

**SMMSO 2017**

June 4-9, 2017 - Acaya (Lecce), Italy

11th Conference on Stochastic Models of Manufacturing and Service Operations

# The Impact of Orbit Dependent Return Rate on the Control Policies of a Hybrid Production System

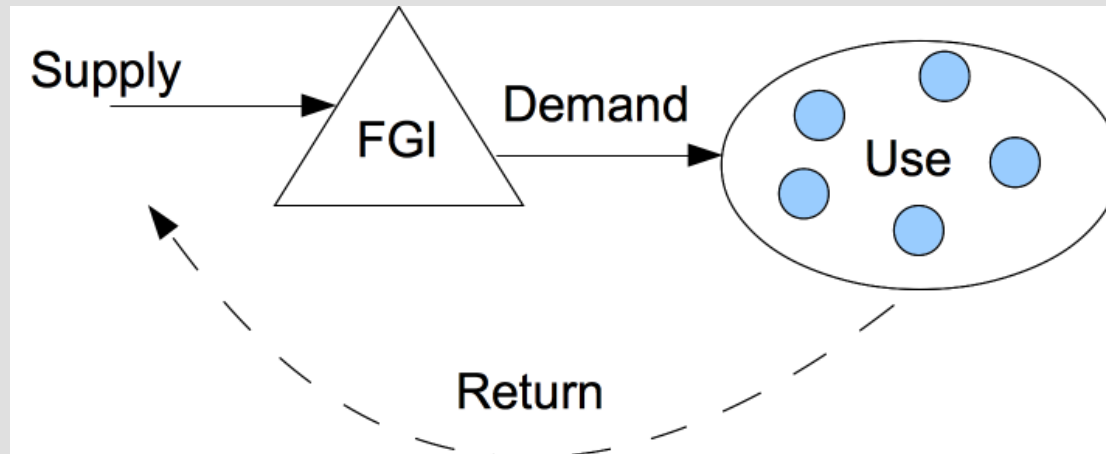
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# Production Control and Return Flows



Objective: Production control under uncertain return flows.

- Returns may affect inventory levels.
- Return flow estimates are required.
- Returns and supplied demand are dependent through products in use

# Production Control with Return Flows

- ⌘ Literature assumes demand and return flows to be independent in most cases.
- ⌘ Demand generates new users when the sales is not for replacement.
- ⌘ The more new products are sold to new users the higher the number of products in use (in orbit).
- ⌘ The size of the orbit determines the return rate.
- ⌘ When some of the products in orbit are disposed off rather than being returned an unobservable decrease occurs in the products in use.
- ⌘ The return rate only gives a partial information about this fact.

# Literature on Optimal Control Models

How would the information about orbit affect the production system control?

∞ Toktay et. *al* (2000)

- CLSC model using QN with  $M/\infty$  queue representing users

∞ Flapper et. *al* (2012)

- Optimal production control with advanced return info.

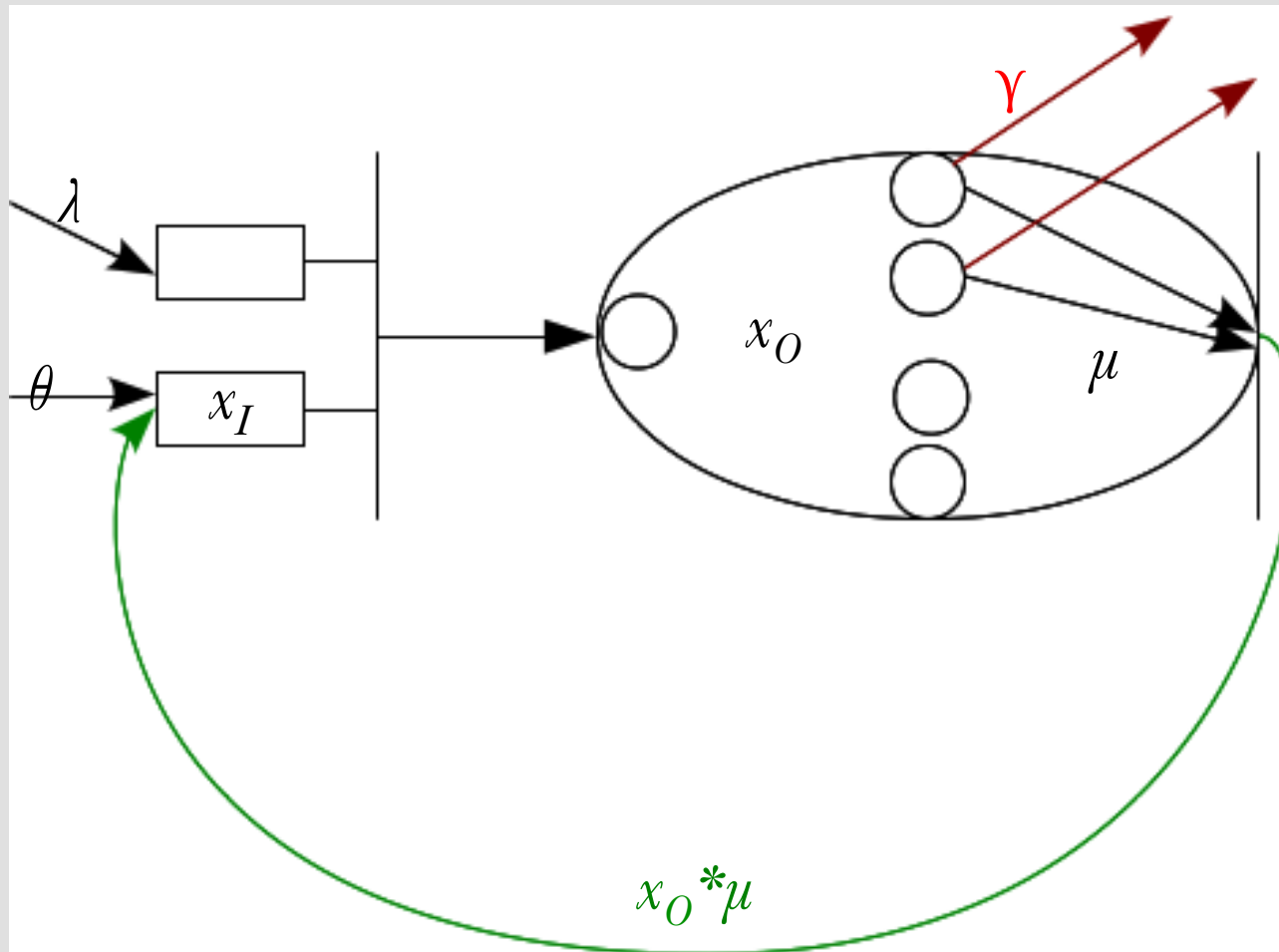
∞ Zerhouni et. *al* (2013)

- Comparison of dependent vs. independent returns under base stock control.
- Continuous review base stock optimal for independent returns
- Instantaneous usage rate assumption for dependent returns

# Objectives of the Study

- ∞ How does the production decision change under return flows dependent on the size of the orbit?
- ∞ What is the value added of using return flows in manufacturing?
- ∞ What is the cost of ignoring/inaccurately estimating the information of orbit size?

# The Model



$\lambda$  : Demand rate

$\mu$  : Usage rate

$\gamma$  : Disposal rate

$\theta$  : Production rate

$$p = \mu / (\gamma + \mu)$$

$x_0$  : Orbit size

$x_I$  : Inv. level

# Assumptions

- Each product in orbit can be disposed of with a positive probability.
- No need for remanufacturing (or instant remanufacturing to as good as new).
- All returns are accepted.
- The system is continuously observed.
- $x_0$  is fully observable.

# Value Function

$$v(x) + g = \Lambda_2^{-1} \left[ hx_I^+ + \mu x_O v(x_O - 1, x_I + 1) \right. \\ \left. + \gamma x_O v(x_O - 1, x_I) \right.$$

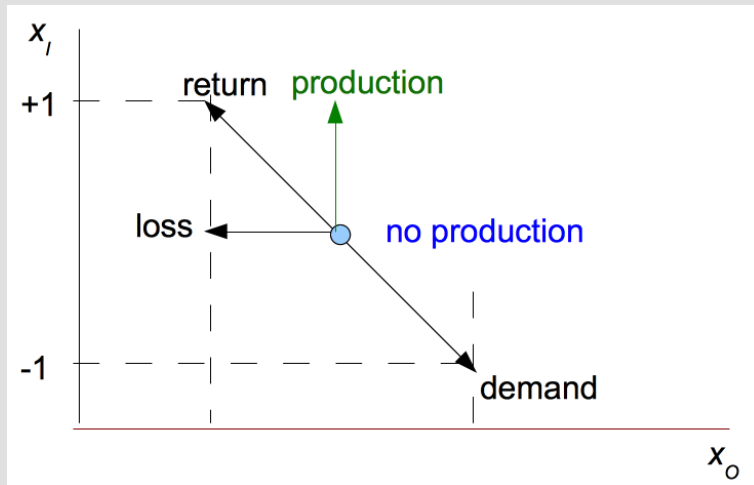
$$+ \lambda \begin{cases} v(x_O + 1, x_I - 1), & x_I > 0 \\ v(x_O, x_I) + c_L, & x_I = 0 \end{cases}$$

$$+ \min \theta \{ v(x_O, x_I + 1), v(x_O, x_I) \} \Big]$$

$$\Lambda_1 = (\mu + \gamma) x_O^{\max} + \lambda + \theta$$

$h$  : holding cost

$c_L$  : lost sales cost







# A Basis for Comparison



- Zerhouni et. al (2013) show for a state independent stationary return flow the optimal control policy is base stock.
- State independent return rate estimates are needed.
- An estimate can be given as  $\hat{\mu} = \alpha \cdot \bar{\mu}$ ,  $\alpha \in [0, \infty)$  where  $\bar{\mu} = E[x_0] \cdot \mu$ .
- Thus the average orbit size is required.

# LP Model

∞ The LP equivalent of the value function calculates the stationary distribution  $P(x_I, x_0)$  (Büyükdaglı and Fadiloğlu, 2016).

$$\min Z = \sum_{x \in S} (c_{x,0} \cdot y_{x,0} + c_{x,1} \cdot y_{x,1})$$

*s.t.*

$$\sum_{x \in S} (y_{x,0} + y_{x,1}) = 1$$

$$y_{x,0} + y_{x,1} - \Lambda^{-1} \left\{ \theta(y_{x,0} + y_{x-e_I,1}) + \lambda(y_{x-e_0+e_I,0} + y_{x-e_0+e_I,1}) \right. \\ \left. + x_0 \mu(y_{x+e_0-e_I,0} + y_{x+e_0-e_I,1}) + x_0 \gamma(y_{x-e_I,0} + y_{x-e_I,1}) \right. \\ \left. + (M - x_0)(\gamma + \mu)(y_{x,0} + y_{x,1}) \right\} = 0$$

$$\forall y_{x,0} \geq 0, y_{x,1} \geq 0$$

$$\forall x \in S$$

$$x = (x_0, x_I), e_0 = (1, 0), e_I = (0, 1)$$

$$c_{x,d} = h \cdot x_I + (1 - x_I)^+ \cdot \lambda \cdot c_L, d = \{0, 1\}$$

# The Procedure

$$\hat{\mu} = \alpha \cdot E[x_0] \cdot \mu \text{ from LP}$$

$$S_{\hat{\mu}} \text{ from } v(x_I)$$

$$S_{\hat{\mu}} \text{ to } v(x) = v_{S_{\hat{\mu}}}(x)$$

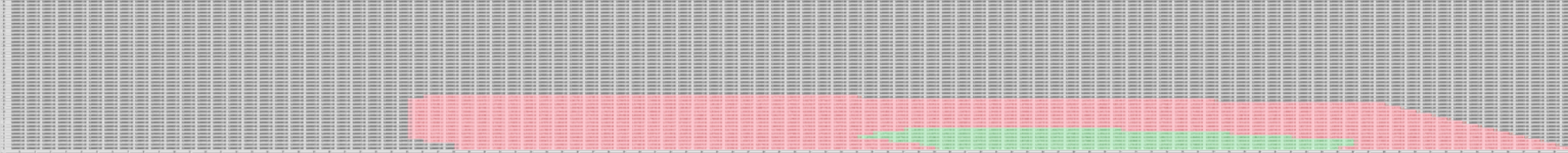
Compare  $v(x)$  &  $v_{S_{\hat{\mu}}}(x)$

$v(x_I)$  is the value function for optimizing the FGI without using the orbit information.

$S_{\hat{\mu}}$  is the optimal base-stock level for  $v(x_I)$ .




# Stationary Distribution from LP



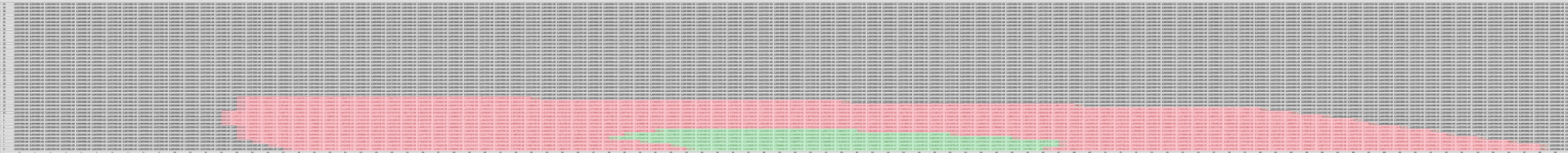
A plot showing the stationary distribution for  $\mu = \gamma = 0.006$ . The x-axis represents a state space, and the y-axis represents probability density. A red shaded area indicates the distribution, which is unimodal and slightly skewed to the right. A green shaded area is nested within the red one, also unimodal and slightly skewed to the right.

$$\mu = \gamma = 0.006$$



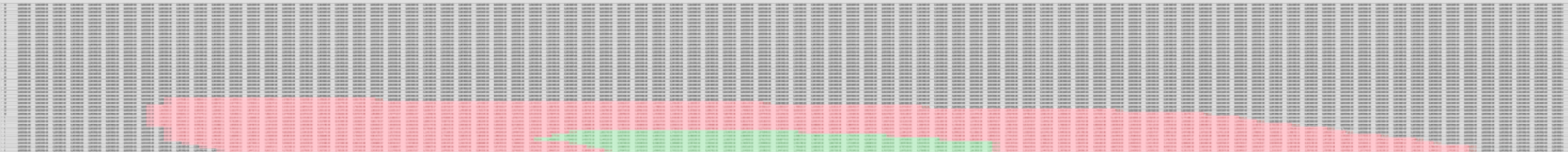
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$$\mu = \gamma = 0.007$$



A plot showing the stationary distribution for  $\mu = \gamma = 0.008$ . The x-axis represents a state space, and the y-axis represents probability density. A red shaded area indicates the distribution, which is unimodal and slightly skewed to the right. A green shaded area is nested within the red one, also unimodal and slightly skewed to the right.

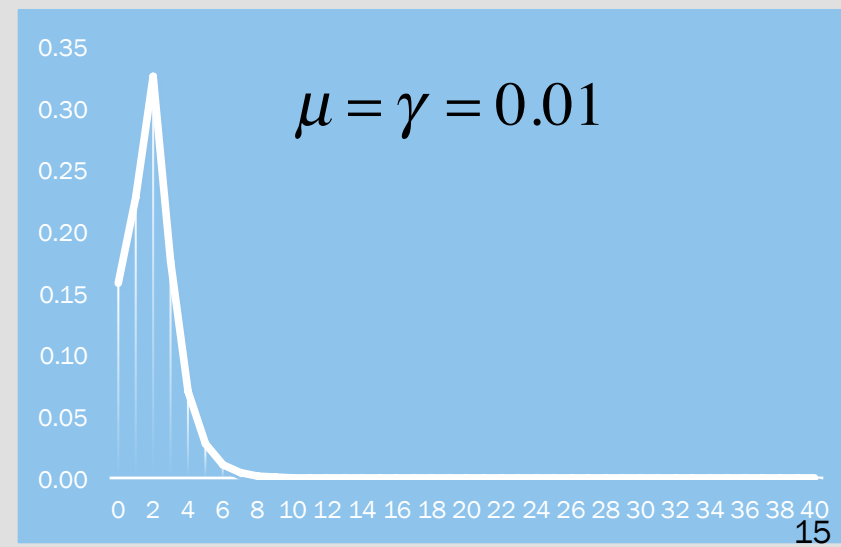
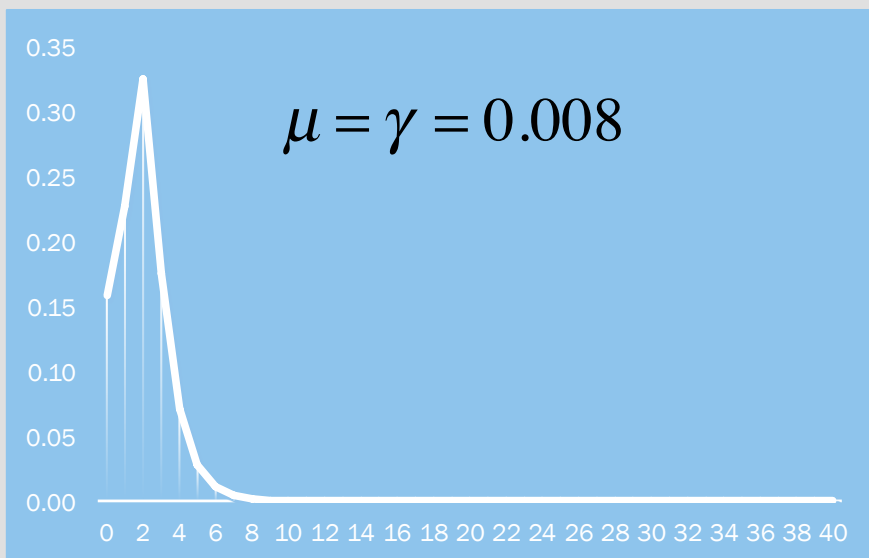
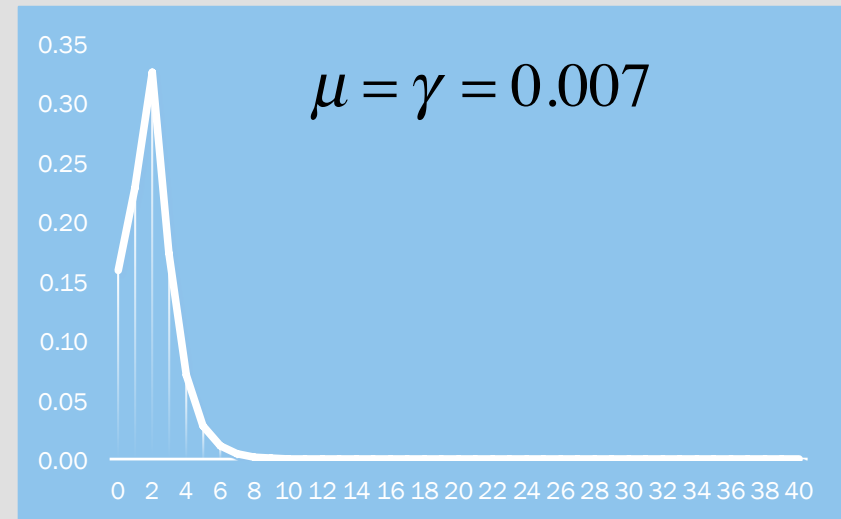
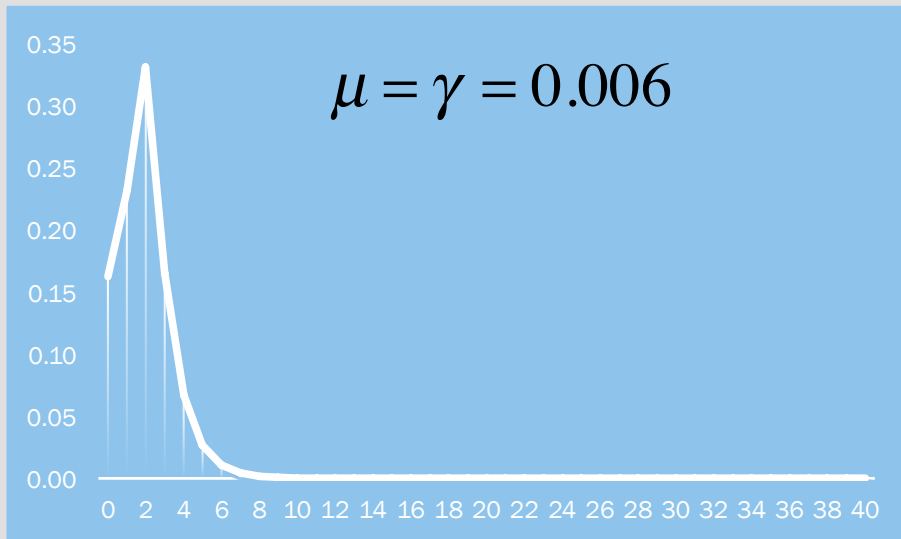
$$\mu = \gamma = 0.008$$



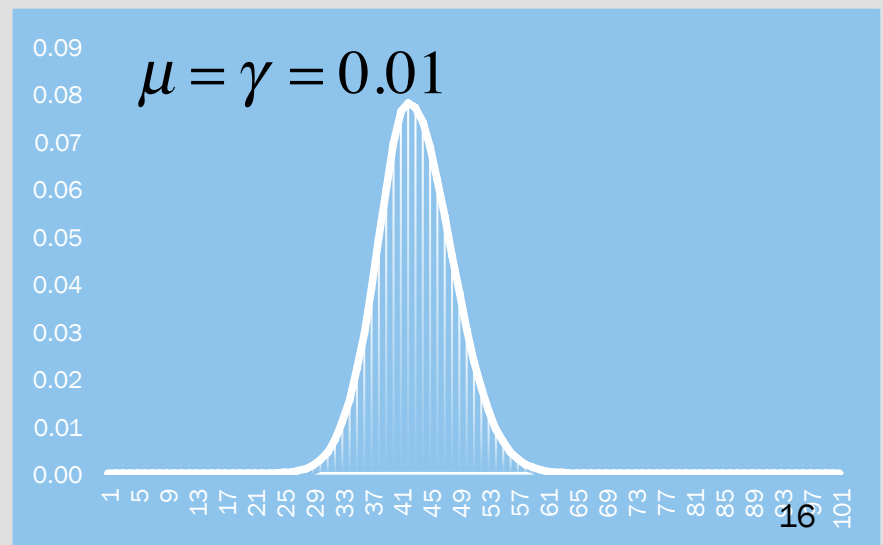
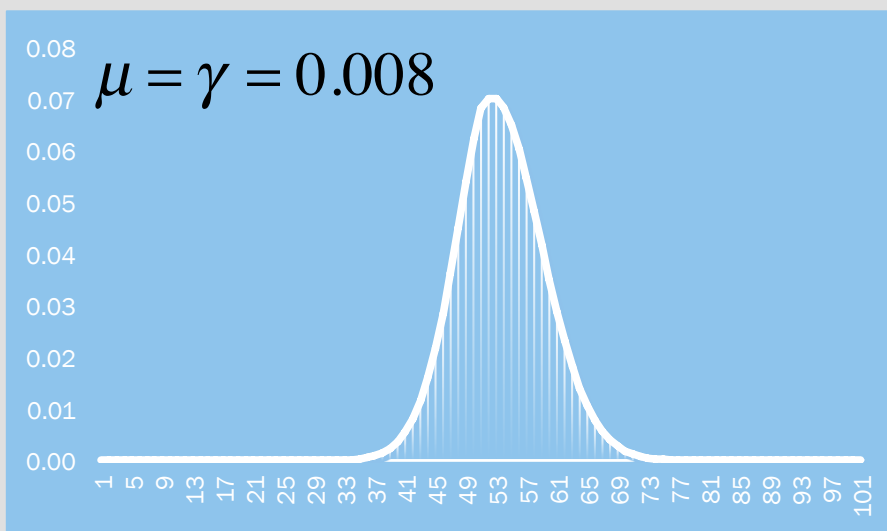
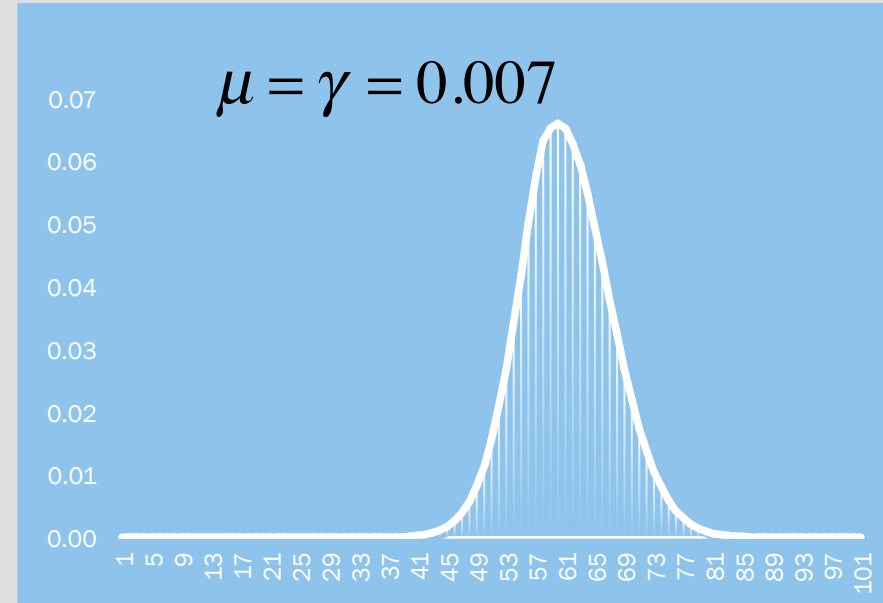
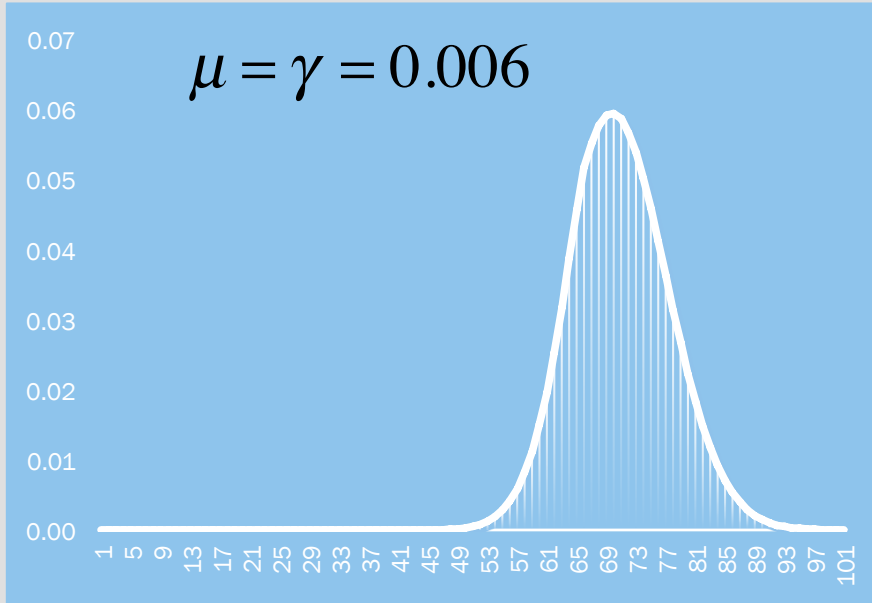
A plot showing the stationary distribution for  $\mu = \gamma = 0.01$ . The x-axis represents a state space, and the y-axis represents probability density. A red shaded area indicates the distribution, which is unimodal and slightly skewed to the right. A green shaded area is nested within the red one, also unimodal and slightly skewed to the right.

$$\mu = \gamma = 0.01$$

# Marginal Probability of Inventory Level



# Marginal Probability of Orbit Size





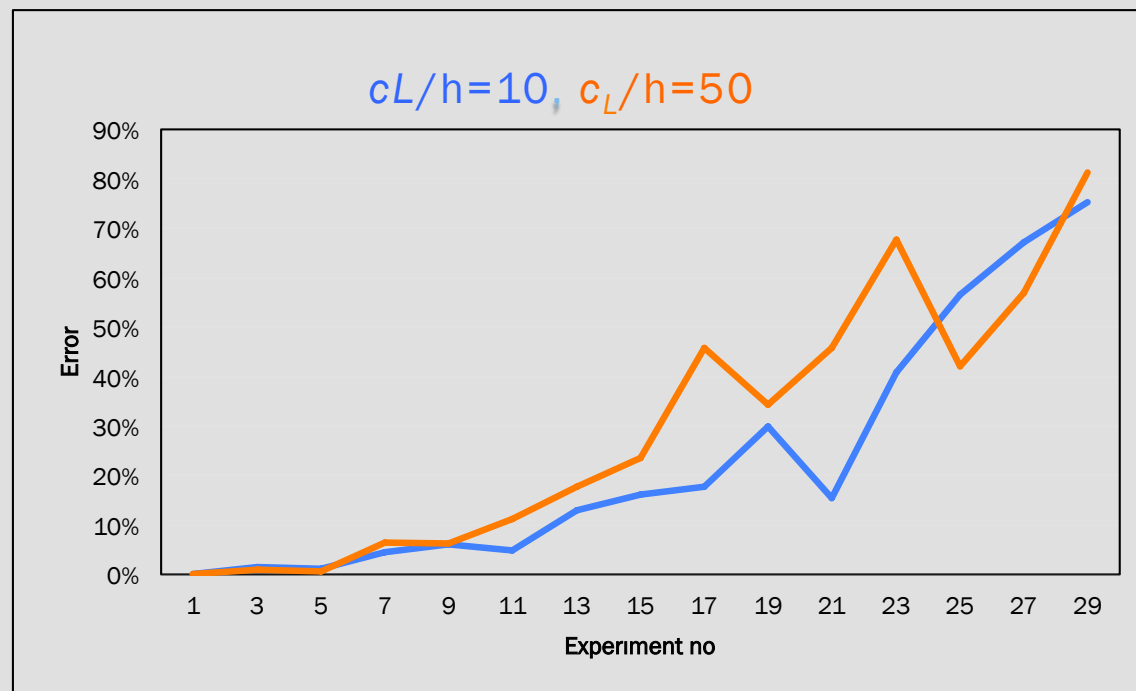
# Pure vs. Hybrid Production

For  $\alpha = 0$  we can observe the added value of using returns.

If no returns are accepted  $S_{max}$  is the optimal control policy.

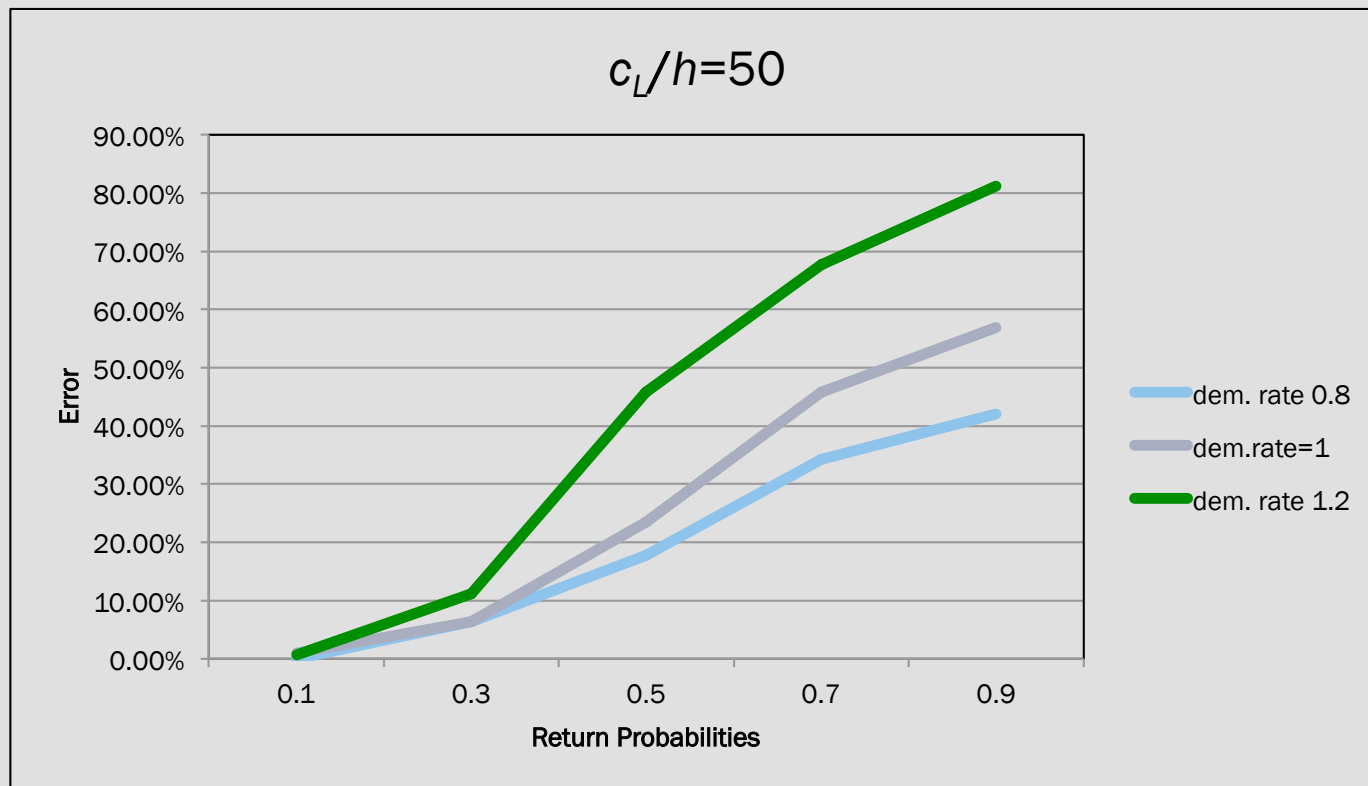
$$\theta = 1, \lambda = \{0.8, 1, 1.2\}, p = \mu / (\mu + \gamma) = \{0.1, 0.3, 0.5, 0.7, 0.9\},$$

$$\mu = \{0.002, \dots, 0.018\}, c_L/h = \{10, 50\}, h = 1$$



# Pure vs. Hybrid Production

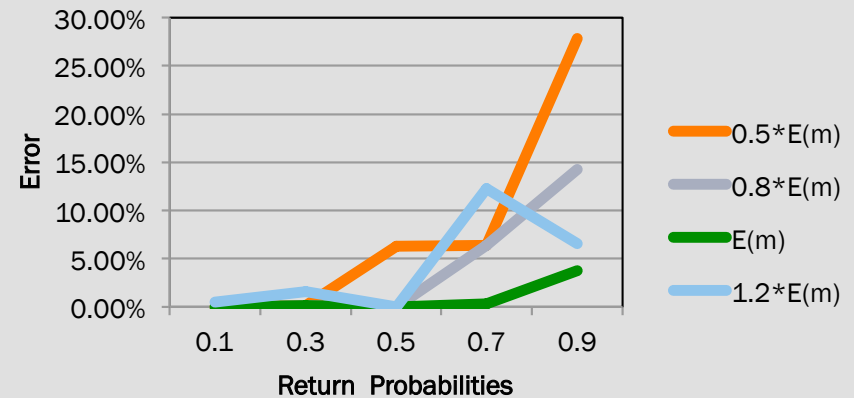
- The impact of  $p$  and  $\lambda/\theta$
- As  $\lambda/\theta$  increases the impact of returns increase since the orbit expands. For  $p=0.9$  and  $\lambda = 1.2, \bar{\mu} = 1.047$



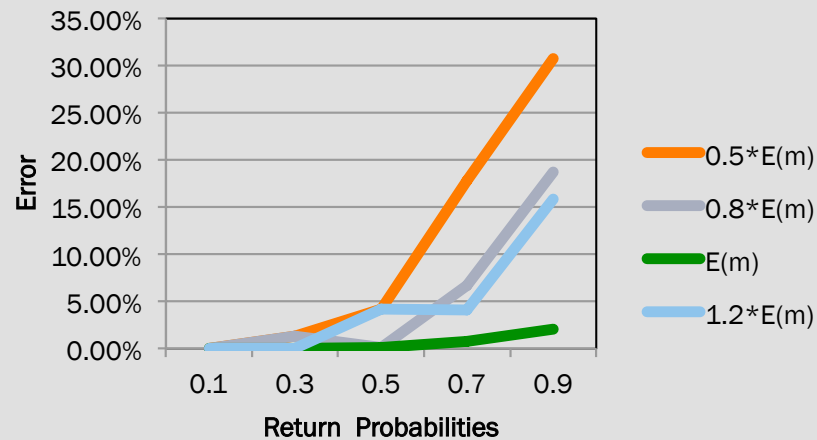
# Value of Orbit State

$$\hat{\mu} = \bar{\mu} \cdot \{0.5, 0.8, 1, 1.2,\}$$

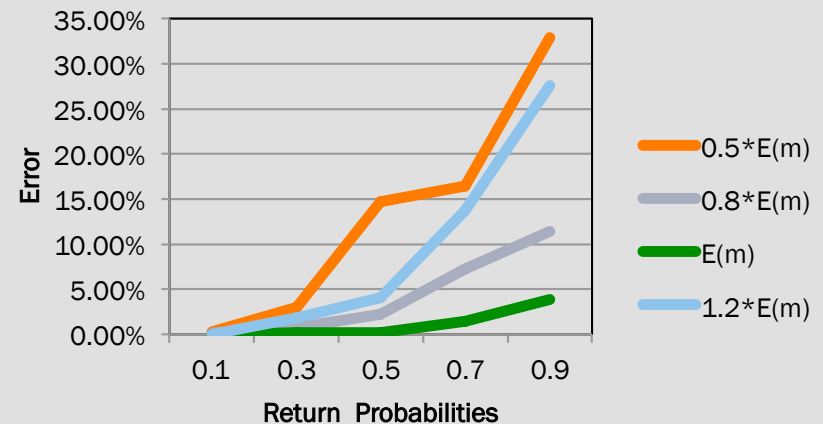
$c_L/h=50$ , demand rate=0.8



$c_L/h=50$ , demand rate=1



$c_L/h=50$ , demand rate=1.2



# Remanufacturing vs Manufacturing Cost

Control levels for  $\bar{\mu}$  model under the impact of production costs.

$$\theta = 1, c_L/h = 50, c_M = 5, c_R = 2$$

	$\lambda = 0.8$		$\lambda = 0.8$		$\lambda = 0.8$	
p	S with $C_R - C_M$	S	S with $C_R - C_M$	S	S with $C_R - C_M$	S
0.1	4	5	6	7	10	10
0.3	3	4	5	5	7	7
0.5	3	3	4	4	5	5
0.7	2	2	3	3	3	4
0.9	1	1	1	1	2	2

# Conclusions

- A testbed for measuring the impact of the number of products in use (orbit) on the production control of hybrid production systems is introduced.
- The added value of using returns for supplying the demand is clearly displayed.
- Results indicate that the base stock policy is heavily dependent on the orbit size and return probabilities.
- The difference between complete information and no information present the value of the partial observation effort.

# Future Work

Model structure extensions:

- Return admission control
- Remanufacturing lead time
- Demand differentiation
- The feedback relation between the orbit and the demand rate.

Testbed extensions:

- Automated LP model
- LP model for upto  $S$  levels
- Measure the effectiveness of proposed methods so far



# Thank You

## References:

Toktay et. al (2000), Inventory Management Of Remanufacturable Products, Mgt. Sci., 46.

Flapper et. al (2012), Control Of A Production-inventory System With Returns Under Imperfect Advance Return Information, EJOR, 218.

Zerhouni et. al (2013), Influence Of Dependency Between Demands And Returns In A Reverse Logistics System, IJPE 143.

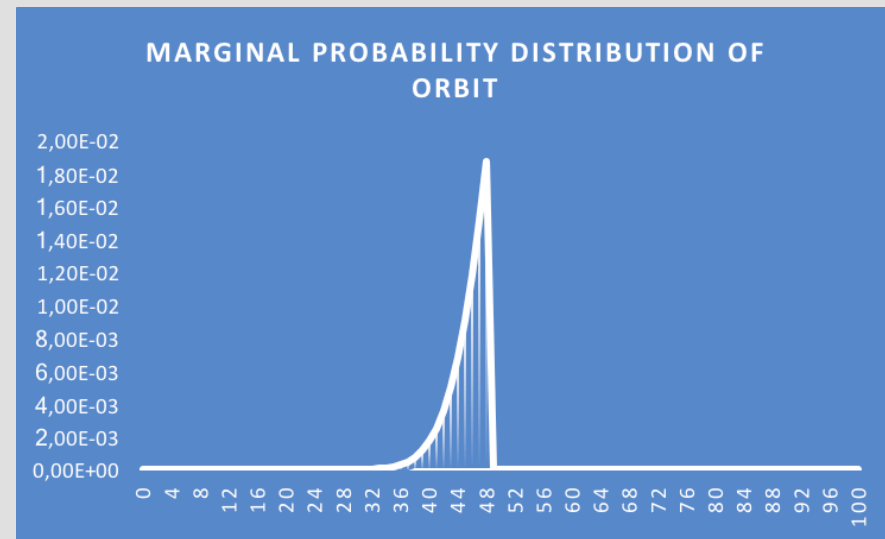
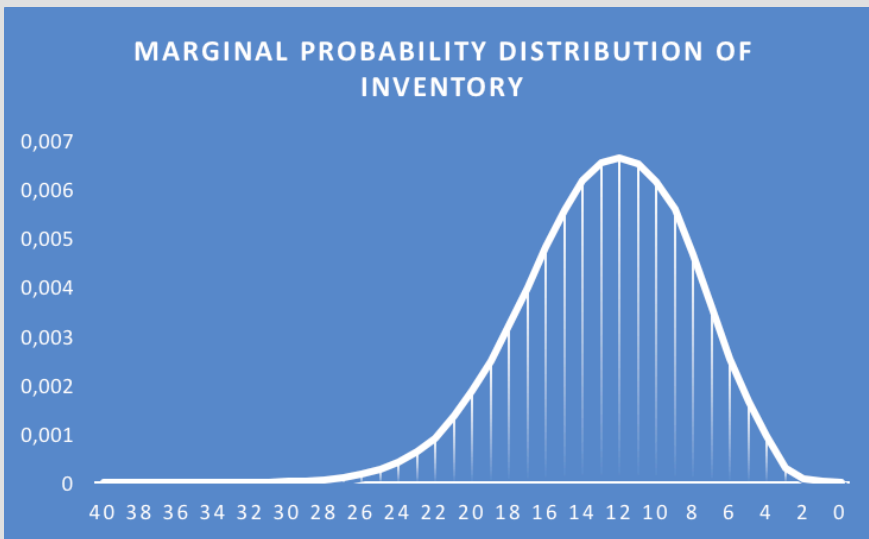
Büyükdağlı and Fadiloğlu (2016) A Mathematical Programming Framework for Control of Tandem Production Lines, Research Report.

# Distributions of the given example.

$$\theta = 1, \lambda = 1.2, \gamma = 0.002, \mu = 0.018, (p = 0.9), c_L/h = 50$$

$$E[x_I] = 6,195$$

$$\bar{\mu} = 1,047, E[x_0] = 58$$



$$P(x_I, x_0)$$