T; t = \tau + 1, \tau + 2, \dots, T - 1\}$ 
3: for all  $((\tau, t) \in \mathcal{E} \text{ with } t \in \mathcal{U})$  do
4:     Predecessor( $t$ ):= 1;  $C(t) := E\{C_{1t}\}$ 
5: end for
6: while  $(\mathcal{M} \neq \emptyset)$  do
7:     Select  $\tau \in \mathcal{M}$  with minimum  $C(\tau)$ 
8:      $\mathcal{M} := \mathcal{M} \setminus \tau$ ;  $\mathcal{U} := \mathcal{U} \setminus \tau$ 
9:     if  $(\tau = T)$  then
10:         end
11:     else
12:         for all  $((\tau, t) \in \mathcal{E} \text{ with } t \in \mathcal{U})$  do
13:              $\mathcal{M} := \mathcal{M} \cup t$ 
14:             if  $(C(\tau) + E\{C_{\tau t}\} < C(t))$  then
15:                 Predecessor( $t$ ):=  $\tau$ 
16:                  $C(t) := C(\tau) + E\{C_{\tau t}\}$ 
17:             end if
18:         end for
19:     end if
20: end while
    
```

Dynamic Lot Sizing Heuristic

```
1:  $\tau := 1$ 
2: while ( $\tau < T$ ) do
3:    $t := \tau$ 
4:   while ( $t < T$ ) do
5:     if ( $C_{\tau t} \leq C_{\tau, t+1}$ ) then
6:        $t := t + 1$ 
7:     else
8:       Make current lotsize for period  $\tau$  permanent.
9:        $\tau := t + 1$ 
10:    end if
11:  end while
12: end while
```


Silver-Meal Rule

$$E\{C_{\tau t}\} = \frac{s + h \cdot \sum_{\ell=\tau}^t E \left\{ \left[I_{\tau-1}(P_{\tau-1}) + q_{\tau t}^* - \sum_{i=\tau}^{\ell} D_i \right]^+ \right\}}{t - \tau + 1} \quad (16)$$

Least-Unit-Cost rule

$$E\{C_{\tau t}\} = E \left\{ \frac{s + h \cdot \sum_{\ell=\tau}^t \left[I_{\tau-1}(P_{\tau-1}) + q_{\tau t}^* - \sum_{i=\tau}^{\ell} D_i \right]^+}{\sum_{i=\tau}^t D_i} \right\} \quad (17)$$

Least-Total-Cost rule

$$E\{C_{\tau t}\} = E \left\{ s + h \cdot \sum_{\ell=\tau}^t \left[I_{\tau-1}(P_{\tau-1}) + q_{\tau t}^* - \sum_{i=\tau}^{\ell} D_i \right]^+ \right\} \quad (18)$$

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Expected Demands

Series #	$E\{D_t\}$												
1	92	92	92	92	92	92	92	92	92	92	92	92	93
2	80	100	125	100	50	50	100	125	125	100	50	100	
3	50	80	180	80	0	0	180	150	10	100	180	95	
4	10	10	15	20	70	180	250	270	230	40	0	10	

Parameters

Series #	T	s	TBO	CV_D	β_c^*
1	12	500	1–12	{0.1, 0.2, 0.3, 0.4}	{0.5, 0.525, 0.05, ..., 0.975}
2	12	500	1–12	{0.1, 0.2, 0.3, 0.4}	{0.5, 0.525, 0.05, ..., 0.975}
3	12	500	1–12	{0.1, 0.2, 0.3, 0.4}	{0.5, 0.525, 0.05, ..., 0.975}
4	12	500	1–12	{0.1, 0.2, 0.3, 0.4}	{0.5, 0.525, 0.05, ..., 0.975}

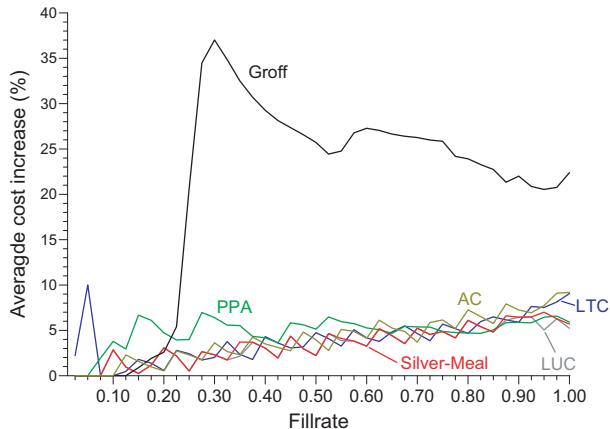
Results

Series #	Heuristic	Average cost increase (%)	Maximum cost increase (%)	% Optimal	% Worst
1	LUC	4.9	30.4	56.6	3.76
	SM	5.0	30.4	56.0	3.97
	LTC	5.3	29.1	48.6	7.00
	PPA	5.8	43.5	57.8	19.23
	AC	6.1	30.1	48.6	7.52
	Groff	24.5	52.7	3.9	75.24
2	SM	5.9	32.7	42.8	4.79
	LTC	7.2	28.7	41.5	5.31
	LUC	7.5	31.3	39.9	5.83
	PPA	7.9	49.9	51.4	17.50
	AC	7.9	32.9	38.3	5.31
	Groff	24.8	60.2	0.4	69.90

Results

Series #	Heuristic	Average cost increase (%)	Maximum cost increase (%)	% Optimal	% Worst
3	AC	9.7	37.0	35.4	3.02
	PPA	10.2	55.8	46.0	3.96
	SM	10.7	36.6	29.5	1.04
	LTC	11.5	41.1	33.2	5.83
	Groff	23.2	58.9	10.3	33.13
	LUC	29.0	67.9	1.8	54.69
4	SM	6.0	30.9	49.9	1.88
	AC	13.2	50.3	27.8	0.73
	PPA	15.3	59.0	34.2	14.58
	LTC	15.8	52.3	22.5	2.19
	Groff	17.1	47.6	14.3	7.81
	LUC	39.1	61.2	3.4	74.48

Results for Demand Series 1



Parameters

$$E\{D_t\} \simeq \text{Uniform}(0, 100)$$

Series #	T	s	TBO	CV_D	β_c^*
5	5	500	1–5	{0.1, ..., 0.4}	{0.5, 0.525, 0.05, ..., 0.975}
6	10	500	1–5	{0.1, ..., 0.4}	{0.5, 0.525, 0.05, ..., 0.975}
7	15	500	1–15	{0.1, ..., 0.4}	{0.5, 0.525, 0.05, ..., 0.975}
8	20	500	1–15	{0.1, ..., 0.4}	{0.5, 0.525, 0.05, ..., 0.975}

Results

Series #	Heuristic	Average cost increase (%)	Maximum cost increase (%)	% Optimal	% Worst
5	SM	4.7	35.4	65.4	14.47
	AC	5.6	39.7	61.2	13.07
	LTC	5.9	63.4	62.6	11.16
	LUC	10.5	64.4	50.6	25.25
	Groff	13.8	60.5	34.7	36.77
	PPA	15.1	56.7	38.6	42.72
6	SM	4.2	32.7	43.7	3.75
	AC	10.2	41.2	21.6	10.05
	LTC	11.5	44.7	16.5	13.35
	Groff	13.8	70.5	13.9	27.95
	LUC	15.4	55.3	12.6	21.25
	PPA	19.9	60.4	12.2	43.78

Results

Series #	Heuristic	Average cost increase (%)	Maximum cost increase (%)	% Optimal	% Worst
7	LTC	8.5	42.8	40.4	3.14
	AC	8.6	44.8	39.5	2.26
	PPA	9.1	64.9	53.1	15.04
	LUC	11.0	53.9	35.3	8.18
	SM	12.1	40.3	25.6	5.96
	Groff	30.3	64.4	2.3	68.47
8	SM	8.8	38.8	21.1	1.54
	AC	10.7	46.5	17.9	1.48
	LTC	11.6	46.2	17.0	3.64
	PPA	12.5	66.2	32.5	16.42
	LUC	13.9	49.4	14.7	9.33
	Groff	29.9	64.6	1.0	69.41

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Conclusion

- Exact solution for the stochastic Wagner-Whitin problem
- Adjusted cost criteria used in standard dynamic lot sizing heuristics
- Silver-Meal rule superior to Groff rule
- Directly applicable in ERP/AP systems
- Static uncertainty strategy: no nervousness, no bullwhip effect
- Target service level (instead of backorder costs)
- Possible extension: Capacities (done)