

Operations & Supply Planning
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Demand Forecasting

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Agenda

- Understanding Demand: nature and components
- Forecasting - motivation
- Forecasting techniques
 - Stationary series
 - Error measures
 - Trend
 - Seasonality
- Forecasting in the real world

Opening question

What is the key role of an operations manager?

Making the right product/service available at the right place at the right time for the right customer at the right price

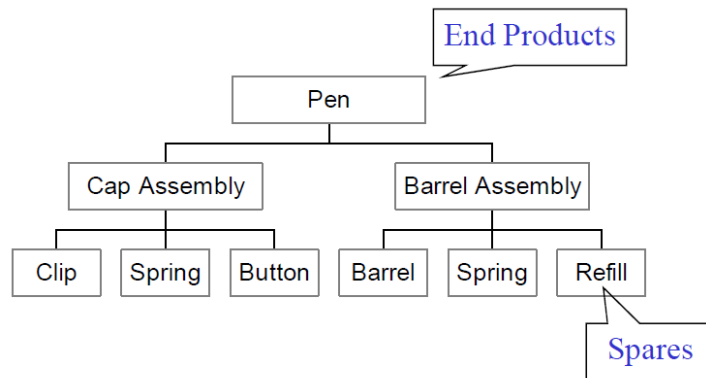
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Matching Supply with Demand



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Independent – Dependent Demand



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Demand Management

- Not much a firm can do about dependent demand
 - It is demand that must be met
- There is a lot a firm can do about independent demand
 1. Take an active role to influence demand
 2. Take a passive role and respond to demand

Understanding Independent demand!



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Independent Demand – Nuances

- What are the characteristics of demand?
 - Demand is not the same as sales.
 - It's random and uncertain
 - Depends several factors
 - Time of the year
 - Economic environment
 - Weather
 - Price fluctuation – upward or downward
 - Sales Promotion
 - Other product demand

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What Forecasting is not...



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What is Forecasting?



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What is forecasting?



Predicting the future based on past data!!

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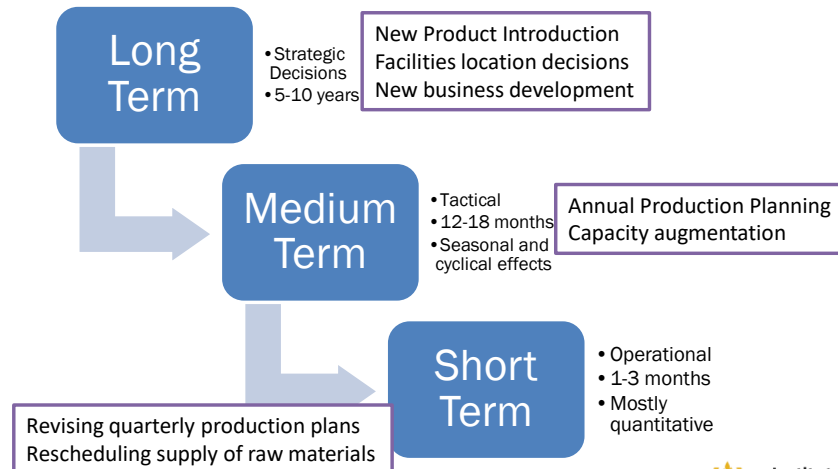
Questions to ponder

- Why forecast?
 - It's expensive not to. Why?
- What to forecast?
 - Level of aggregation.
 - What use is a forecast to you?
- How to forecast?
 - What methodology are you going to follow?
 - What data do you need?



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Forecasting time horizon



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Subjective Forecasts

- Sales force composites
 - Salesforce provides optimistic, pessimistic and most likely forecasts
- Jury of executive opinions
 - Top executives from different functional areas
- Customer surveys
 - Select a sample population of customers
- The Delphi method

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Subjective forecasts

- Form group of experts – anonymous
- Send questionnaires
- Collect and tabulate findings
- Redistribute and seek justification of outer quartiles
- Re-collect and tabulate findings
- Repeat until consensus is reached

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Common biases in subjective forecasting

- Inconsistency
 - Applying different decision criteria in similar situations
- Conservatism; Recency
- Anchoring; Confirmation bias
- Uncertainty underestimation
- Reputation effect

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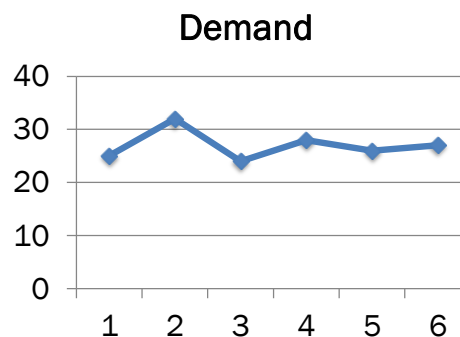
Objective forecasting methods

- Extrapolative methods or time series models
 - Using past data of same metric to predict future
- Causal models
 - Using cause effect relationship

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Demand for Diapers at a local store

- Month 1 - 25
- Month 2 - 32
- Month 3 - 24
- Month 4 - 28
- Month 5 - 26
- Month 6 - 27



How much should the store keep in stock for the next month?

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Next month demand

- Average
- Average of last four periods
- Last period demand
- Weighted average of last five periods
- Up down up down
- Remove outliers, find average

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Components of Demand

1. Average demand for a period of time
2. Trend
3. Seasonal element
4. Cyclical elements
5. Random variation
6. Autocorrelation

Notation Conventions

- Let $A_1, A_2, \dots, A_n, \dots$ be the past values of the series to be predicted (demands?). If we are making a forecast during period t (for the future), assume we have observed A_t, A_{t-1} etc.
 - Let $F_{t,t+\tau}$ = forecast made in period t for the demand in period $t + \tau$ where $\tau = 1, 2, 3, \dots$
 - Then $F_{t-1,t}$ is the forecast made in $t-1$ for t and
 - $F_{t,t+1}$ is the forecast made in t for $t+1$. (one step ahead) Use shorthand notation $F_t = F_{t-1,t}$

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Simple averages

- The arithmetic average of the n most recent observations. For a one-step-ahead forecast:

$$F_t = (1/N) (A_{t-1} + A_{t-2} + \dots + A_1)$$

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Moving Averages

- The arithmetic average of the n most recent observations. For a one-step-ahead forecast:

$$F_t = (1/N) (A_{t-1} + A_{t-2} + \dots + A_{t-n})$$

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Summary of Moving Averages

- *Advantages of Moving Average Method*
 - Easily understood
 - Easily computed
 - Provides stable forecasts
- *Disadvantages of Moving Average Method*
 - Requires saving lots of past data points: at least the N periods used in the moving average computation
 - Lags behind a trend
 - Ignores complex relationships in data

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What about Weighted Moving Averages?

- This method looks at past data and tries to logically attach importance to certain data over other data
- Weighting factors must add to one
- Can weight recent higher than older or specific data above others
 - If forecasting staffing, we could use data from the last four weeks where Tuesdays are to be forecast.
 - Weighting on Tuesdays is: T_{-1} is .25; T_{-2} is .20; T_{-3} is .15; T_{-4} is .10 and Average of all other days is weighed .30.

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Exponential Smoothing Method

- A type of weighted moving average that applies declining weights to past data.
1. $F_t = \alpha (A_{t-1}) + (1 - \alpha) (F_{t-1})$
Forecast = α (most recent demand) + $(1 - \alpha)$ (last forecast)
- OR -
 2. $F_t = F_{t-1} + \alpha (A_{t-1} - F_{t-1})$
Forecast = last forecast + α (last forecast error)
- where $0 < \alpha < 1$ and generally is small for stability of forecasts (around .1 to .2)

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Exponential Smoothing (cont.)

In symbols:

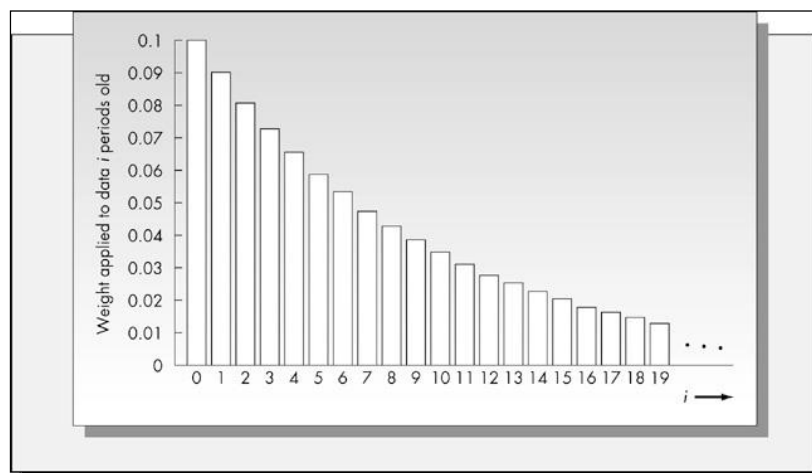
$$\begin{aligned}
 F_{t+1} &= \alpha A_t + (1 - \alpha) F_t \\
 &= \alpha A_t + (1 - \alpha) (\alpha A_{t-1} + (1 - \alpha) F_{t-1}) \\
 &= \alpha A_t + (1 - \alpha) (\alpha A_{t-1} + (1 - \alpha)^2 (\alpha A_{t-2} + \dots)
 \end{aligned}$$

- Hence the method applies a set of exponentially declining weights to past data. It is easy to show that the *sum of the weights* is exactly one.

$$\{\text{Or: } F_{t+1} = F_t - \alpha (F_t - A_t)\}$$

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Weights in Exponential Smoothing:



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Effect of seed forecast on the overall forecasting

- What should be the value of the forecast of the first period?
- Does it really matter which value is taken?

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Effect of α value on the Forecast

- Small values of α means that the forecasted value will be stable (show low variability)
 - Low α increases the lag of the forecast to the actual data if a trend is present
- Large values of α mean that the forecast will more closely track the actual time series

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Comparison of MA and ES

- Similarities
 - Both methods are appropriate for stationary series
 - Both methods depend on a single parameter
 - Both methods lag behind a trend

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Comparison of MA and ES

- Differences
 - ES carries all past history (forever!)
 - MA eliminates “bad” data after N periods
 - MA requires all N past data points to compute new forecast estimate while ES only requires last forecast and last observation of ‘demand’ to continue

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Accuracy of Forecasts

- Forecast error
 - $SFE = \sum_{i=1}^t (D_t - F_t)$
- Mean Absolute Deviation (MAD)
 - $MAD = \frac{1}{t} \times \sum_{i=1}^t |D_t - F_t|$
- Mean Absolute Percentage error (MAPE)
 - $MAPE = \frac{1}{t} \times \sum_{i=1}^t \frac{|D_t - F_t|}{D_t} \times 100$
- Mean Squared error
 - $MSE = \frac{1}{t} \times \sum_{i=1}^t (D_t - F_t)^2$

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Tracking Signal

- The tracking signal (TS) is a measure that indicates whether the forecast average is keeping pace with any genuine upward or downward changes in demand
- Depending on the number of MAD's selected, the TS can be used like a quality control chart indicating when the model is generating too much error in its forecasts

$$TS = \frac{RSFE}{MAD} = \frac{\text{Running sum of forecast errors}}{\text{Mean absolute deviation}}$$