

The more, the better, usually: An analytical investigation of the yield rate paradox in closed-loop supply chains

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Motivation:

Sustainability Perspective: Aquafil Group

"the first rule in being sustainable is being profitable, because if you are not profitable as a company, sooner or later you go out of business, and yes, you aren't damaging the environment - but also you don't exist anymore."



Mr. Giulio Bonazzi, CEO and
President of Aquafil Group

https://www.econyl.com/assets/uploads/ElleMacArthur_case-study-aquafil-group.pdf

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Higher return (yield) positively impacts on the dynamics of CLSC

- [Tang and Naim, 2004]
- [Zhou and Disney, 2006]
- [Cannella et al., 2016]
- [Zhou et al., 2017]
- [Ponte et al., 2019]

Impacts of return yield may vary depending on parameters

- [Hosoda et al., 2015]

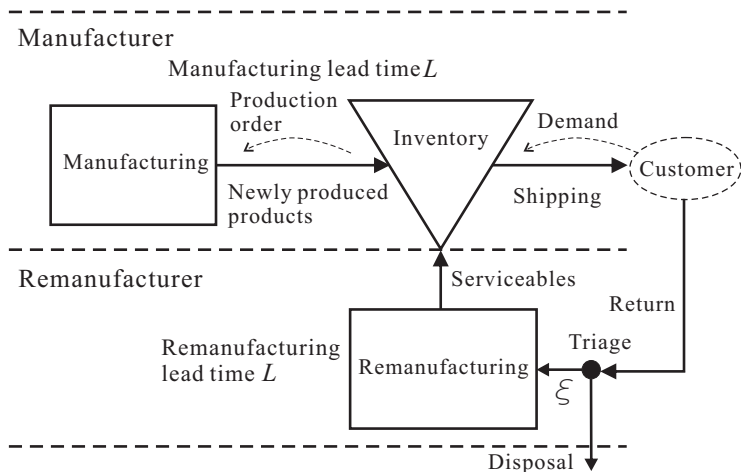
A comprehensive review of the CLSC literature:
[Akçalı and Çetinkaya, 2011], [Govindan et al., 2015],
[Cannella et al., 2016] and [Goltsos et al., 2018].

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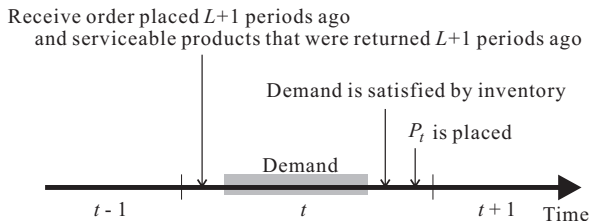
Model

Schematic of Model: Remanufactured products are considered as good as new.



Model

Sequence of Events and the Inventory Balance Equation at the Manufacturer



$$NS_t = NS_{t-1} + \underbrace{\xi}_{\text{Yield rate}} \underbrace{R_{t-(L+1)}}_{\text{Return}} + \underbrace{P_{t-(L+1)}}_{\text{Production order}} - \underbrace{D_t}_{\text{Demand}}$$

Serviceable products

where

- L : Manufacturing and remanufacturing lead times

Vector Autoregressive Demand and Return Model

$$\begin{bmatrix} D_t \\ R_t \end{bmatrix} = \begin{bmatrix} \mu_d \\ \mu_r \end{bmatrix} + \begin{bmatrix} \phi_d & 0 \\ \theta_r & \phi_r \end{bmatrix} \begin{bmatrix} D_{t-1} - \mu_d \\ R_{t-1} - \mu_r \end{bmatrix} + \begin{bmatrix} \varepsilon_{d,t} \\ \varepsilon_{r,t} \end{bmatrix}$$

- VAR model origin:
 - [Box and Tiao, 1977] [Tiao and Box, 1981]
- Application of VAR model to CL supply chain
 - [Hosoda and Disney, 2018]
- D_t : Demand at time period t
- R_t : Returns at time period t
- μ_d, μ_r : Mean of the demand and the returns, respectively
- ϕ_d, ϕ_r : Auto-correlation coefficients for the demand and the returns, respectively
- θ_r : Cross-correlation coefficient between returns and previous demand
- $\varepsilon_{d,t}, \varepsilon_{r,t}$: i.i.d. random variables with zero mean and constant SDs σ_d and σ_r , respectively
- Demand is not correlated with previous returns

Vector Autoregressive Demand and Return Model

Variances and Covariances

$$\mathbb{V}[D] = \frac{\sigma_{\varepsilon_d}^2}{1 - \phi_d^2}$$

$$\mathbb{V}[R] = \frac{\theta_r^2(1 + \phi_d\phi_r)\mathbb{V}[D]}{(1 - \phi_d\phi_r)(1 - \phi_r^2)} + \frac{\sigma_{\varepsilon_r}^2}{1 - \phi_r^2}$$

$$COV_0 = \mathbb{E}[(D_t - \mu_d)(R_t - \mu_r)] = \frac{\phi_d\theta_r\mathbb{V}[D]}{1 - \phi_d\phi_r}$$

$$COV_{L+1} = \mathbb{E}[(D_t - \mu_d)(R_{t-(L+1)} - \mu_r)] = \phi_d^{L+1} COV_0.$$

Closed-Loop Order-Up-To (CL-OUT) Policy

$$\begin{aligned} P_t = & \mathbb{E}[D_{t+L+1}] - \xi R_t \\ & + TNS - NS_t \\ & + \mathbb{E}\left[\sum_{i=1}^L D_{t+i}\right] - (WIP_t + WIPR_t), \end{aligned}$$

where

- $\mathbb{E}[D_{t+L+1}] - \xi R_t$: Forecast of products required at time $t + L + 1$.
- $TNS - NS_t$: Inventory feedback loop.
- $\mathbb{E}\left[\sum_{i=1}^L D_{t+i}\right] - (WIP_t + WIPR_t)$: Total WIP feedback loop.

CL-OUT Policy

Variances of the net stock levels $\mathbb{V}[NS]$ and production of new items $\mathbb{V}[P]$

$$\mathbb{V}[NS] = \frac{(L+1)(1-\phi_d^2) + \phi_d(1-\phi_d^{L+1})(\phi_d^{L+2} - \phi_d - 2)}{(1-\phi_d)^2(1-\phi_d^2)} \sigma_{\varepsilon_d}^2$$

$$\mathbb{V}[P] = \mathbb{V}[D] - 2\xi \frac{\theta_r \phi_d^{L+2}}{1-\phi_d \phi_r} \mathbb{V}[D] + \xi^2 \mathbb{V}[R] + 2 \frac{\phi_d(1-\phi_d^{L+1})(1-\phi_d^{L+2})}{(1-\phi_d^2)(1-\phi_d)} \sigma_{\varepsilon_d}^2$$

- The variance of the net stock levels, $\mathbb{V}[NS]$, is independent of the re-manufacturing parameters, $\{\xi, \phi_r, \theta_r, \text{ and } \sigma_{\varepsilon_r}\}$.
- Thus, inventory related costs are not influenced by the decision variable ξ and the uncertainty in the returns, $\varepsilon_{r,t}$.

Objective Cost Function: System-wide Cost

$$C(\xi) = \mathbb{E}[h(NS_t)^+ + b(-NS_t)^+] + \min_{\xi} \left\{ \underbrace{\mathbb{E}[uk + w(P_t - k)^+] + \mathbb{E}[u'k' + w'(\xi R_t - k')^+] + \mathbb{E}[(1 - \xi)R_t \times g]}_{\text{Function of } \xi} \right\},$$

where

- ξ : yield rate (decision variable)
- h : unit holding cost, b : unit backlog cost.
- u : unit manufacturing cost when P_t is less than the capacity k .
- w : unit manufacturing cost when P_t is greater than the capacity k .
- u' : unit remanu. cost when ξR_t is less than the capacity k' .
- w' : unit remanu. cost when ξR_t is greater than the capacity k' .
- $u > u'$ and $w > w'$.
- g : unit disposal cost.

Derivatives of Objective Cost Function

First and Second derivatives

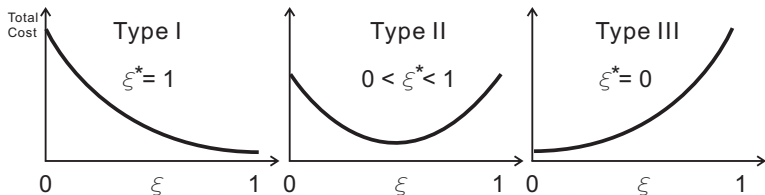
$$\begin{aligned}\frac{\partial C(\xi)}{\partial \xi} &= w\varphi \left[\Phi^{-1} \left[\frac{w - u}{w} \right] \right] \frac{\xi \mathbb{V}[R] - COV_{L+1}}{\sqrt{\mathbb{V}[P]}} \\ &\quad + w'\varphi \left[\Phi^{-1} \left[\frac{w' - u'}{w'} \right] \right] \sqrt{\mathbb{V}[R]} - (u - u')\mu_r - g\mu_r.\end{aligned}$$

$$\begin{aligned}\frac{\partial^2 C(\xi)}{\partial \xi^2} &= w\varphi \left[\Phi^{-1} \left[\frac{w - u}{w} \right] \right] \frac{\mathbb{V}[R]\mathbb{V}[P|\xi = 0] - (COV_{L+1})^2}{\mathbb{V}[P]\sqrt{\mathbb{V}[P]}} \\ &> 0, \text{ if } 1 > \phi_d \geq 0.\end{aligned}$$

Convexity of Objective Cost Function

Corollary

When the CL-OUT policy is exploited by the manufacturer, the system-wide cost is convex in the triage yield, ξ .



"Yield Rate Paradox" could exist

System-wide cost could increase in the triage yield rate ξ , even though the unit remanufacturing cost is lower than that of the manufacturing.

First-order derivative is decreasing in μ_r as long as $u > u'$

Corollary

When the CL-OUT policy is exploited by the manufacturer, the first-order derivative of the system-wide cost is decreasing in μ_r as long as $u > u'$.

$$\begin{aligned} \frac{\partial C(\xi)}{\partial \xi} = & w\varphi \left[\Phi^{-1} \left[\frac{w - u}{w} \right] \right] \frac{\xi \mathbb{V}[R] - \text{COV}_{L+1}}{\sqrt{\mathbb{V}[P]}} \\ & + w'\varphi \left[\Phi^{-1} \left[\frac{w' - u'}{w'} \right] \right] \sqrt{\mathbb{V}[R]} - (u - u')\mu_r - g\mu_r. \end{aligned}$$

Larger average returns could help profitability

Larger average returns might shift the cost function curve from Type III to Type II (or, from Type II to Type I).

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Example of Numerical Analysis: Data Set

• Demand

- $\mu_d = 20$
- $\phi_d = 0.4$
- $\sigma_{\varepsilon_d} = 3$

• Return

- $\mu_r = \{10, 15\}$
- $\phi_r = 0.7$
- $\theta_r = 0.5$
- $\sigma_{\varepsilon_r} = 1$

• Lead time

- $L = 1$

• Inventory related cost

- $h = 1$
- $b = 9$

• Cost of manufacturing new products

- $w = 11$
- $u = 4$

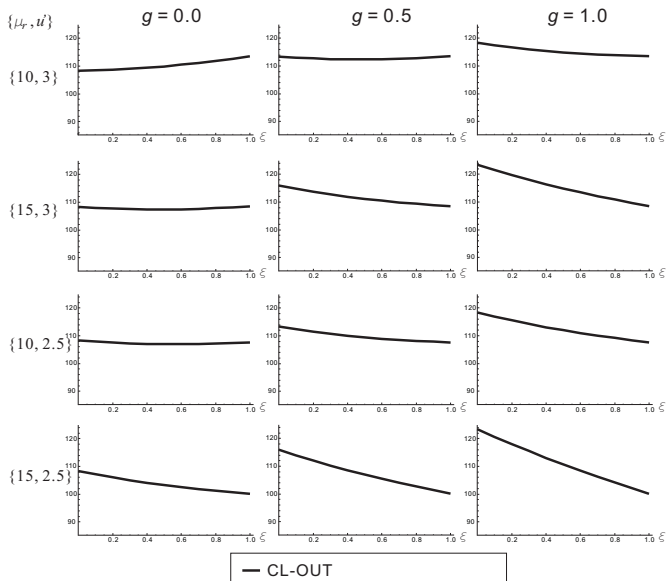
• Remanufacturing cost

- $w' = 9$
- $u' = \{3, 2.5\}$

• Disposal cost

- $g = \{0, 0.5, 1\}$

Impact of μ_r , u' , g with CL-OUT Policy



Closed-Loop APIOBPCS (CL-APIOBPCS)

Automatic Pipeline Inventory Order Based Production Control System for Closed-Loop Supply Chain

A modified OUT policy with feedback proportional controllers [John et al., 1994].

$$P_t = \mathbb{E}[D_{t+L+1}] - \xi R_t \\ + (1 - \alpha)(TNS - NS_t) \\ + (1 - \alpha) \left(\mathbb{E} \left[\sum_{i=1}^L D_{t+i} \right] - (WIP_t + WIPR_t) \right),$$

where $|\alpha| < 1$. Note that when $\alpha = 0$, CL-APIOBPCS degenerates into the CL-OUT policy.

[Lin et al., 2017] provides a comprehensive review of the APIOBPCS family of production planning archetypes.

CL-APIOBPCS

Variances of the net stock levels $\mathbb{V}_A[NS]$ and production of new items $\mathbb{V}_A[P]$

$$\mathbb{V}_A[NS] = \mathbb{V}[NS] + \frac{(\phi_d^{L+1} - 1)^2 \alpha^2 \sigma_{\varepsilon_d}^2}{(\phi_d - 1)^2 (1 - \alpha^2)} \geq \mathbb{V}[NS].$$

$$\mathbb{V}_A[P] = \mathbb{V}[P] + 2\alpha\sigma_{\varepsilon_d}^2 \left(\phi_d^{L+1} - 1 \right) \times \frac{\left(\xi\theta_r(\alpha^2 - 1)(\phi_d - 1) + (\phi_d(\alpha - \phi_d^L(\alpha - \phi_d + \phi_d(\phi_d - 1)(1 + \alpha)))) - 1 \right) (\alpha\phi_r - 1)}{(1 + \alpha)(\phi_d - 1)^2(\alpha\phi_d - 1)(\alpha\phi_r - 1)}.$$

- $\mathbb{V}_A[NS]$ is still independent of the re-manufacturing parameters, $\{\xi, L, \phi_r, \theta_r, \text{ and } \sigma_{\varepsilon_r}\}$ as in the case of CL-OUT policy.
- It is easy to recognize that $\mathbb{V}_A[NS] = \mathbb{V}[NS]$ and $\mathbb{V}_A[P] = \mathbb{V}[P]$ given that $\alpha = 0$.

Advantage of CL-APIOBPCS

- Objective cost function for **CL-OUT** policy:

$$C(\xi) = \mathbb{E}[h(NS_t)^+ + b(-NS_t)^+] + \min_{\xi} \left\{ \underbrace{\mathbb{E}[uk + w(P_t - k)^+] + \mathbb{E}[u'k' + w'(\xi R_t - k')^+] + \mathbb{E}[(1 - \xi)R_t \times g]}_{\text{Function of } \xi} \right\}$$

- Objective cost function for **CL-APIOBPCS** policy:

$$C(\alpha, \xi) = \min_{\alpha, \xi} \left\{ \begin{array}{l} \mathbb{E}[h(NS_t)^+ + b(-NS_t)^+] \\ + \mathbb{E}[uk + w(P_t - k)^+] \\ + \mathbb{E}[u'k' + w'(\xi R_t - k')^+] \\ + g\mathbb{E}[(1 - \xi)R_t] \end{array} \right\} \left. \begin{array}{l} \} \text{Function of } \alpha \\ \} \text{Function of } \xi \end{array} \right.$$

CL-APIOBPCS can reduce the cost further

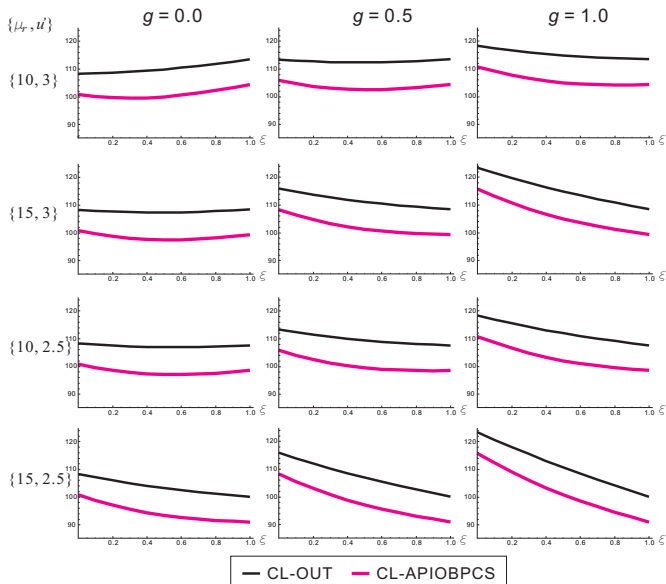


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Summary and Conclusion

Major Contributions

Existence of Yield Rate Paradox and its Mitigation Policies

- System-wide cost is convex in the triage yield rate ξ .
- Larger μ_r and/or smaller u' can mitigate *yield rate paradox*.
- Higher disposal cost g can induce one to remanufacture all returns.
- Introduce CL-APIOBPCS
 - Proportional controller α can reduce the cost further.

Social/Managerial Implication

- Larger mean of the returns is good for not only the sustainability but also a cost-conscious system.

Future Research Direction

- When/how does the yield rate paradox actually emerge in real life?
- Are there any other mitigation policies?

Thank you for your attention. Any questions?

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