

Process Management (OPIM 631)

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I. Basic Performance Measures

- **process** – transforms **inputs** (RM, unserved customers) into **outputs** (FG, customers served) with **resources** (machines / workers / capital)
- **flow unit** – unit of analysis
 - types
 - physical –
 - monetary – facilitates aggregating across different inventory types
 - challenges
 - flow unit breaks up in process
 - multiple types of flow units present (~ **product mix**)
- **capacity** – # workers / activity time
 - limit = minimum (bottleneck capacity, demand)
 - maximum rate at which process can generate supply
 - max rate at which process produces during given time
 - maximum amount a resource can produce in a given time
 - depends on – product mix
 - consider – number of resources
- **product mix** -
- **Basic performance measures**
 - **WIP / Inventory** – flow units contained within a process
 - represents a liability for operations
 - to evaluate – look at process over time (rather than at specific time); watch unit through whole process
 - measuring – can measure by tracking inputs and outputs alone
 - **why hold?** –
 - ◆ i) pipeline ~ due to in process time
 - ◆ ii) seasonal ~ build up inventory in advance
 - ◆ iii) cycle (EOS in processing batches) – more common if set-up required
 - ◆ iv) decoupling / buffer – allows mgmt to run steps independently; reduces effect of variations in flow
 - ◆ v) safety – addresses stochastic demand
 - > **stochastic demand** ~ variation in demand about predicted demand
 - problem in retailing / finished goods
 - **why important?** –
 - ◆ costly – obsolescence + perishable + shrinkage + overhead storage costs
 - > individual product's inventory cost = total inventory cost per yr / inventory turns

- ♦ ops sees as a liability
- ♦ hinders process improvement (multiplies the effects of any defects)

- **Flow time** = time in queue + activity time = $T_q + p$
 - time takes flow unit to get through project
 - consider if flow units pass each other during process
- **Flow rate** (through-put rate) = flow units / time
 - rate at which process delivers output
 - **limit** = minimum (input rate, demand, bottleneck capacity)
 - most important
 - if inventory is in\$ → look at COGS (not sales)
- **inventory turns** – frequency that inventory turns over during yr
 - = 365 days / flow time (in days)

- **Little's Law** – $Inv = Avg\ Flow\ Rate * Avg\ Flow\ Time$

- always holds BUT deals with avg
- caveat – not impacted by order

- **Consider**

- higher flow rate ~ higher revenues
- shorter flow time ~ more sales / higher P
- lower inv ~ lower WC + lower flow time

II. Supply Process

- **bottleneck** – resource with the smallest capacity
 - limits the capacity of the overall process
 - **true bottleneck** - has highest **implied process utilization** (flow rate / capacity)
- **demand constrained** – when demand < supply
- **supply constrained** – when demand > supply (limit is either bottleneck or inputs)
- **steps for analyzing a process**
 - **i)** define process boundaries – part of business will focus on
 - **ii)** identify appropriate level of detail
 - **iii)** define flow unit
 - consider if focus amounts = input or output
 - convert amounts for consistency by following unit volume / weight throughout process (front to back or vice versa)
 - input maybe better if process itself consumes resources (i.e., more goes in than comes out)
 - **iv)** draw process flow chart – consider from flow unit perspective
 - symbols
 - ♦ box – process step
 - ♦ triangle – buffer holding inventory
 - ♦ arrows – route of flow
 - ROT – include only processes / steps that affect process flow or economics
 - **v)** determine capacity of each step (flow units / period of time)
 - **vi)** find Bottleneck – using Little's Law → calculate capacity of each step (~ flow rate per unit of time)
 - consider how long takes to process x units = $x / Flow\ Rate$
 - **true-bottleneck** –
 - ♦ a) find **requested capacity** – trace capacity from beginning to current step
 - > **requested capacity** ~ total capacity needed t/b processed per time period
 - ♦ b) determine **implied utilization** ~ (**capacity requested** / **available capacity**)

- ◆ c) true bottleneck – has highest implied utilization

– **if product mix exists** – again compare requested capacity (across products) to available capacity (resources * x / flow rate)

- **vi) calculate utilization**
 - **process utilization** = Flow Rate / Process Capacity
 - ◆ ~ how much actually produced relative to how much could be produced
 - ◆ underutilization occurs if – Demand < Supply or insufficient Supply or process step has periods of limited availability or non-bottleneck
 - ◆ Bottleneck has highest process utilization
- **vii) Calculate performance measures**
- **Remember**
 - always fully utilize the bottleneck step
 - corp objective = max π <> max utilization

III. Balancing Process Steps

- **cycle time** = 1 / Flow Rate
 - time between the process spitting out finished product
 - ~ every x units of time, a flow unit enters / exits process (w/ x determined by the bottleneck)
- **idle time (step i)** = cycle time – the activity time at a given step (i)
- **span of control** = summed activity times of individual activities at a process step
- **specialization** – reduces span of control
 - bene = lower activity times
 - con = lower utilization (but we care about π / not utilization)
- **work cell** –
 - will have 100% utilization b/c must also be the bottleneck
 - but – reduces specialization (worker must master longer span of control)
 - may lead to longer activity times (due to less specialization)
- **goal** = max π <> max utilization
- **line balancing** = avoid mismatch b/w supply and demand of process steps
 - effect
 - = increase efficiency of process (better utilize resources) + increase capacity of process (reallocating b/w over / under used resources)
 - = reduce direct labor cost
 - consider – can re-sequence tasks
 - harder – with increase in specialization (easier with decrease in specialization)
- **sequential process steps** -
 - **worker paced line** – inventory exists b/w process steps
 - **machine paced line** – no inventory between process steps
 - time to process x units =
 - **a) starting cold** = time to finish first unit + [(x-1) / Flow Rate]
 - ◆ time to finish first unit – depends
 - > worker paced line – sum of activity times for each step
 - > machine paced line – number of steps * activity time of Bottleneck step

- **b) continuous flow** – time to finish first unit + $[x / \text{Flow Rate}]$ **why different (p. 176)**
- **labor** -
 - **direct labor content** = sum of activity times of labor process steps
 - ignores how process operated (i.e., labor cost w/b different)
 - **direct labor cost** – s/b higher than labor content
 - = (total wages/ unit of time) / (# product / unit of time)
 - = (labor cost) / process flow rate
 - **difference** = from under utilization
 - **remember** –
 - adding more people <> change labor content (may change labor cost if added to non-bottleneck)
 - assumption = labor cost is fixed (but can day be shortened?)
 - **average labor utilization** -
 - **a)** labor content / (labor content + idle time)
 - **b)** average utilization across all workers = $1/m * (u_1 + \dots + u_m)$
- **Scaling up to higher volume** (p. 183-)
 - **balance line** – increases utilization (reduces direct labor content)
 - **add labor** –
 - **i)** replicate balanced line (to new line) – but current line may still not be best way after more workers added
 - ◆ avg utilization remains unchanged
 - **ii)** add people to individual process steps
 - ◆ capacity = # workers / activity time (with more workers → capacity rises)
 - **iii)** divide work further (increase specialization)
 - ◆ **a)** determine targeted throughput / cycle time
 - ◆ **b)** how many workers does it take to met target (a)?
 - > **1)** break each process steps into individual activities (w/ activity times)
 - > **2)** disperse activities evenly across all workers
 - > **3)** **span of control** = summed activity times of individual activities at a process step
 - ◆ **c)** consider – greater specialization (less training / quicker learning) BUT labor utilization may go down due to bottleneck

IV. Batching / Flow Interruptions

- **batch processing** – as opposed to mass production systems
 - **generally** - uses general purpose technology to produce larger variety of products
 - **set up** –
 - leads to lower capacity (b/c no production occurs during set-up)
 - ~ interruption in process flow (stealing capacity)
 - **only one goal** = eliminate / reduce time required for each one
 - **capacity w/ set up** ~ $\text{Batch size} / [\text{setup time} + (\text{batch size} * \text{activity time per unit})]$ ~ a function of batch size
 - ◆ as increase batch size (spread set up over more units) and approach max = $1 / \text{activity time}$
 - **Single Minute Exchange of Die (SMED) Method** –
 - ◆ divide all set up tasks
 - > internal setup tasks ~ only executed while machine is **stopped**
 - > external set-up tasks ~ c/b done before machine stops
 - ◆ bene = if applied to bottleneck (not helpful if improving flow of non-bottleneck)
 - **production batch** – collection of flow units that are processed before a resource needs to go through another set-up
 - **transfer batch** – collection of flow units transferred as group from one process step to another
 - may be beneficial to reduce below size of production batch (e.g., to reduce inventories & flow time)

- batch size
 - increasing size → increases capacity (output / time period)
- problems –
 - larger batch size = more inventory
 - more waiting (from flow unit perspective) throughout process
 - customer buying patterns may not coincide with batches
 - ◆ solution = **Mixed Model production** = batch size of 1 (eliminates cycle inventory + perfect align S & D)
- ROT –
 - if set-up occurs at bottleneck → increase batch size (to increase process capacity)
 - if set up at non-bottleneck → decrease batch size (to decrease inventory)
- Choosing a batch size – key = focus on set-up times (not set-up costs)~to capture effect of set-up on process capacity
 - i) calculate capacity of each step
 - ii) determine **optimal batch size** – equate Batch step capacity with next lowest capacity step and solve for B
 - ◆ **optimal batch size** – smallest batch size that does not adversely affect process capacity
- Economic Order Quantity (EOQ) Model –
 - always check to see if Bottleneck shifted

- **Other Flow interruption issues**

- **production run** – quantity between two flow interruptions
 - **capacity** ~ (amt processed b/w stops) / [duration of stop + (time to produce one flow unit * amt processed b/w stops)]
- machine down-time – step may be limited by down time required in other steps if no inventory buffers exist
 - capacity ~ $B / S + B * p$ (where S is machine downtime)
 - i.e., constraint may be different while running vs. during shutdown
- Buffers – increase process capacity in face of flow interruptions

- **remember**

- capacity at bottleneck = extremely valuable
- capacity at non-bottleneck = free
- investment in machinery = sunk (focus on time / not cost in choosing batch size)
- focus on time of set-up (not cost) b/c this captures the effect on the overall process capacity

V. Variability – Simple Process

- **Key point** – variability causes flow units to incur wait time even if utilization < 100%
- **inter-arrival time** ~ time between two consecutive arrivals in a process
 - follows **exponential distribution** → Prob ($A > t$) = $\exp(-t/a)$
 - A – random inter arrival time
 - a – avg inter-arrival time
- **exponential distribution** –
 - includes - customers arriving independently
 - ROT –
 - σ = avg inter-arrival time (a) → $CV_a = 1$
 - memoryless property – number of arrivals in next time slot is independent of when last arrival occurred
 - limits to ROT = if
 - arrivals are scheduled
 - arrivals are batched
 - arrivals are negatively correlated (go to different line b/c see someone just entered)
 - arrivals negatively correlated due to limited population of customers
- **Variability**
 - effect –
 - **only relevant in presence of dependence**
 - even with under-utilization (or match of S & D) → flow units wait
 - averages out over time **IFF observations are independent of each other** (i.e., independence b/w S & D)
 - inventory – introduces independence
 - **sequential dependence** – not possible to start step until prior step finishes
 - **critical chain effect** – mgrs adjust task durations to include contingency buffer (due to high cost of moving scheduled resources when flow unit not finished in earlier step)
 - source –
 - inflow of units –
 - ◆ biggest source in service organizations
 - activity times
 - ◆ human operators – better at some things than others
 - ◆ customer involvement – level of knowledge / attention varies
 - resources – random breakdowns
 - routing – randomness in path of flow unit (e.g., hospital)
 - measurement –
 - σ - no good b/c is an absolute number (same unit as mean)
 - **coefficient of variation** ~ σ / mean
 - ◆ measures variability relative
 - ◆ $CV_p = CV$ activity time
 - ◆ $CV_a = CV$ of **inter-arrival time**
 - problem ~ **seasonality** ~ predictable pattern
 - ◆ has nothing to do with variability
 - Solution – break process into step / periods during which seasonality does not occur
- **waiting time**
 - simple process – one buffer (unlimited space) + single activity
 - **i**) test reasonableness of model (HO 7)
 - ◆ **a**) presence of dependence
 - ◆ **b**) stationary process – any seasonality w/i process step? test by
 - > **1**) sort all arrival so increasing in time (label - $a_1 \dots a_n$)

- > 2) plot graph ($y =$ number customers arriving; $x =$ time)
 - > 3) diagonal line from first to last arrival (if stationary \rightarrow little deviation b/w graph & line)
 - > **solution to seasonality** = divide up process into smaller intervals w/ mean for each
 - ◆ c) exponential distribution of inter-arrival times?
 - ii) compute CV_a
 - ◆ a – avg inter-arrival time (\sim mean)
 - ii) calculate capacity \sim # **resource** / p
 - ◆ p – time to process flow unit
 - iii) calculate utilization \sim **flow rate** / **capacity**
 - iv) calculate wait time – w/ either
 - ◆ a) time flow unit waits in buffer (flow unit perspective) $\sim T_q$ (**measures LR avg wait time**)
 - > $T_q = \text{Activity time} * (u / (1-u)) * [(CV_a^2 + CV_p^2) / 2] -$
 - is per resource
 - is exact so long as inter-arrival times follow exponential distribution
 - > utilization (u) $m/b < 100\%$ \rightarrow if utilization $> 100\%$, then real issue is capacity (not variability)
 - > waiting time grows linearly with waiting time in system
 - ◆ b) number units waiting in buffer (system perspective) = ???
 - > Total inventory (I_T) = $I_q + I_p$
 - > I_p – b/w 0 & 1 (flow unit is either in or not)
 - > probability that process is busy working on unit = utilization
 - v) calculate total process time = $T_q + p$
 - ◆ w/b LR avg – first customers might be less than others
 - vi) calculate inventory – $I = R * T = 1/a(T_q + p)$
 - ◆ buffered inventory – $I_q = 1/a * T_q$
 - ◆ processing inventory – $I_p = (1-u) * 0 + u * 1 = u$
 - > $I_p = \text{prob (0 processed)} * 0 + \text{prob (1 processed)} * 1$
 - vii) account for variance in waiting times –
 - ◆ service level = prob (waiting time $\leq x$)
 - ◆ **target wait time** = $-\ln(1-\text{service level}) * T_q$
- remember
 - cannot just match avg D to avg S – variability causes wait times to arise
 - **seasonality** \sim predictable pattern
 - has nothing to do with variability
 - if utilization $> 100\%$ \rightarrow we do not care about variability b/c inventory builds up (real issue is capacity)
 - increasing utilization \rightarrow increases wait time
 - variability impacts quality of service (in addition to wait times)

VI Variability – Multiple Processes

- **focus** – only utilizations less than 100% (otherwise is a capacity issue)
- **utilization** = flow rate / capacity
 - = 1 / inter-arrival time / (number of resources * activity time)
 - = activity time / (inter arrival time * number of resources)
 - = p / am
- **avg waiting time** = $(\text{activity time} / m) * (\text{utilization}^{(2(m+1)^{.5}) - 1} / (1-u)) * (C_a^2 + C_p^2) / 2$ **is an approximation**
 - once compute \rightarrow can compute (with Little's Law) flow units in buffer (I_q) \rightarrow compute I_p and I_T
- **Inventory in process** = $m * u$ (\sim number of resources * utilization) – is cumulative across resources

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VII Matching Supply & Demand (OPIM 632) – Newsvendor Model (One Production Opportunity = Make to Stock)

- **key steps for profit maximizing order Q** –

- i) gather financial inputs (data)
 - ii) generate demand model / forecast (choose distribution function)
 - source – history, opinions, etc
 - iii) choose objective (\sim maximize Π)
 - iv) choose order Q to achieve objective
- **Process for Π maximizing order Q**
 - if NOT normal distribution
 - 1) find & rank A/F ratios in ascending order
 - 2) calculate empirical distribution [$F(A/F) = \% = \text{rank} / \text{total \# AF ratios}$]
 - 3) match critical ratio to A/F distribution %
 - 4) $Q = \text{Forecast} * \text{matched A/F ratio}$
 - if normal distribution
 - 1) Expected A/F ratio = avg of all A/F ratios
 - 2) σ of A/F ratio = σ of all ratios
 - 3) Expected Actual Demand = $E(A/F) * \text{Forecast}$
 - 4) Expected $\sigma = \sigma$ of A/F ratio * Forecast
 - 5) find Z based on critical ratio
 - 6) $Q = E(\text{Actual Demand}) + [E(\sigma) * Z]$
- **Process**
 - **1) Gather Inputs - Forecast Demand** (not sales)
 - **a)** need distribution for every possible demand
 - ◆ why? reveals accuracy / variance
 - ◆ solution – choose a defined distribution to describe your forecast
 - ◆ $F(Q) \sim$ distribution function \sim probability outcome will be value or smaller
 - ◆ $f(Q) \sim$ density function \sim probability outcome will be exactly a value
 - **b)** make demand forecast \sim opinions, past data, etc.
 - ◆ **i)** assume normal distribution for simplicity
 - ◆ **ii) A/F** = actual demand / forecast - focus on ABS errors (not relative errors)
 - > **Expected Actual Demand** = Expected A/F ratio * forecast
 - > **σ of Actual Demand** = σ of A/F Ratio * Forecast
 - ◆ **iii)** rank, sort A/F, calc rank / total # different products
 - ◆ **iv)** find $\mu = E(\text{Demand}_{\text{New}}) = \text{avg}(A/F) * \text{Forecast}_{\text{New}}$
 - ◆ **v)** find $\sigma = E(\sigma_{\text{New}}) = E(\sigma_{A/F}) * \text{Forecast}$
 - **c)** goodness of fit test - plot observed distribution function of A/F ratio
 - **2) Find Order Q** -
 - **a) Profit maximizing Order Q**
 - ◆ A/F ratio - Observe Empirical Distribution
 - > **i)** find critical ratio
 - > **ii)** find distribution – observe historical relative forecast errors \rightarrow A/F ratios for numerous products
 - > **iii)** rank products (low to hi) based on AF ratio
 - > **iv)** evaluate empirical distribution function $F(A/F)$ for each item
 - $F(A/F) = \text{each item's rank} / \text{number of items}$
 - > **v)** find $F(A/F) \sim$ critical ratio and identify the appropriate A/F ratio (choose the larger)
 - > **vi)** $Q = \text{selected A/F ratio} * \text{Forecast}$
 - ◆ A/F ratio - Fit Normal Distribution
 - > **i)** assume normal distribution for simplicity
 - > **ii) A/F** = actual demand / forecast - focus on ABS errors (not relative errors)
 - > **iii)** rank, sort A/F, calc rank / total # different products
 - > **iv)** find $\mu = E(\text{Demand}_{\text{New}}) = \text{avg}(A/F) * \text{Forecast}_{\text{New}}$
 - > **v)** find $\sigma = E(\sigma_{\text{New}}) = E(\sigma_{A/F}) * \text{Forecast}$
 - > **vi)** find Z stat for critical ratio
 - > **vii)** $Q = \mu + \sigma * Z$

- ◆ Use Std Normal Distribution
 - > i) find **critical ratio** → $C_O * F(Q) = C_U * (1 - F(Q))$ → $F(Q) = C_U / (C_O + C_U)$ [focus is on C_U b/c using distrib funct.]
 - > ii) find Q that satisfies → $F(Q) = C_U / (C_O + C_U)$
 - A) graph
 - B) equation for $F(Q)$ ~ we have for Standard Normal distribution (and some others)
 - 1) find standard normal order Q assuming $Z =$ critical ratio (**always pick larger Z**)
 - 2) convert Z from standard normal to normal distribution ($\sigma_{\text{new order}} = Z * \sigma_{\text{historical}}$)
 - 3) $Q = \mu + Z \sigma$
- b) Minimize Losses -
 - ◆ key = order less than π max Q
- c) Maximize customer service
 - ◆ consider - which measure to use depends on industry (fill rate = avg; in stock = conservative)
 - ◆ i) Target Fill Rate → $Q = \mu + \sigma (Z)$
 - > A) Recall →
 - Fill Rate = Expected Sales / μ
 - Expected Sales = $\mu -$ Expected Lost Sales = $\mu - \sigma * L(Z)$
 - > B) Combining equations → $L(Z) = (\mu / \sigma) (1 - \text{Fill Rate})$
 - > C) find Z – by solving for $L(Z)$ and examining $L(Z)$ table
 - > D) solve for Q
 - > **remember – choose lower Z stat b/c fill rate increases if expected loss decreases (190)**
 - this ensures that will achieve at least the target fill rate
 - ◆ ii) Target In Stock Probability → $Q = \mu + \sigma (Z)$
 - > A) find Z where $F(Z) =$ desired in-stock probability
 - > B) solve for Q
 - ◆ iii) Minimize Out of Stock Probability -
- remember
 - ◆ expected gain from selling 1st unit is much higher than for selling last unit
 - ◆ $E(\text{Gain})$ decreases relative to $E(\text{Loss})$ as we order more
 - ◆ order more until $E(\text{Gain}) = E(\text{Loss})$ of ordering next unit → $C_O * F(Q) = C_U * (1 - F(Q))$
- 3) Other Performance Measures
 - a) Expected Lost Sales = $\sigma * L(Z)$
 - ◆ i) find Z (for a given Q) → $Z = (Q - \mu) / \sigma$ (just reversing what did earlier)
 - > **remember** - link b/w Q & Z ~ probability that demand < Q is same as probability that standard normal distribution is less than Z
 - ◆ ii) find $L(Z)$ → look up Z in **Standard Normal Loss Function Table**
 - ◆ iii) multiply - $\sigma * L(Z)$
 - b) Expected Sales = $\mu -$ Expected Lost Sales
 - ◆ remember –
 - > each unit of Demand = either sale or lost sale
 - > Expected demand = μ
 - > Expected Sales ~ always less than Expected Demand (μ) b/c Expected lost sales will always be positive unless you order an unlimited Q (very expensive)
 - c) Expected Left Over Inventory = $Q -$ Expected Sales
 - ◆ remember
 - > $E(\text{Left Over Inventory})$ and $E(\text{Lost Sales})$ can both be positive (b/c dealing w/ expectation)
 - > In reality → can only be either left over inventory or lost sales
 - d) Expected $\Pi = (C_U * \text{Expected Sales}) - (C_O * \text{Expected Leftover Inventory})$

- ◆ remember
 - > $C_U = \text{opportunity cost of a lost sale} = \text{per sale margin}$

- e) **Expected GM %** = $1 - (\text{Cost} / \text{Expected Revenue}) = 1 - \text{Cost} / (P * \text{Expected Sales} + V * \text{Expected L/O Inv})$
 - ◆ **GM%** = $(\text{Revenue} - \text{Cost}) / \text{Revenue} = 1 - C / R$
 - > never higher than $\sim 1 - C / P$
 - ◆ **Revenue** = $(P * \text{Sales}) + (V * \text{left over inventory})$ (~ where V = salvage value)
 - ◆ **Cost** = $c * Q$
- f) **Expected Fill Rate** = $\text{Expected Sales} / \text{Expected Demand} (\mu)$
 - ◆ **Fill Rate** = $\text{Sales} / \text{Demand}$ (~ % of demand that is satisfied = sales)
 - ◆ measures - avg customer service (treats all customers equal)
- g) **In-Stock Prob** = $F(Q)$ (~ distribution function)
 - ◆ ~ probability store ends season satisfying all demand
 - ◆ ~ probability that $Q_D < F(Q)$
 - ◆ ~ fill rate at end of season (i.e., prob that last customer demand c/b satisfied)
 - ◆ measures - customer service conservatively (i.e., b/c evaluates when at its worst)
 - ◆ process -
 - > i) find Z stat for given order $Q \rightarrow Z = (Q - \mu) / \sigma$
 - > ii) find $F(Q) \sim$ the % for the Z stat
- h) **Out of Stock Probability** = $1 - F(Q)$ (~ 1- in stock probability)

- **Remember**

- Newsvendor model requires - $p + c + v$ + demand forecast w/ probability for every possible demand
- Actual Demand = only historic sales made at regular P (~ discounted sales are not real demand)
- Prior data (A/F) \rightarrow provides distribution (standard deviation)
- Profit Maximizing Order $Q \leftrightarrow$ Expected Demand (μ) \rightarrow if $C_U > C_O \rightarrow Q >$ Expected Demand
- separate forecasting from order Q decision (means two products w/ same mean forecast can have different $\pi \max Q$)
- Opportunity Costs ~ just as important as explicit costs

VIII Quick Response w/ Reactive Capacity (Make to Order)

- **Mismatch Cost**

- what is it ~ measures cost of supply / demand mismatches ~ the most you are willing to pay for perfect info
- calculation -
 - a) = $(C_O * \text{Expected Leftover Inventory}) + (C_U * \text{Expected Lost Sales}) = \text{Tangible} + \text{Intangible Costs}$
 - b) = $E(\pi)$ with a perfect forecast - $E(\pi)$ with actual forecast
 - ◆ $E(\pi)$ with a perfect forecast = $(p-c) * \text{Expected Demand} (\mu) = C_U * \mu$ (~ b/c Expected Demand = Q)
 - ◆ $E(\pi)$ with actual forecast = using Newsvendor Model
 - c) = Maximum Π - Expected Π (Newsvendor)
 - ◆ **Maximum Π** = $(p-c) * \mu = C_U * \mu$
 - d) = **mismatch tax rate** = mismatch cost / unit of Demand = $C_U * [f(Z) / F(Z)] * [\sigma / \mu]$
 - ◆ tells the per unit % cost of not having perfect information (also maximum benefit from perfect info)
 - ◆ $f(Z) / F(Z)$
 - > ~ decreases as Q increases (if we graph the ratio for Z)
 - > ~ decreases as critical ratio increases (b/c critical ratio drives Q)
 - > **Key Point - product with lower critical ratio has higher mismatch tax rate**
 - ◆ σ / μ = coefficient of variation
 - > ~ good measure of variability of demand
 - > needed b/c σ alone does not capture the relative variability in demand (b/c relative size of μ is not captured)
 - > ROT - low < 0.25; high > 0.75
 - > reflects quality of firms forecasting ability
 - > recall - coefficient of variation w/ exponential distribution = 1 (always) (differs from normal distribution)

- ◆ key points – mismatch cost increases if either
 - > **a)** critical ratio decreases
 - > **b)** demand variability increases (as captured in CoV)

- evaluation – compare to Expected Demand (μ) (b/c everything else depends on μ)
 - assume – $Q = \Pi \max Q$ (this minimizes the mismatch cost)
 - mismatch cost / unit of Demand = $C_U * [f(Z) / F(Z)] * [\sigma / \mu]$
 - ordering the $\pi \max Q$ minimizes the mismatch cost
 - $E(\Pi \text{ Newsvendor}) = \text{Maximum } \Pi - \text{Mismatch Cost}$
 - $C_U = \text{maximum } \Pi / \text{unit of expected demand}$
 - $\text{Expected } \Pi / \text{Unit of demand} = C_U - C_U * \text{Mismatch Tax Rate}$
- why – understand the per unit supply / demand mismatch cost

- **Make to order (reduces Mismatch Costs)**

- pros – reduces cost from lost sales / excess inventory
- cons – expensive + time lags are unavoidable + idle capacity from queuing theory (must have idle capacity to eliminate wait time, but idle capacity is supply / demand mismatch)

- **Reactive Capacity**

- what is it – capacity allowing a firm to make one added replenishment during the season
- why bother – is intermediate solution less costly than make to order but better than newsvendor

- Types

- **A) Unlimited Reactive Capacity** –

- ◆ cons ~ expensive (to mfr who passes cost to retailer - designer) + complex (much more difficult to choose order Q's with multiple replenishments) + does not help if over order
- ◆ simplifying assumptions – don't stock out before second replenishment + after observe initial sales → can predict Demand perfectly (may be close to true depending on historical results)
- ◆ Key – use Newsvendor to make initial order
- ◆ C_O = same as newsvendor
- ◆ C_U = premium must pay for under ordering in a prior round
- ◆ **Maximum Π** = same = $(P-C) * \mu$
- ◆ **Expected Π** = Maximum $\Pi - (C_O * \text{Expected Left Over Inventory}) - (C_U * \text{Expected Lost Sales})$
- ◆ **Expected Replenishment Q** = Newsvendor's Expected Lost Sales
- ◆ **Π max initial order Q** –
 - > **i)** find C_O & C_U (see above)
 - > **ii)** calculate critical ratio
 - > **iii)** find Z value for critical ratio (Standard Normal Distribution Function $F(Z)$)
 - > **iv)** convert Z value into order quantity for actual demand distribution function → $Q = \mu + \sigma(Z)$
- ◆ metrics (measuring value of replenishment) ~ % Π increases + reduction in mismatch cost + Fill Rate increase + reduction in leftover inventory

- **B) Limited Reactive Capacity** –

- ◆ key – postpone Supply until after Demand is known
- ◆ cons – capacity limit on replenishment + minimum order Q on replenishment
- ◆ simplifying assumptions – don't stock out before second replenishment + after observe initial sales → can predict Demand perfectly (may be close to true depending on historical results) + order once constraint (to ensure can get above - and can ignore - the minimum order Q in replenishment)
- ◆ remember
 - > anything in initial order = Newsvendor Optimal Order Q ~ generates Newsvendor Expected Π
 - > anything in replenishment ~ generates max Π
 - > key – choose initial order - use either
 - **a) mismatch quantity ratio**
 - = mismatch cost / Q
 - initial order s/b those items with the lowest ratio

- **b)** enumerate all possible initial order permutations and compare for max Π
- > ideal candidate for initial order = low P, low GM, low CoV, high critical ratio

- **Remember**

- $E(\Pi \text{ Newsvendor}) < E(\Pi \text{ reactive capacity}) < E(\Pi \text{ perfect info})$
 - consider this if reactive capacity cost difficult to calculate
- information is valuable only if you are willing and able to react to it
- products w/ low gross margins tend to have low mismatch quantity ratios
- products w/ low CoV and high critical ratios have low mismatch tax rates → have low mismatch –quantity ratios

IX. Service Levels & Lead Times (Order Up to Model)

- **Situation** – obsolescence is not problem + not limited to single production run
- **Goal** – balance too much inventory w/ too little
- **Solution** = Order up to Model (~ 1-for-1 ordering policy)
 - Defn
 - **period** ~ order interval (assume all are same)
 - **lead time** (ℓ) ~ number of periods required to receive orders (always measured in periods ~ matches frequency in which orders can be made / orders received)
 - >
 - **backorders** = (negative on-hand inventory)
 - ◆ ~ when demand arises and no units in current inventory
 - ◆ caused by – excess demand over short time (temp supply / demand imbalance)
 - ◆ causes a **stock out** (see below)
 - **inventory position** = on-order inventory + on-hand inventory
 - ◆ on-order inventory ~ ordered in prior periods but not received (may be 0, negative only if returning inv)
 - ◆ on-hand inventory = $S - [\mu (\ell + 1)]$
 - > ~ [order up to level - expected avg demand over (lead time + 1)]
 - > ~ [periods units in inventory - backorders]
 - **order up to level** (S) ~ max # units willing to have (on-hand + on order) = **base stock level** = inventory BoP
 - **Poisson distribution** – defined by mean (not σ); BUT $\sigma = (\mu)^{0.5}$
 - **stock out** –
 - ◆ arises if - (a) out of stock and (b) demand arrives
 - ◆ causes a backorder
 - **out of stock** – when have no inventory on-hand
 - **in stock** – if inventory available to meet all demand
 - Assumes – demand is random + same demand distribution applies in all periods + replenishment is at beginning of period (before any demand) + order always received in ℓ periods + no limit to Q + no obsolescence in inventory + all demand is filled eventually + all unmet Demand is backordered + no seasonality in Demand [expected demand (μ) for any period is same as any other]
 - Choosing demand distribution
 - **a) covering the relevant period** (day, week, mo, etc.) – convert from long to short if necessary (or vice versa w/ reciprocal)
 - ◆ Expected period demand (μ) = expected (monthly) demand / periods per (month)
 - ◆ σ of period demand = σ of (monthly) demand / [periods per (month)]^{0.5}
 - **b) with right distribution**
 - ◆ Poisson – discrete + only positive
 - ◆ Normal – continuous (more of issue w/ low avg demand) + 50% is negative
 - ◆ if inter-arrival time is exponential distribution → # arrivals in fixed interval is a Poisson distribution
 - Quantity per period (Q) = **S – Inventory position** = last period's demand where
 - ◆ **inventory position** = on-order inventory + on-hand inventory
 - ◆ **S** - depends on the target metric . . .

- minimize inventory, maintain target in-stock probability → simply eval in-stock prob for increasing S until > than desired probability
- minimize inventory, maintain target F/R → eval E(F/R) at increasing S until > than desired probability
 - ◆ = $E(F/R) \leq E(\text{demand in one period}) * (1 - F/R)$
 - ◆ **BUT** → E(B/O) table depends on (t+1) periods
- minimize inventory holding costs (h)
 - ◆
- maintain minimum service level (w/ minimum F/R)
 - ◆

➤ Performance Measures (Poisson Distribution) -

- In-stock Probability = Prob [demand over (t+1) periods ≤ S] = 1- Stock-out Probability
 - ◆ Expected demand over (t+1) periods = $\lambda = (t+1) * \mu$ demand per period
 - ◆ Probability = Poisson (S, expected demand over (t+1) periods, 1) → see table
- Stock-Out Probability = 1-Prob [$\lambda \leq S$] = Prob [$\lambda > S$]
 - ◆ works better if service level target is high (~ stock outs are rare)
- Expected Backorder (at end of any period) = $[\lambda - S] + \sum_{s=1}^{\lambda} [\text{prob}(\lambda \leq (S-1))]$
 - ◆ = $[\lambda - S] + \sum_{s=1}^{\lambda} [\text{Poisson}(S_{(0 \text{ to } S-1)}, \lambda, 1)]$ (~ expected lost sales in NV)
- Expected Fill Rate = $1 - [E(B/O) / E(\text{Demand in one period})]$
 - ◆ ignore S = 0 → F/R of 100% is unreasonable
 - ◆ theory ~ # of 1 period customers = μ in one period → # customers not served in period = E(B/O) → $[E(B/O) / E(\text{Demand})]$ = fraction of customers not served → $1 - [E(B/O) / E(\text{Demand})]$ = fraction of customers that are served
 - ◆ remember – lead time influences number of customers not served (numerator) / does not influence demand in one period
- Expected Inventory = $S - \lambda + E(B/O)$
 - ◆ cf pg. 31 – fill this in
 - ◆ $\text{Expected Inventory}_{\text{BoP}} = E(\text{Demand in one period}) + E(\text{inventory})$
 - ◆ $\text{Expected Inventory}_{\text{random t}} = [E(\text{Demand in one period}) / 2] + E(\text{inventory})$
- Expected on-order inventory = pipeline inventory = $E(\text{Demand for one period}) * t$
 - ◆ cf Little's Law ~ Inventory = FR * FT

➤ Performance Measures (Normal Distribution) -

• Lead Time Issues

- reducing lead time → reduces expected inventory (especially as service requirement increases) + reduces pipeline inventory

• Remember

- Order up to is pull system (NV is push system) = better when avg demand is steady
- Poisson Distribution – is different for each mean
- after place order, inventory position → increases to S