

*D-Lab*

# D-Lab: Supply Chains

## *Inventory Management, Part II*



October 22, 2014  
Annie Chen

### Agenda:

- Review: **Economic Order Quantity (EOQ)**
- Single-period: **Newsvendor Model**
- Multi-period:
  - **Base Stock Policy**
  - **(R,Q) Policy**
- Project discussion

## Operations Management Seminar



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## Data-driven Operations Research Analyses in the Humanitarian Sector

### **Abstract:**

We briefly discuss six projects in the humanitarian sector:

- (1) allocating food aid for undernourished children using data from a randomized trial in sub-Saharan Africa,
- (2) allocating food aid for undernourished children using data from a nutrition project in Guatemala,
- (3) analyzing the nutrition-disease nexus in the case of malaria,
- (4) allocating aid for health interventions to minimize childhood mortality,
- (5) assessing the impact of U.S.'s failure to use local and regional food procurement on childhood mortality, and
- (6) deriving individualized biometric identification for India's universal identification program.

# Types of inventory models

- **Demand:** constant, deterministic, stochastic
- **Lead times:** “0”, “>0”, stochastic
- **Horizon:** single period, finite, infinite
- **Products:** one product, multiple products
- **Capacity:** order/inventory limits, no limits
- **Service:** meet all demand, shortages allowed

**EOQ**

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EOQ

Newsvendor



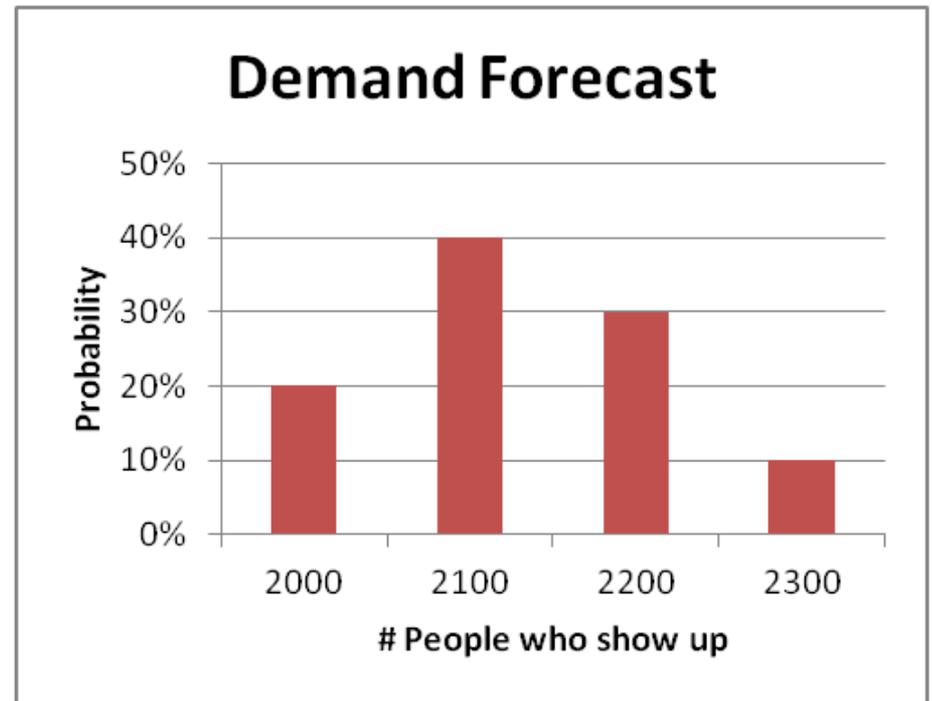
# Example: Vaccine Clinic

In preparation for a one-day flu shot clinic, you need to decide the quantity of vaccines to order in advance.

You are given a **probabilistic forecast** based on historical demand data.

**Exercise:** What is the expected demand?

Sol.  $E[D] = \sum_d d \cdot Prob(D = d)$



# Example: Vaccine Clinic

Optimal order quantity depends on the **forecast** and the **costs**:

$$\text{Expected cost } f(Q) = c_o E[\text{overage}(Q)] + c_u E[\text{underage}(Q)]$$

What is the **overage cost**  $c_o$  and **underage cost**  $c_u$  in the following cases?

1. Suppose vaccines costs you  $c = \$1$  per unit if ordered in advance. For every flu shot you give, you are paid  $r = \$5$ . At the end of the day, leftover vaccines have to be thrown away, so the salvage value is  $v = \$0$ .

$$\Rightarrow c_o = c - v = 1; \quad c_u = r - c = 4 \text{ (value of lost sales)}$$

2. Suppose  $c = \$1$ ,  $r = \$5$ , but you can sell back leftover vaccines to a recycler for  $v = \$0.5$  each.

$$\Rightarrow c_o = c - v = 0.5; \quad c_u = 4$$

3. Suppose again  $c = \$1$ ,  $r = \$5$ ,  $v = \$0$ . In addition, suppose you cannot turn people away if you run out of pre-ordered vaccines; instead, you can make **emergency orders** at double the cost,  $\$2$ .

$$\Rightarrow c_o = 1; \quad c_u = 2 - 1 = 1$$

# Optimal Order Quantity

Expected cost  $f(Q) = c_o E[\text{overage}(Q)] + c_u E[\text{underage}(Q)]$

How do you compute the optimal order quantity?

- **Sol 1:** Brute-force enumeration
  - Calculate  $f(Q)$  for all possible  $Q$
  - See Excel demo
  - May be cumbersome if there are lots of possible  $Q$
- **Sol 2:** Take derivative of  $f(Q)$ 
  - See standard inventory textbooks; a bit tedious
- **Sol 3: Incremental Analysis**

# Optimal Order Quantity

- Incremental analysis:
  - What is the cost/benefit of ordering one additional unit ( $Q \rightarrow Q+1$ )?
    1. Benefit: if the additional unit is used up, you make an extra  $c_u$ . This event happens with probability  $P(D>Q)$ .  
=> Expected benefit:  $P(D>Q) c_u$
    2. Cost: if the additional unit is not used up, you wasted an investment of  $c_o$ . This event happens with probability  $P(D\leq Q)$ .  
=> Expected cost:  $P(D\leq Q) c_o$
  - If the expected benefit outweighs the expected cost, you'd want to continue increasing the order quantity  $Q$ . Conversely, if cost outweighs benefit, you'd want to continue decreasing  $Q$ .
  - At the optimal  $Q$ , the benefit and cost balance each other:
$$P(D>Q) c_u = P(D\leq Q) c_o$$
  - Collecting the terms, we obtain the optimality condition:

$$P(D \leq Q) = \frac{c_u}{c_u + c_o} \left. \vphantom{\frac{c_u}{c_u + c_o}} \right\} \text{Critical ratio}$$

# Example: Vaccine Clinic

Optimal order quantity depends on the forecast and the costs:

$$\text{Expected cost } f(Q) = c_o E[\text{overage}(Q)] + c_u E[\text{underage}(Q)]$$

What is the **critical ratio** in the following cases?

1. Suppose vaccines costs you  $c = \$1$  per unit if ordered in advance. For every flu shot you give, you are paid  $r = \$5$ . At the end of the day, leftover vaccines have to be thrown away, so the salvage value is  $v = \$0$ .

⇒  $c_o = c - v = 1$ ;  $c_u = r - c = 4$  (value of lost sales)

$$P(D \leq Q) = \frac{c_u}{c_u + c_o} = 0.8$$

2. Suppose  $c = \$1$ ,  $r = \$5$ , but you can sell back leftover vaccines to a recycler for  $v = \$0.5$  each.

⇒  $c_o = c - v = 0.5$ ;  $c_u = 4$

$$P(D \leq Q) = \frac{c_u}{c_u + c_o} = 0.89$$

3. Suppose again  $c = \$1$ ,  $r = \$5$ ,  $v = \$0$ . In addition, suppose you cannot turn people away if you run out of pre-ordered vaccines; instead, you can make emergency orders at double the cost,  $\$2$ .

⇒  $c_o = 1$ ;  $c_u = 2 - 1 = 1$

$$P(D \leq Q) = \frac{c_u}{c_u + c_o} = 0.5$$

Note: In this example, since we have a discrete probability, there is no  $Q$  that exactly matches the optimality condition; we need to check the two options that bound it. (This is much more efficient than having to check all possible options for  $Q$ !)

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EOQ

Newsvendor

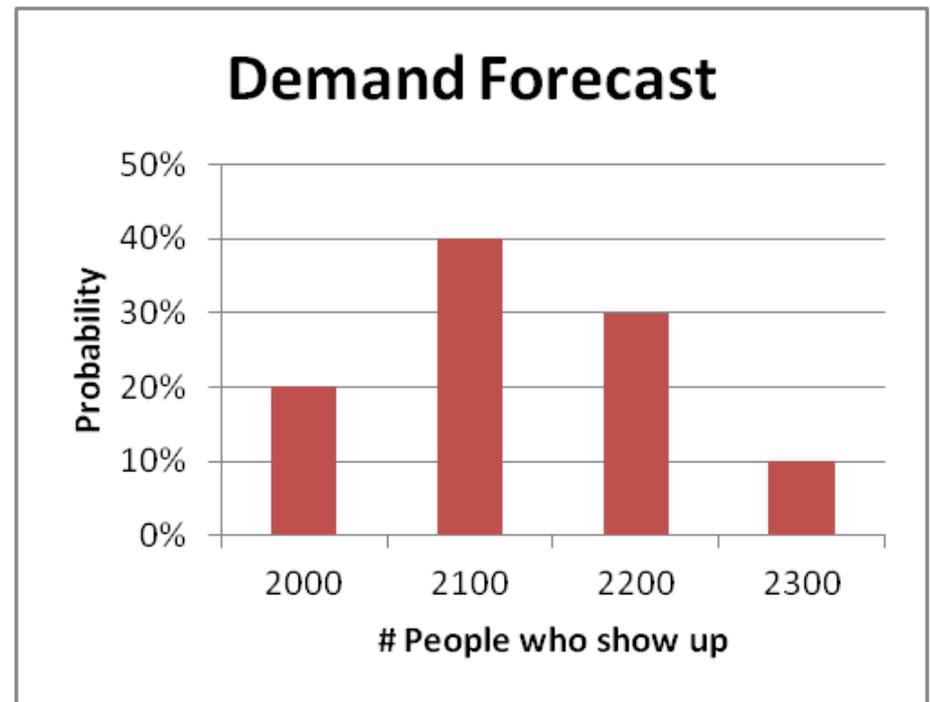
Continuous review: (R,Q)  
Periodic review: (T,S)

# Multi-Period & Stochastic Demand

- **Overage** is no longer a big deal
  - Leftover inventory can be used in the following periods (unlike that in the single-period case)
  - Cost of overage is **holding cost**
  - Possible economies of scale for **fixed ordering cost**
- **Underage** is more serious
  - Performance measure: **service level**
    - Service Level  $\alpha = \text{Prob}(\text{no stock-out})$**
  - Need to hold **safety stock** to achieve service level

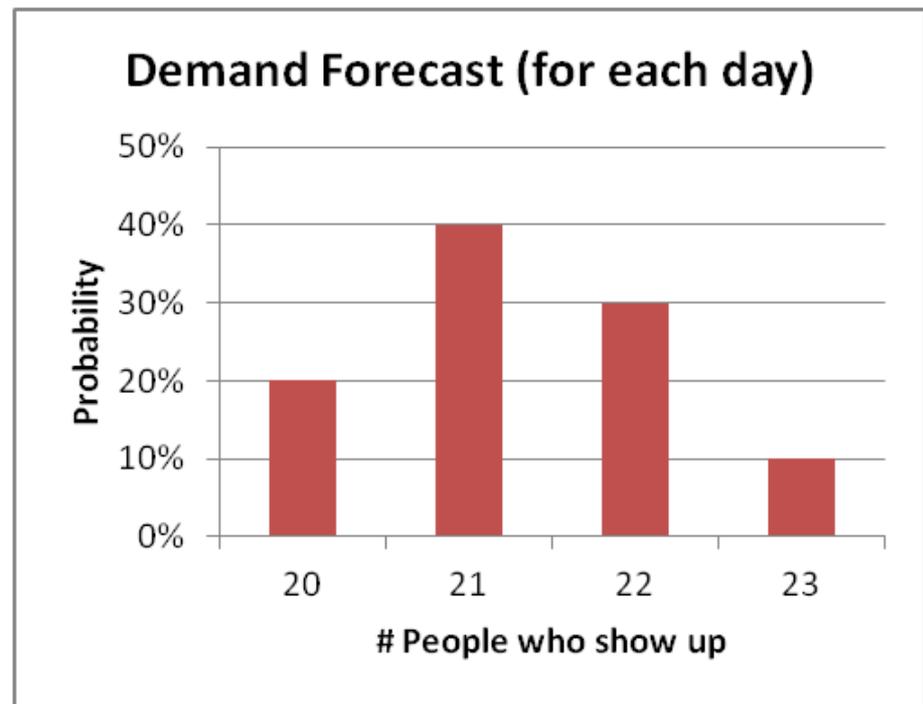
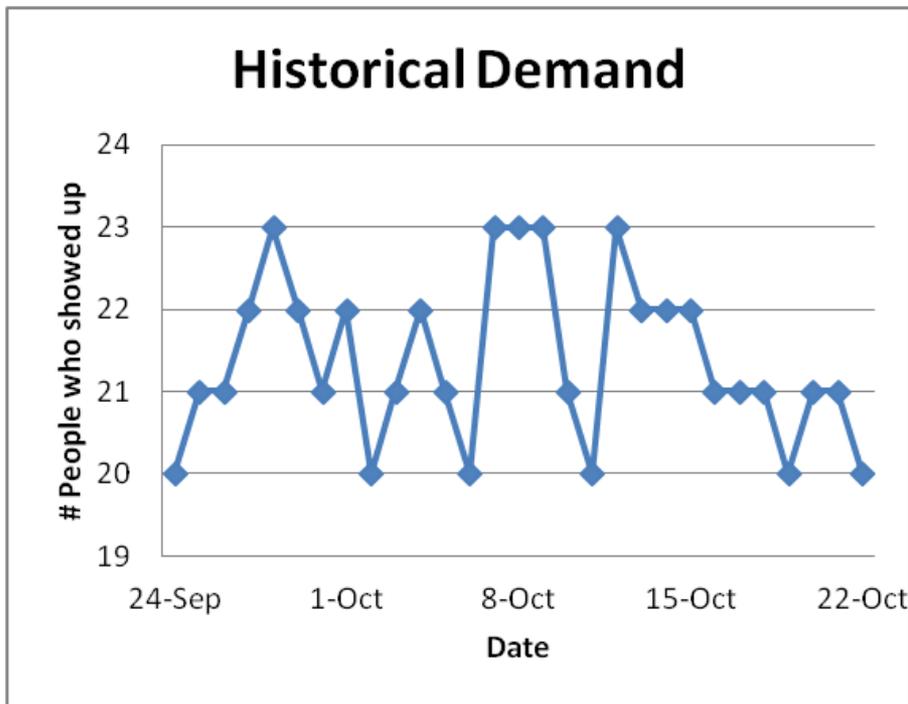
# Service Level

- What is the stock-out probability and service level if you ordered 2100?
- How much should you order to achieve a service level of 90%? 95%?



# Example: Vaccine Clinic

Suppose you are now managing a **daily, non-seasonal** vaccine clinic.



# Ordering for Multiple Periods

- Due to **economies of scale** (e.g. fixed cost per order, as in EOQ), it may be desirable to place one order to cover multiple periods

=> need to know the distribution of **demand over multiple periods**

- The **Central Limit Theorem** provides an approximation:

Let  $D_T =$  demand over  $T$  days, where the daily demand has mean  $\mu_D$  and std  $\sigma_D$

Then by the Central Limit Theorem,

$$D_T \rightarrow \text{Normal}(\mu, \sigma^2)$$

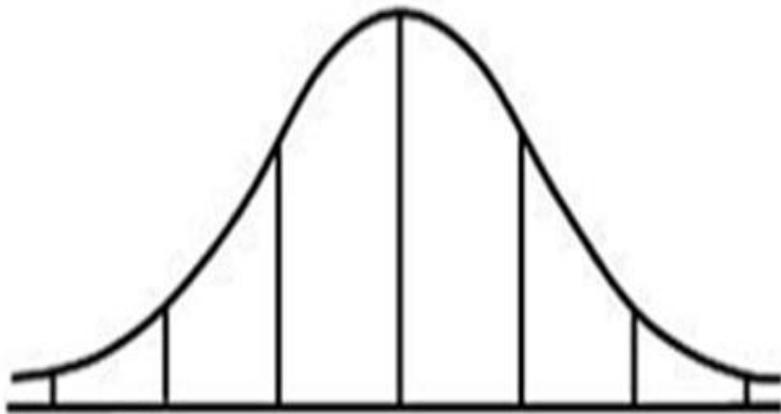
where  $\mu = T\mu_D$

$\sigma = \sqrt{T}\sigma_D$  (as a result of summing the variance:  $\sigma^2 = T\sigma_D^2$ )

# Example: Vaccine Clinic

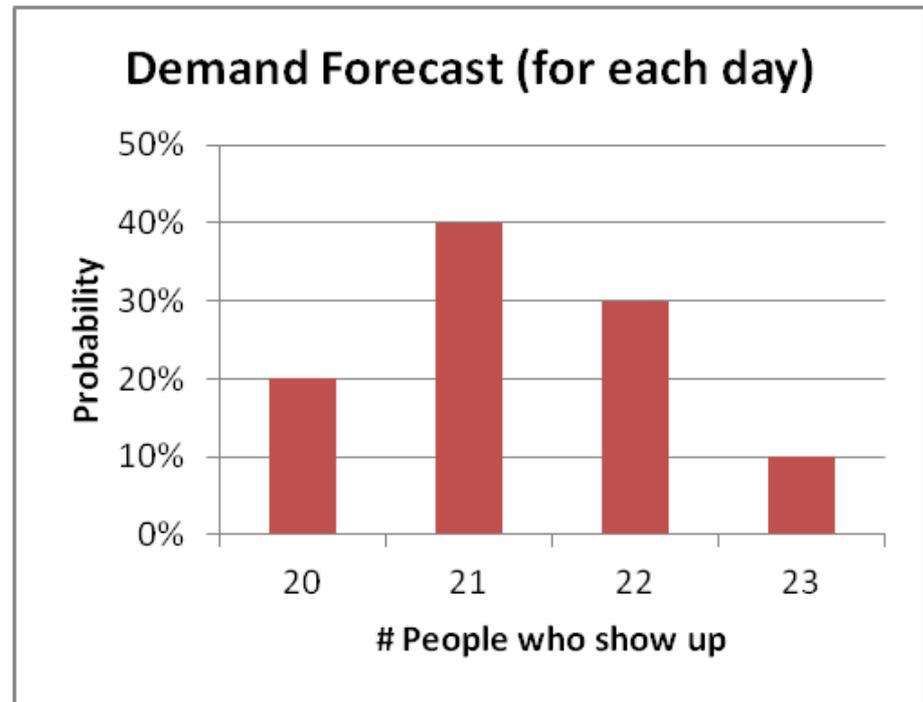
If you reorder every  $T=5$  days:

Aggregate demand:  $D_T \approx N(\mu_T, \sigma_T^2)$



$$\mu_T = 5 * 21.3 = 106.5$$

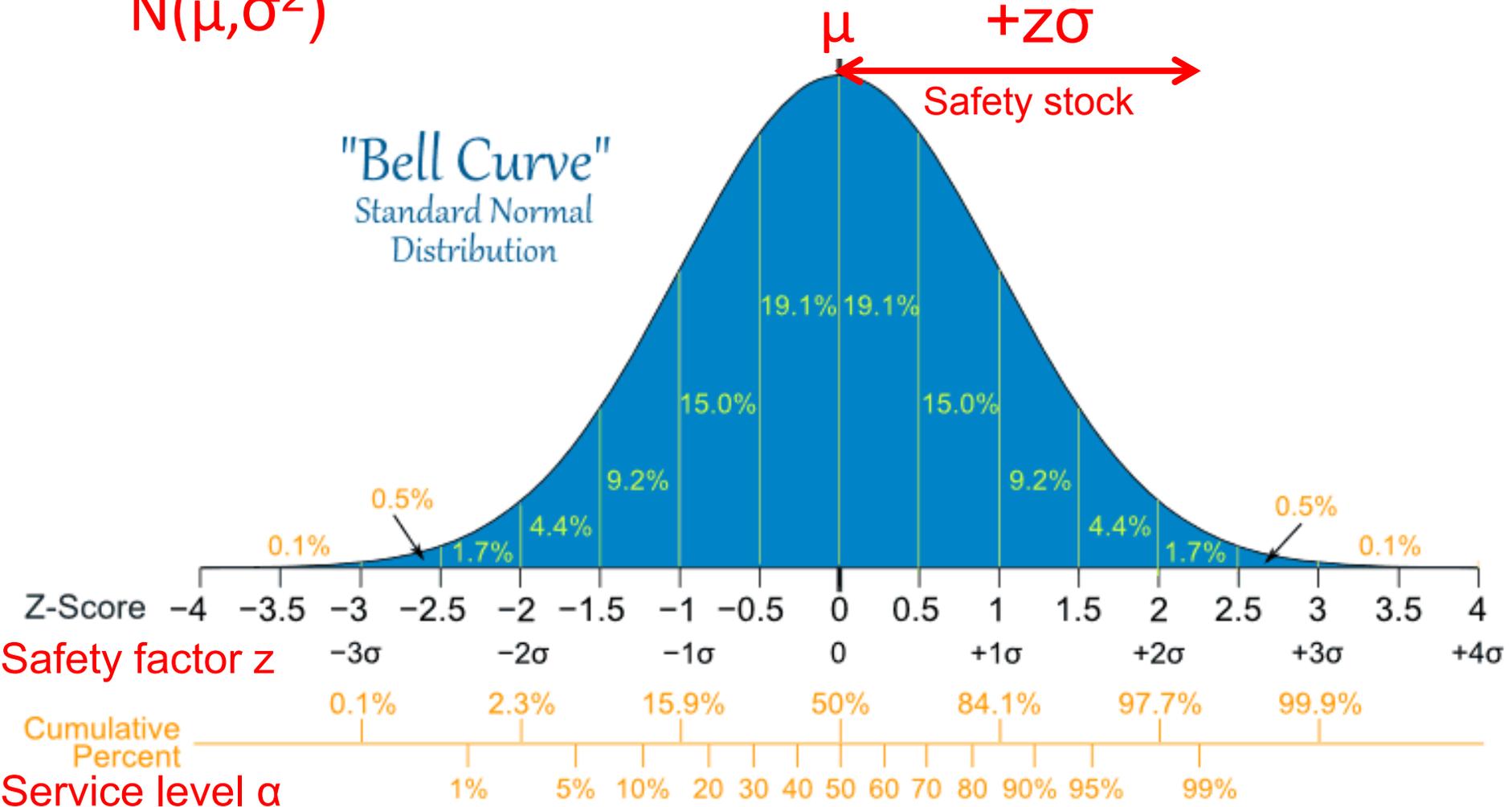
$$\sigma_T = \sqrt{5} * 0.9 = 2.01$$



$$\mu = 21.3, \sigma = 0.9$$

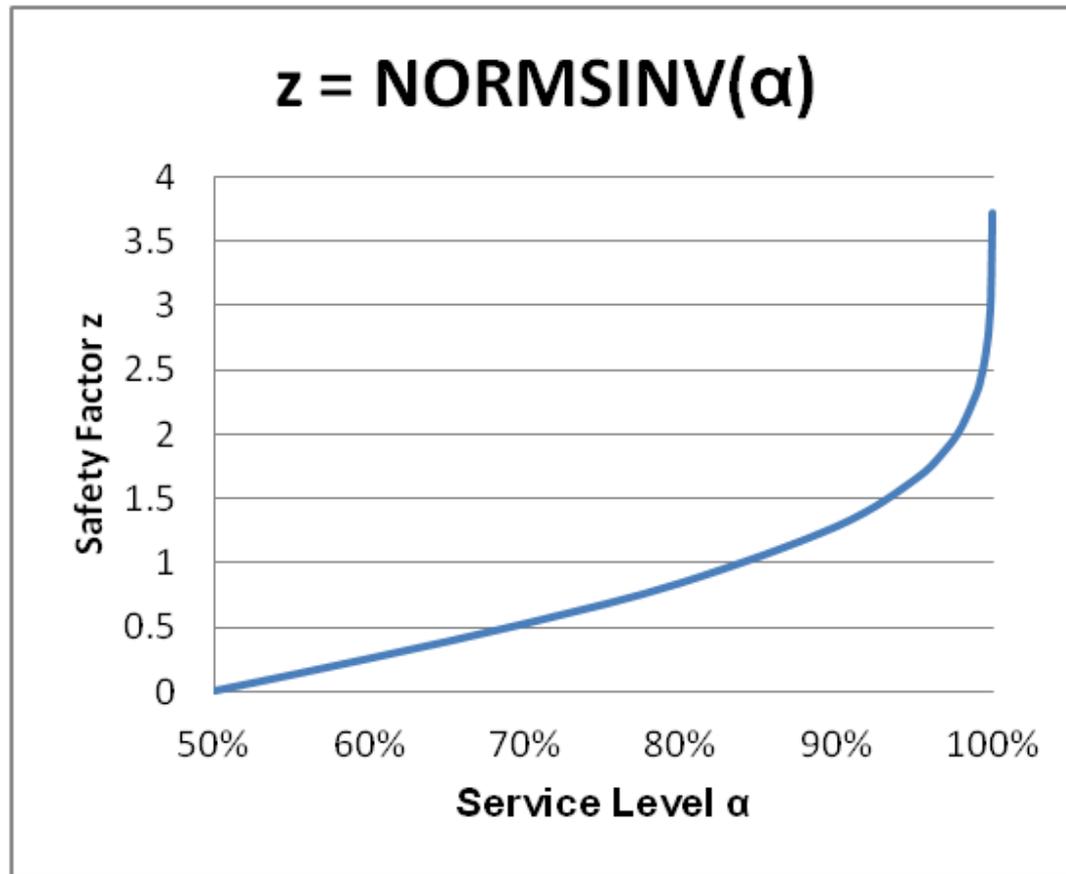
# Service Level for Normal Distribution

$$N(\mu, \sigma^2)$$



In Excel:  $z = \text{NORMSINV}(\alpha)$

# Tradeoff: Service Level vs. Safety Stock

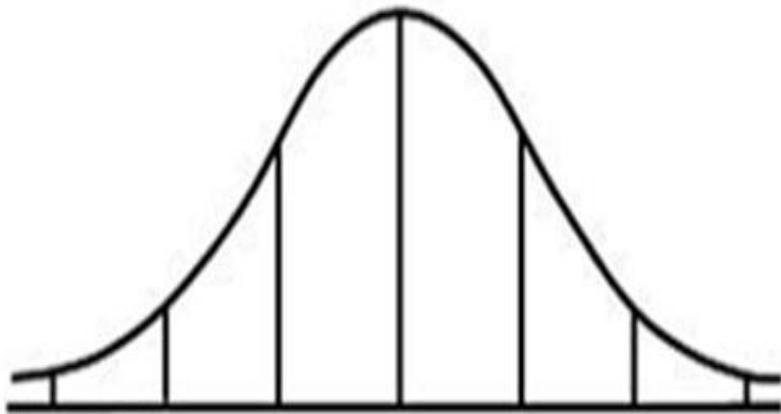


⇒ Relationship is **nonlinear** when the service level is close to 1; i.e., need **disproportionately high safety stock** to achieve very high service level

# Example: Vaccine Clinic

If you reorder every  $T=5$  days:

Aggregate demand:  $D_T \approx N(\mu_T, \sigma_T^2)$

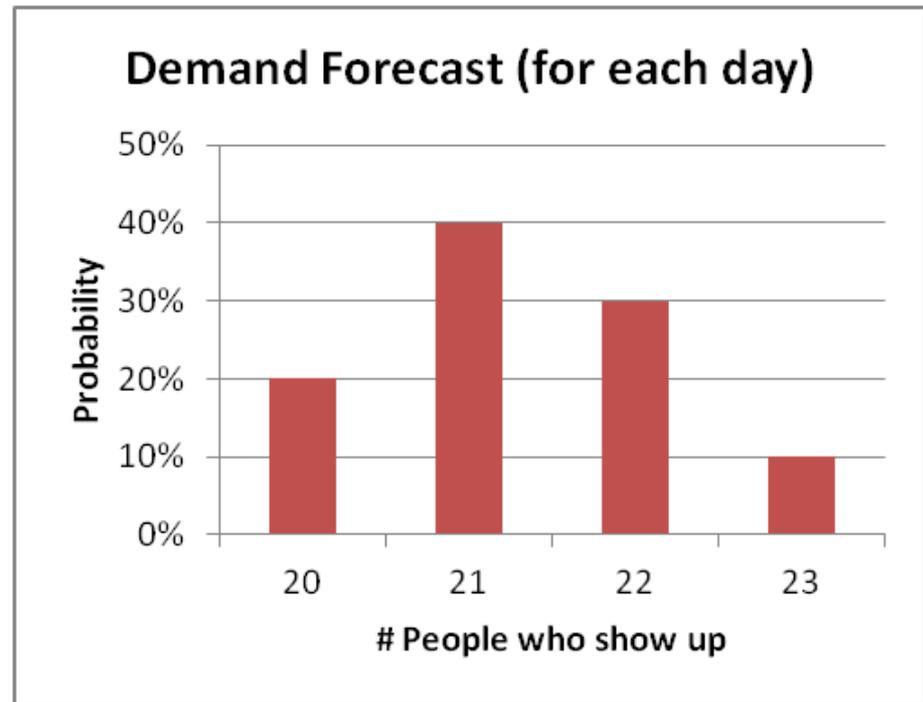


$$\mu_T = 5 * 21.3 = 106.5$$

$$\sigma_T = \sqrt{5} * 0.9 = 2.01$$

Service level  $\alpha = 97.7\%$  (safety factor  $z = 2$ )

$\Rightarrow$  Base stock level  $S = 106.5 + 4.02 = 110.5$



$$\mu = 21.3, \sigma = 0.9$$

# Base Stock Policy

1. Determine **review period T**
  - EOQ:  $T = \sqrt{\frac{2K}{h\mu}}$
2. Find aggregate demand over T
  - Use daily demand data & approximate with Central Limit Theorem =>  $N(\mu_T, \sigma_T)$
3. Find safety factor z
  - Given service level  $\alpha$ ,  $z = \text{NORMSINV}(\alpha)$
4. Compute base stock level  $S = \mu_T + z \sigma_T$

# Base Stock Policy with Lead Time

1. Determine review period  $T$ 
  - EOQ:  $T = \sqrt{\frac{2K}{h\mu}}$
2. Find aggregate demand over  $T+L$ 
  - Use daily demand data & approximate with Central Limit Theorem  $\Rightarrow N(\mu_{T+L}, \sigma_{T+L})$
3. Find safety factor  $z$ 
  - Given service level  $\alpha$ ,  $z = \text{NORMSINV}(\alpha)$
4. Compute base stock level  $S = \mu_{T+L} + z \sigma_{T+L}$

# Periodic vs. Continuous Review

- Period review:  
**Base stock policy**
- At each **review period  $T$** , order up to **base stock level  $S$**
- $T = \sqrt{\frac{2K}{h\mu}}$
- $S = \mu_{T+L} + z\sigma_{T+L}$
- Continuous Review:  
**(R,Q) policy**
- When inventory drops below **reorder point  $R$** , place new order with **order quantity  $Q$**
- $R = z\sigma_{T+L}$
- $Q = \sqrt{\frac{2K\mu}{h}}$

# Summary: Inventory Models

Model	EOQ	Newsvendor	Base Stock	(R,Q)
Decision variable	<ul style="list-style-type: none"> <li>•Order quantity Q</li> <li>•(Order period T)</li> </ul>	<ul style="list-style-type: none"> <li>•Order quantity Q</li> </ul>	<ul style="list-style-type: none"> <li>•Review period T</li> <li>•Order-up-to level S</li> </ul>	<ul style="list-style-type: none"> <li>•Reorder point R</li> <li>•Order quantity Q</li> </ul>
Demand	Constant	Stochastic	Stochastic	Stochastic
Lead time	0	0	L>0	L>0
Horizon	Infinite	Single	Infinite (periodic review)	Infinite (continuous review)
Optimal solution	$Q = \sqrt{\frac{2K\mu}{h}}$ $\left(T = \frac{Q}{\mu} = \sqrt{\frac{2K}{h\mu}}\right)$	$P(D \leq Q) = \frac{c_u}{c_u + c_o}$	$T = \sqrt{\frac{2K}{h\mu}}$ $S = \mu_{T+L} + z\sigma_{T+L}$	$R = z\sigma_{T+L}$ $Q = \sqrt{\frac{2K\mu}{h}}$

# Project discussion

Things to consider:

- What is the **inventory**?
  - Could be raw material, finished goods, workforce, etc.
  - E.g., oxygen, manure/flies, recyclable material/fleet of collectors, charcoal...
- What is the **setup**? Which **inventory model** might be appropriate for this setup?
  - Demand pattern: Constant or stochastic?
  - Order policy: One-shot or multi-period?
- What demand **data** is available?
  - If not available, what data should be collected?

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