

# **ISE 754: Logistics Engineering**

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# Topics

1. Introduction
2. Facility location
3. Freight transport
  - Midterm exam
4. Network models
5. Routing
6. Warehousing
  - Final project
  - Final exam

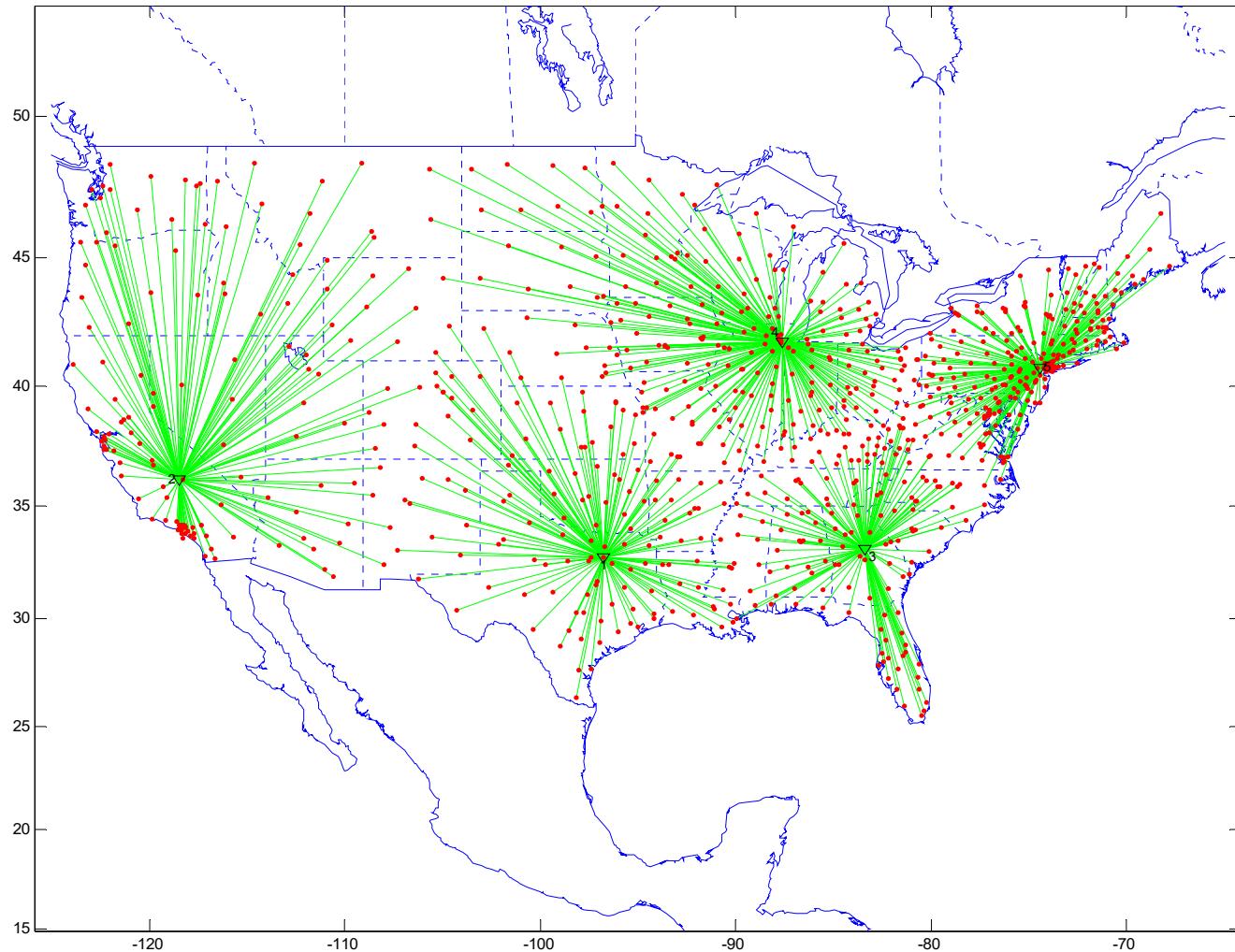
Outside the box

Inside the box

# Scope

- Strategic (years)
  - Network design
- Tactical (weeks-year)
  - Multi-echelon, multi-period, multi-product production and inventory models
- Operational (minutes-week)
  - Vehicle routing

# Strategic: Network Design



**Optimal locations for five DCs serving 877 customers throughout the U.S.**

# Tactical: Production-Inventory Model

	$c^p$	$c^i$	$c^s$	$\mathbf{0}$	$c^p$	$c^i$	$c^s$	$\mathbf{0}$
Product 1	Flow balance $x$	$y$						
	Capacity $x$			$\frac{-K}{k} \leq 0$				
		Setup $z$		$\frac{1}{k} \leq 0$				
							$0$	
Product 2			Flow balance $x$	$y$				
			Capacity $x$		$\frac{-K}{k} \leq 0$			
					$\frac{1}{k} \leq 0$			
Linking			$k_1$	$+$			$k_2$	$= 1$

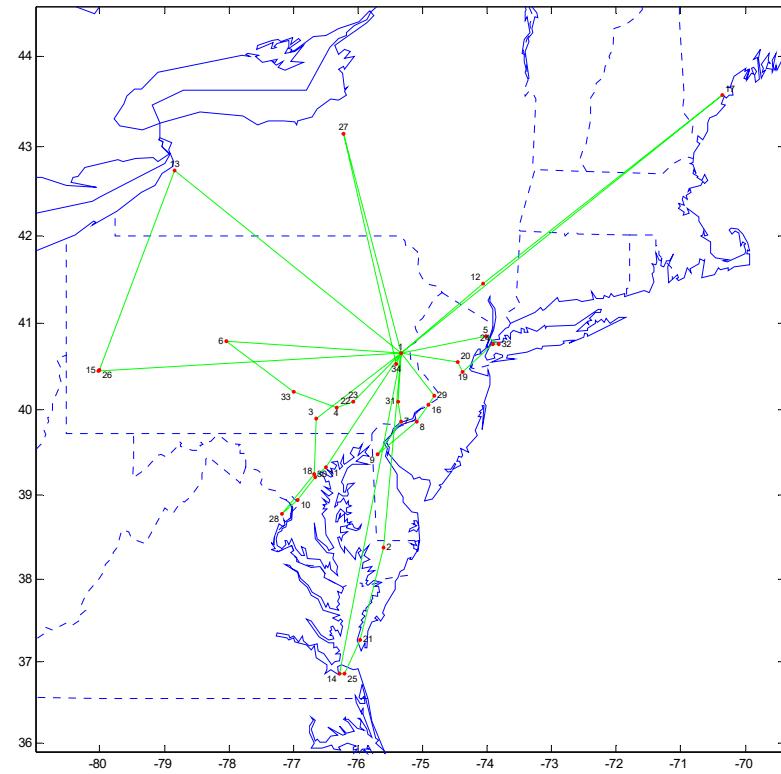
Constraint matrix for a 2-product, multi-period model with setups

# Vehicle Routing

**Eight routes served from DC in Harrisburg, PA**

**Route Summary Information**

Route	Load Weight	Route Time	Customers in Route	Layover Required
1	12,122	18.36	4	1
2	4,833	16.05	2	1
3	9,642	17.26	3	1
4	25,957	13.77	6	0
5	12,512	9.90	2	0
6	15,156	13.70	5	0
7	29,565	11.30	6	0
8	32,496	8.84	5	0
		109.18		3



**Detailed Route Information**

Route	1	Total							
		Start	L/D (hr)	Depart	Time (hr)	Zip Code			
1	23:29	0	23:29	0	18020				
14	2,328	7.51	7:00	399	7:00	0.59	7:35	8.1	23510
25	4,697	0.19	7:46	6	7:46	0.68	8:27	0.87	23502
21	3,682	1.12	9:34	37	9:34	0.64	10:13	1.76	23310
2	1,415	2.51	12:43	93	12:43	0.56	13:17	3.07	21801
1	0	4.57	17:51	196	17:51	0	17:51	4.57	18020
Total	12,122	15.89		731		2.47		18.36	

# Geometric Mean

- How many people can be crammed into a car?
  - Certainly more than one and less than 100: the average (50) seems to be too high, but the geometric mean (10) is reasonable

$$\text{Geometric Mean: } X = \sqrt{LB \times UB} = \sqrt{1 \times 100} = 10$$

- Often it is difficult to directly estimate input parameter X, but is easy to estimate reasonable lower and upper bounds (LB and UB) for the parameter
  - Since the guessed LB and UB are usually orders of magnitude apart, use of the arithmetic mean would give too much weight to UB
  - Geometric mean gives a more reasonable estimate because it is a logarithmic average of LB and UB

# Fermi Problems

- Involves “reasonable” (i.e.,  $\pm 10\%$ ) *guesstimation* of input parameters needed and back-of-the-envelope type approximations
  - Goal is to have an answer that is within an order of magnitude of the correct answer (or what is termed a *zeroth-order approximation*)
  - Works because over- and under-estimations of each parameter tend to cancel each other out as long as there is no consistent bias
- How many McDonald’s restaurants in U.S.? (actual 2013: 14,267)

Parameter	LB		UB	Estimate	
Annual per capita demand	1	$1 \text{ order/person-day} \times 350 \text{ day/yr} =$	350	18.71	(order/person-yr)
U.S. population				300,000,000	(person)
Operating hours per day				16	(hr/day)
Orders per store per minute (in-store + drive-thru)				1	(order/store-min)
Analysis					
Annual U.S. demand		$(\text{person}) \times (\text{order/person-yr}) =$	5,612,486,080		(order/yr)
Daily U.S. demand		$(\text{order/yr}) / 365 \text{ day/yr} =$	15,376,674		(order/day)
Daily demand per store		$(\text{hrs/day}) \times 60 \text{ min/hr} \times (\text{order/store-min}) =$	960		(order/store-day)
Est. number of U.S. stores		$(\text{order/day}) / (\text{order/store-day}) =$	16,017		(store)

# System Performance Estimation

- Often easy to estimate performance of a new system if can assume either perfect or no control
- Example: estimate waiting time for a bus
  - 8 min. avg. time (aka “headway”) between buses
  - Customers arrive at random
    - assuming no web-based bus tracking
  - Perfect control (LB): wait time = half of headway
  - No control (*practical* UB): wait time = headway
    - assuming buses arrive at random

$$\text{Estimated wait time} = \sqrt{LB \times UB} = \sqrt{\frac{8}{2} \times 8} = 5.67 \text{ min}$$

- Bad control can result in higher values than no control

# <http://www.nextbuzz.gatech.edu/>

 SELF-COORDINATING BUSES  
REDUCE BUNCHING

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HOME THE IDEA PROOF OF CONCEPT HOW IT WORKS CONTRIBUTORS

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## A BUS-HEADWAY CONTROLLER

A software system to coördinate buses on a route, based on an [idea](#) by [John J. Bartholdi III](#) and [Donald D. Eisenstein](#). The current version of the software was designed and largely written by Loren K. Platzman. Implementation has been led by [Russ Clark](#), Jin Lee, and David Williamson.

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### THE IDEA

Delaying buses briefly at certain checkpoints equalizes headways

[Read more](#)



### PROOF OF CONCEPT

Coordinating trolleys on Georgia Tech's busiest route

[Read more](#)



### HOW IT WORKS

Tablets, GPS, cellular networks, and web-based control

[Read more](#)

# Crowdsourcing

- Obtain otherwise hard to get information from a large group of online workers
- Amazon's Mechanical Turk is best known
  - Jobs posted as HITs (Human Information Tasks) that typically pay \$1-2 per hour
  - Main use has been in machine learning to create tagged data sets for training purposes
  - Has been used in logistics engineering to estimate the percentage homes in U.S. that have sidewalks (sidewalk deliveries by Starship robots)

# Available 2016: Starship Technologies

- Started by Skype co-founders
- 99% autonomous
- Goal: “deliver ‘two grocery bags’ worth of goods (weighing up to 20lbs) in 5-30 minutes for ‘10-15 times less than the cost of current last-mile delivery alternatives.”



# Matrix Multiplication

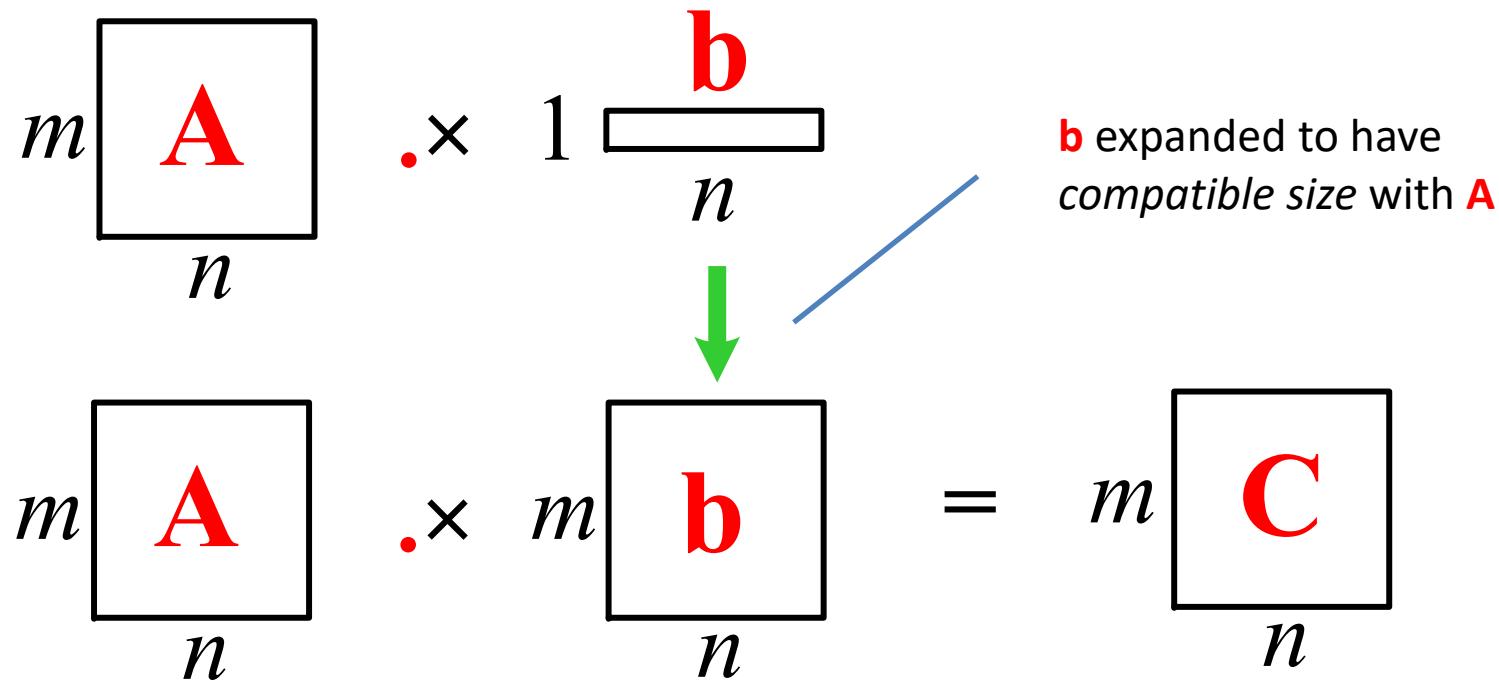
$$m \begin{array}{|c|} \hline \text{A} \\ \hline n \\ \hline \end{array} \times n \begin{array}{|c|} \hline \text{B} \\ \hline p \\ \hline \end{array} = m \begin{array}{|c|} \hline \text{C} \\ \hline p \\ \hline \end{array}$$

$$\cancel{(m \times n) \times (n \times p)} = (m \times p)$$



Arrays must have same  
*inner dimensions*

# Element-by-Element Multiplication



$$(m \times n) \times (1 \times n) = (m \times n)$$

# Compatible Sizes

- Two arrays have compatible sizes if, for *each respective dimension*, either
  - has the same size, or
  - size of one of arrays is one, in which case it is automatically duplicated so that it matches the size of the other array

$$\mathbf{A}_{m \times n} \bullet^* \mathbf{B}_{m \times n} = \mathbf{C}_{m \times n}$$

$$\mathbf{A}_{m \times n} \bullet^* \mathbf{b}_{1 \times n} = \mathbf{C}_{m \times n}$$

$$\mathbf{a}_{m \times 1} \bullet^* \mathbf{B}_{m \times n} = \mathbf{C}_{m \times n}$$

$$\mathbf{a}_{m \times 1} \bullet^* \mathbf{b}_{1 \times n} = \mathbf{C}_{m \times n}$$

$$\mathbf{a}_{1 \times n} \bullet^* \mathbf{b}_{m \times 1} = \mathbf{C}_{m \times n}$$

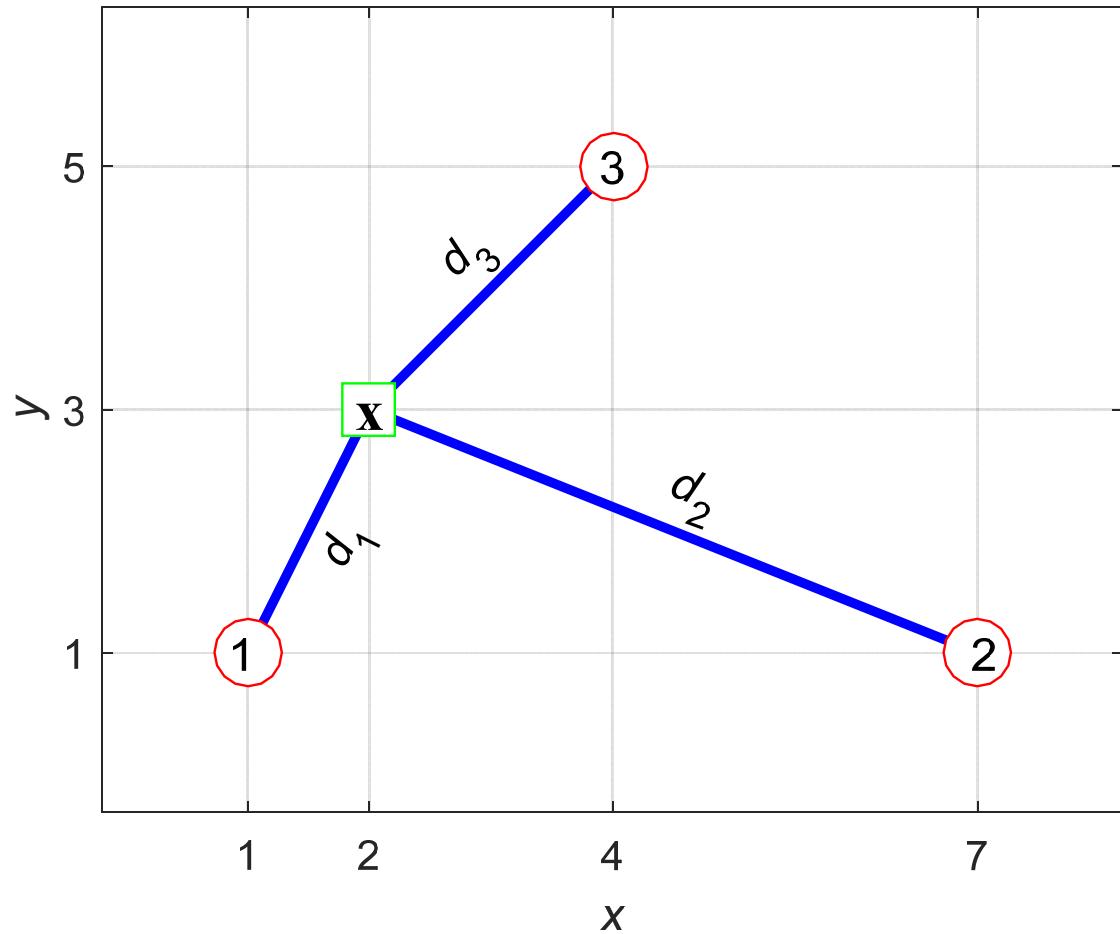
~~$$\mathbf{A}_{m \times n} \bullet^* \mathbf{B}_{m \times p}$$~~

~~$$\mathbf{A}_{m \times n} \bullet^* \mathbf{b}_{n \times 1}$$~~

~~$$\mathbf{a}_{m \times 1} \bullet^* \mathbf{B}_{p \times n}$$~~

~~$$\mathbf{a}_{m \times 1} \bullet^* \mathbf{b}_{n \times 1}$$~~

# 2-D Euclidean Distance



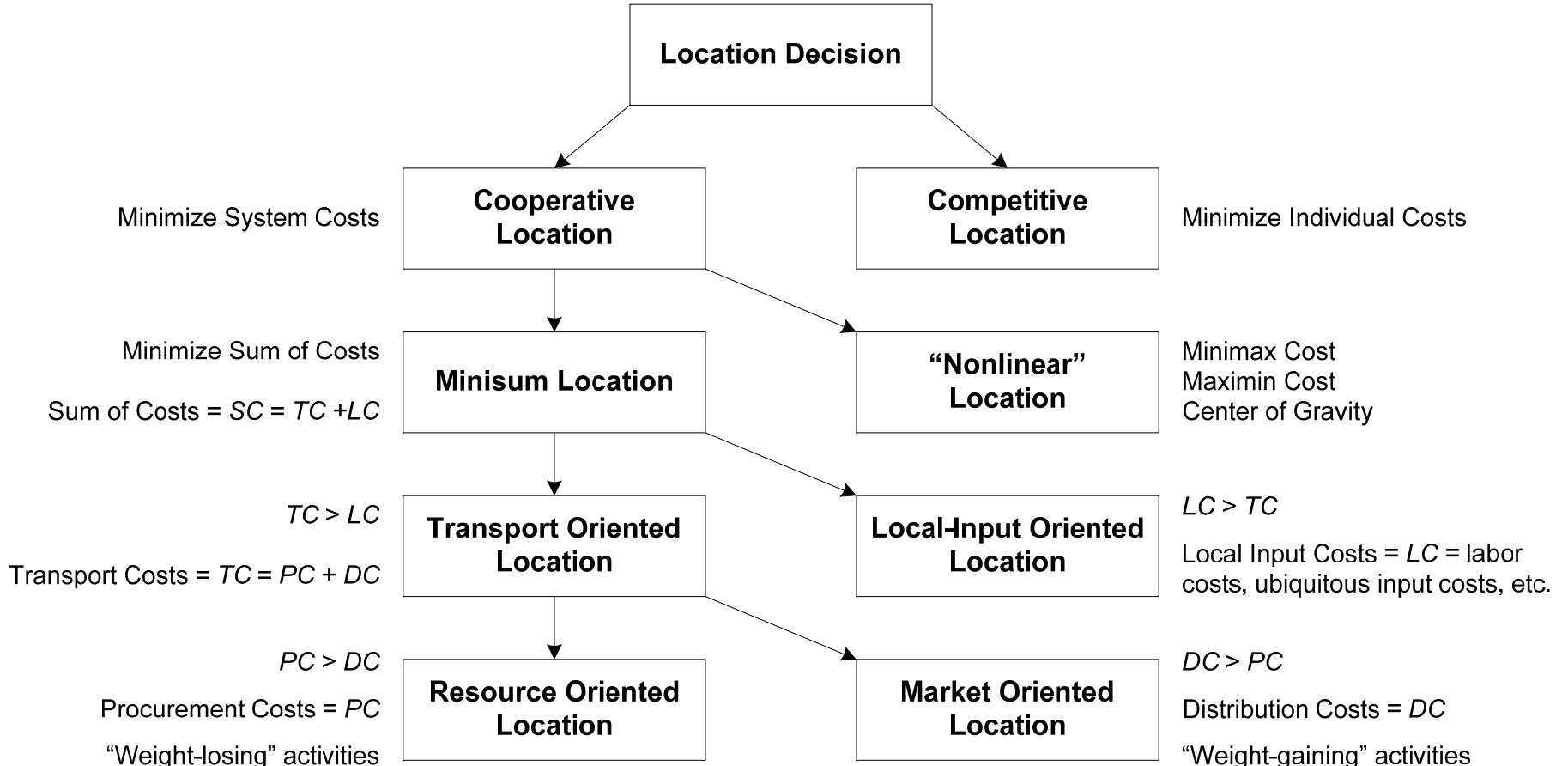
$$\mathbf{x} = \begin{bmatrix} 2 & 3 \end{bmatrix}, \quad \mathbf{P} = \begin{bmatrix} 1 & 1 \\ 7 & 1 \\ 4 & 5 \end{bmatrix}$$

$$\mathbf{d} = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} = \begin{bmatrix} \sqrt{(x_1 - p_{1,1})^2 + (x_2 - p_{1,2})^2} \\ \sqrt{(x_1 - p_{2,1})^2 + (x_2 - p_{2,2})^2} \\ \sqrt{(x_1 - p_{3,1})^2 + (x_2 - p_{3,2})^2} \end{bmatrix}$$

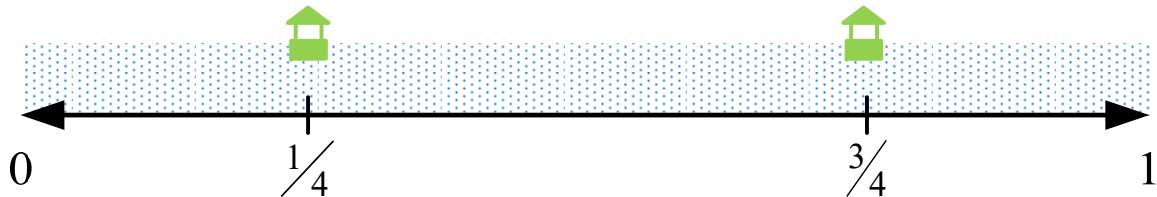
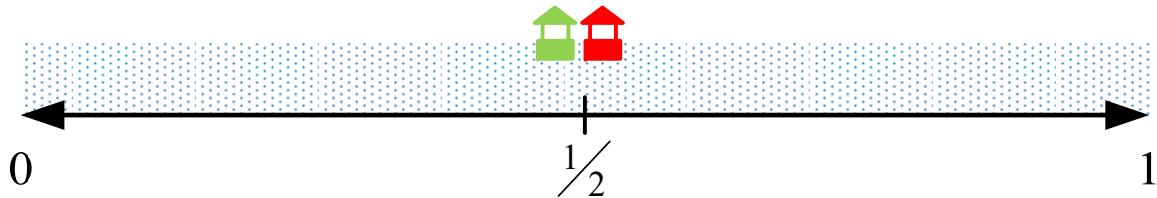
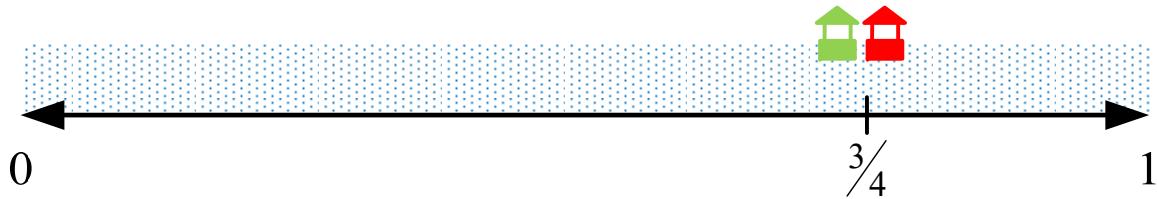
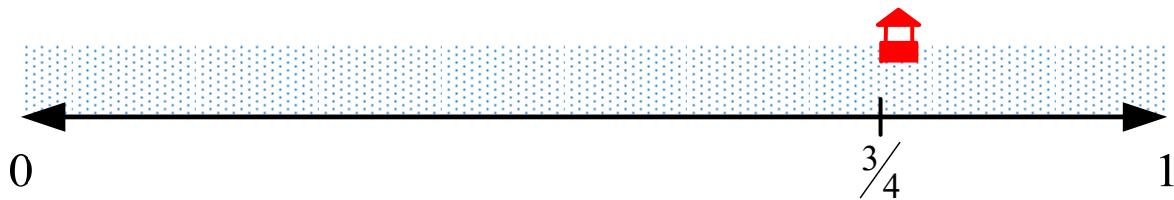
# Basic Matlab Workflow

- Given problem to solve:
  1. Test critical steps at Command Window
  2. Copy working critical steps to a cell (&&) in script file (myscript.m) along with supporting code
    - Repeat using new cells for additional problems
- Once all problems solved, report using:
  - `>> diary hw1soln.txt`
  - Evaluate each cell in script:
    - To see code + results: select text then Evaluate Selection on mouse menu (or F9)
    - To see results: position cursor in cell then Evaluate Current Section (Ctrl+Enter)
  - `>> diary off`
- Can also report using Publish (see Matlab menu) as html or Word
- Submit all files created, which may include additional
  - Data files (myscript.mat) or spreadsheet files (myexcel.xlsx)
  - Function files (myfun.m) that can allow use to re-use same code used in multiple problems
    - All code inside function isolated from other code except for inputs/outputs:  
`[out1,out2] = myfun(input1,input2)`

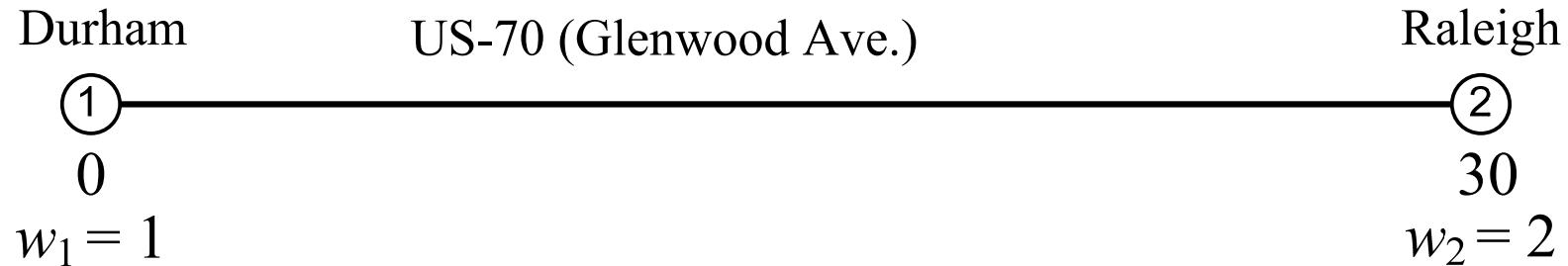
# Taxonomy of Location Problems



# Hotelling's Law



# 1-D Cooperative Location



$$\text{Min } TC = \sum w_i d_i$$

$$a_1 = 0, \quad a_2 = 30$$

$$\text{Min } TC = \sum w_i d_i^2$$

$$TC = \sum w_i d_i^2 = \sum w_i (x - a_i)^2$$

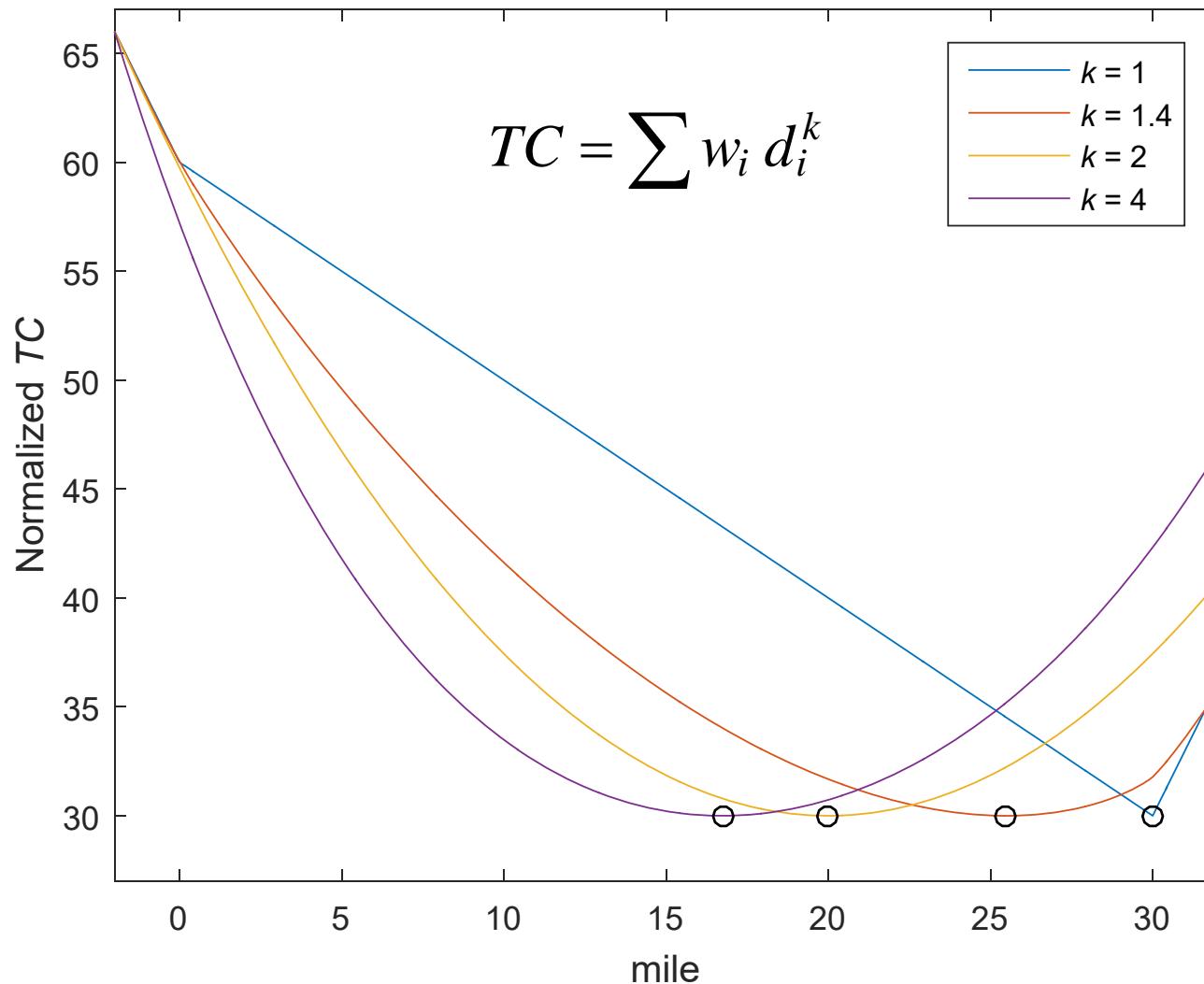
$$\text{Min } TC = \sum w_i d_i^k$$

$$\frac{dTC}{dx} = 2 \sum w_i (x - a_i) = 0 \Rightarrow$$

$$x \sum w_i = \sum w_i a_i \Rightarrow$$

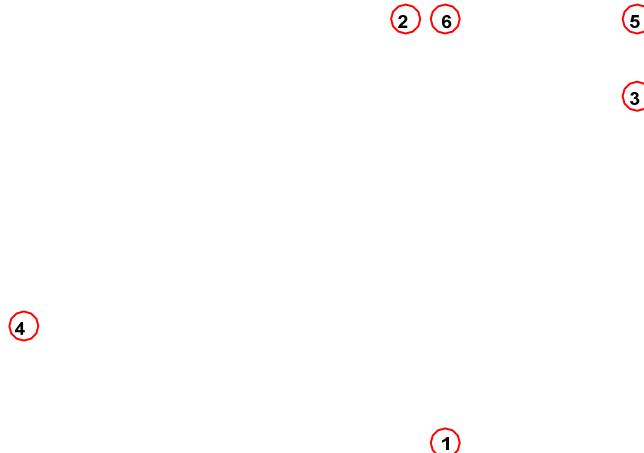
Squared-Euclidean Distance  $\Rightarrow$  Center of Gravity:  $x^* = \frac{\sum w_i a_i}{\sum w_i} = \frac{1(0) + 2(30)}{1+2} = 20$

# “Nonlinear” Location

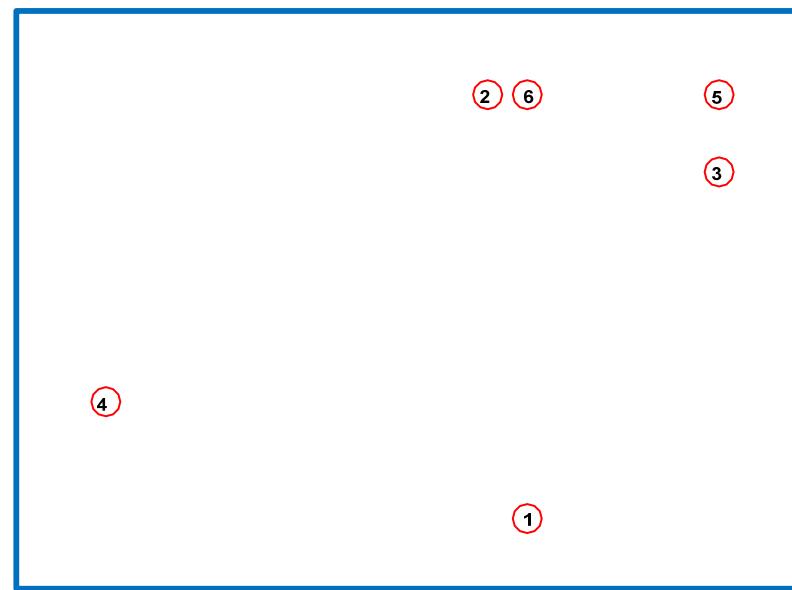


# Minimax and Maximin Location

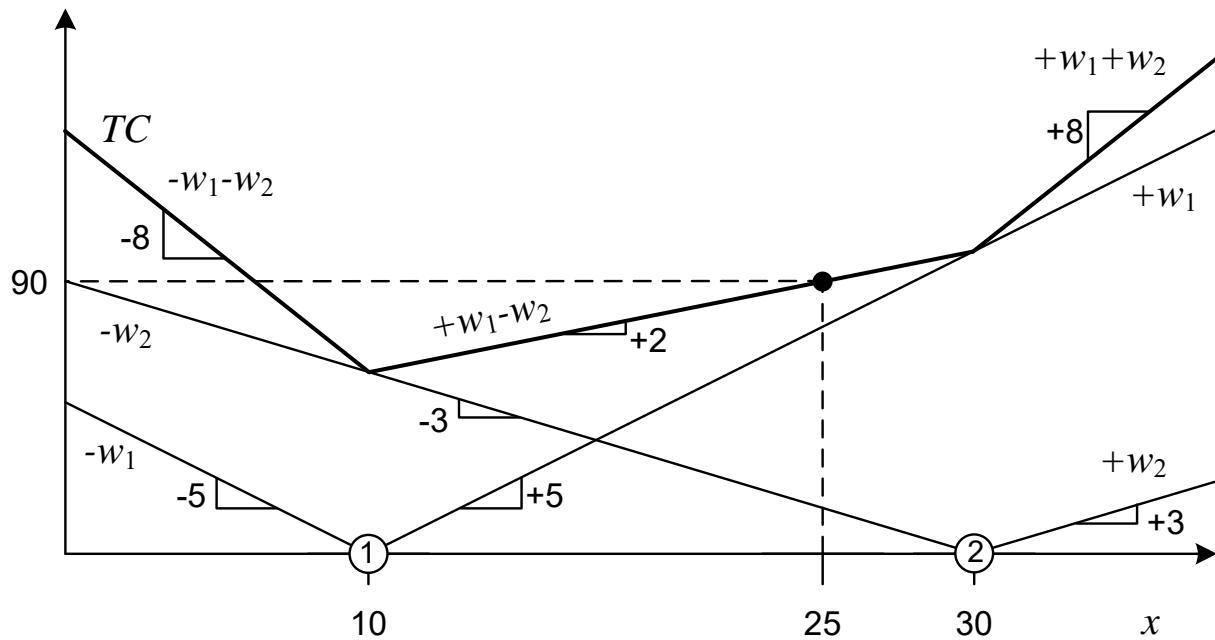
- Minimax
  - Min max distance
  - Set covering problem



- Maximin
  - Max min distance
  - AKA obnoxious facility location



# 2-EF Minisum Location



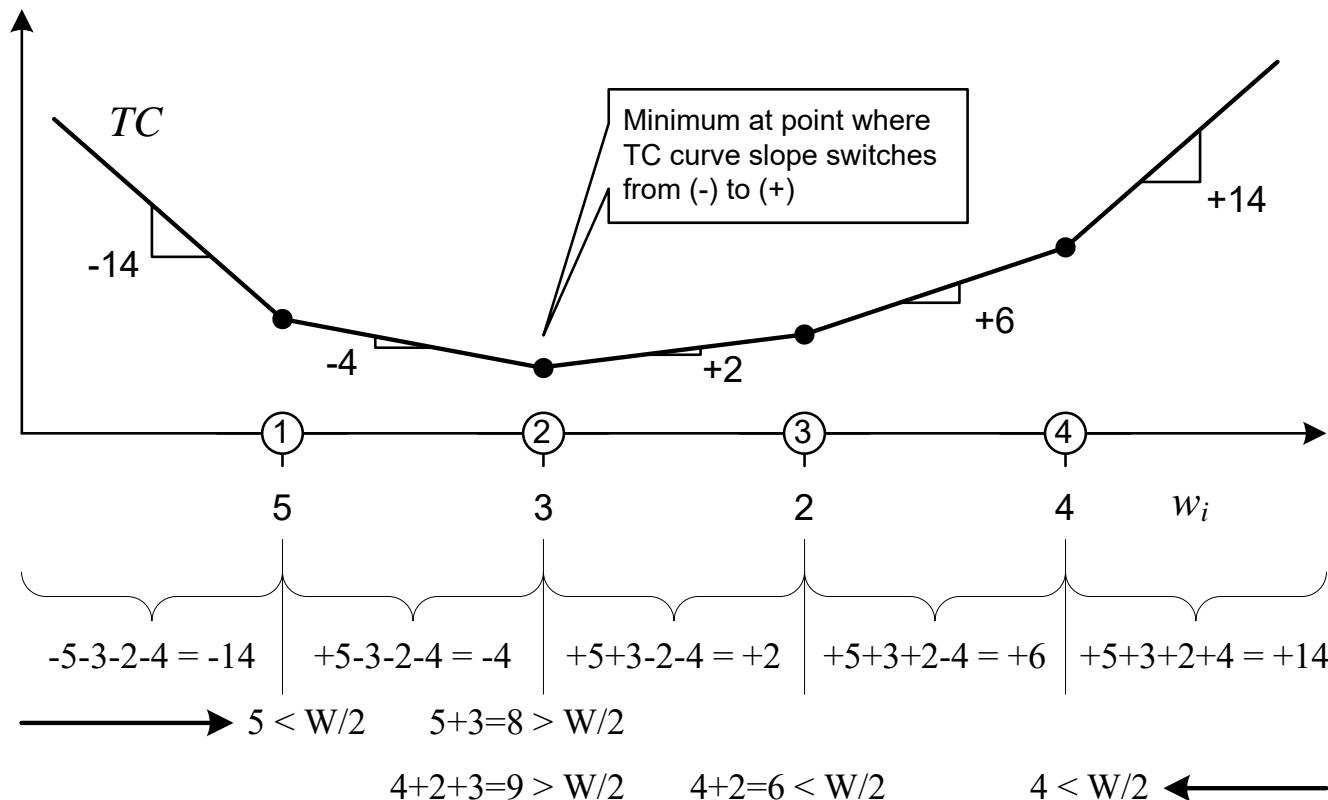
$$TC(x) = \sum w_i d_i = \beta_1(x - x_1) + \beta_2(x - x_2), \quad \text{where } \beta_i = \begin{cases} w_i, & \text{if } x \geq x_i \\ -w_i, & \text{if } x < x_i \end{cases}$$

$$\begin{aligned} TC(25) &= w_1(25 - 10) + (-w_2)(25 - 30) \\ &= 5(15) + (-3)(-5) = 90 \end{aligned}$$

# Median Location: 1-D 4 EFs

*Median location:* For each dimension  $x$  of  $X$ :

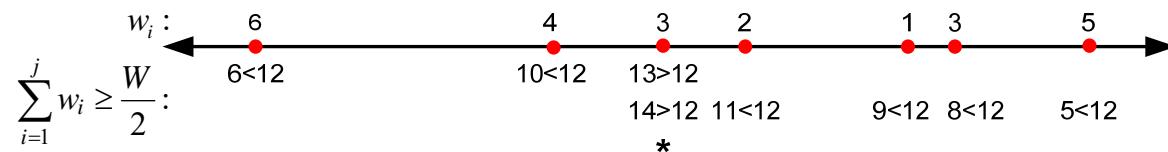
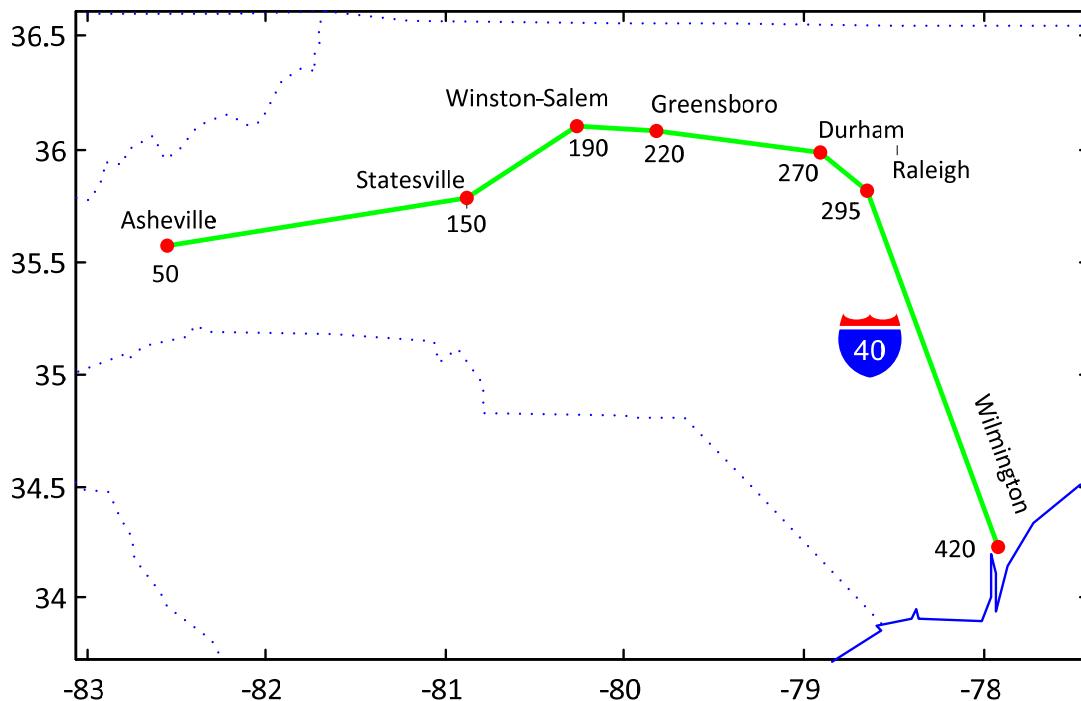
1. Order EFs so that  $|x_1| \leq |x_2| \leq \dots \leq |x_m|$
2. Locate  $x$ -dimension of NF at the first EF $j$  where  $\sum_{i=1}^j w_i \geq \frac{W}{2}$ , where  $W = \sum_{i=1}^m w_i$



# Median Location: 1-D 7 EFs

*Median location:* For each dimension  $x$  of  $X$ :

1. Order EFs so that  $|x_1| \leq |x_2| \leq \dots \leq |x_m|$
2. Locate  $x$ -dimension of NF at the first  $\text{EF}_j$  where  $\sum_{i=1}^j w_i \geq \frac{W}{2}$ , where  $W = \sum_{i=1}^m w_i$

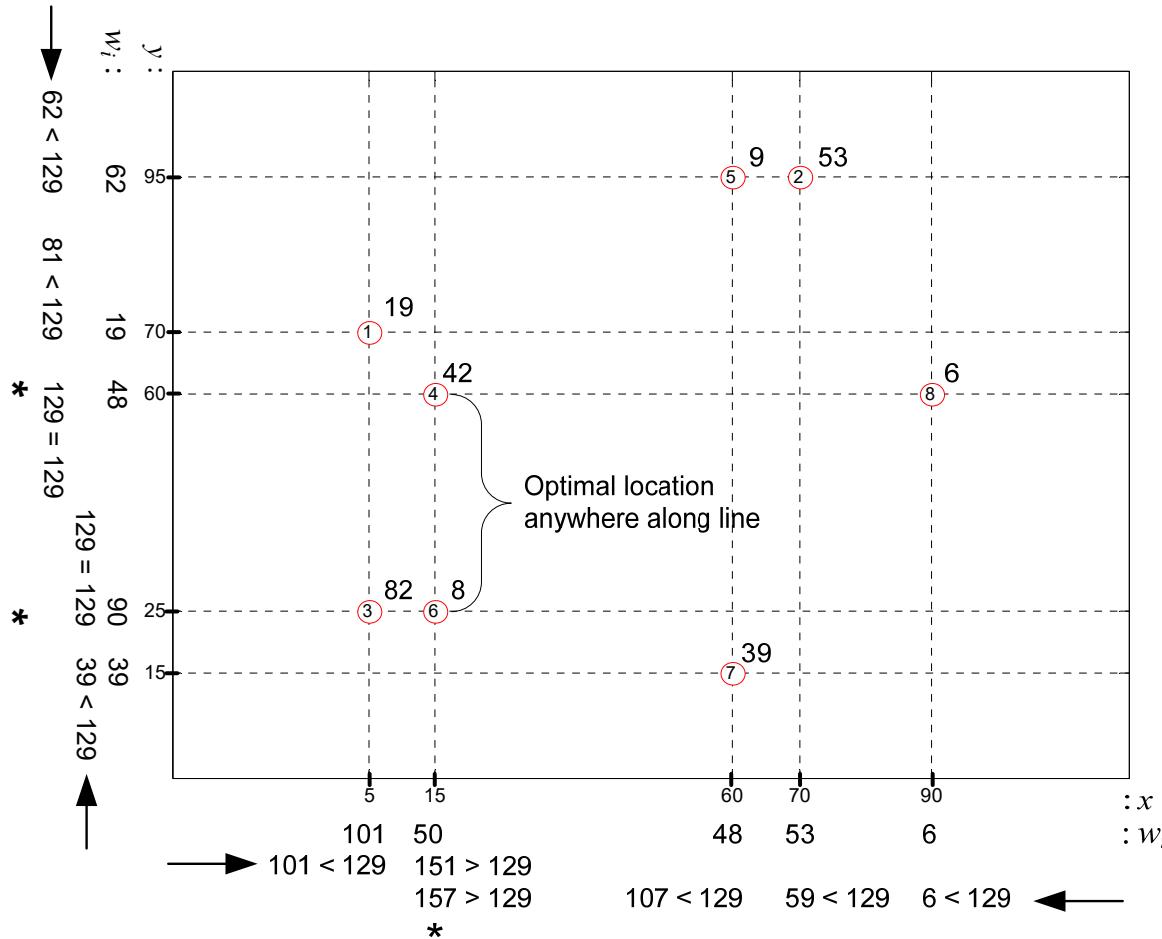


# Median Location: 2-D Rectilinear Distance 8 EFs

*Median location:* For each dimension  $x$  of  $X$ :

1. Order EFs so that  $|x_1| \leq |x_2| \leq \dots \leq |x_m|$

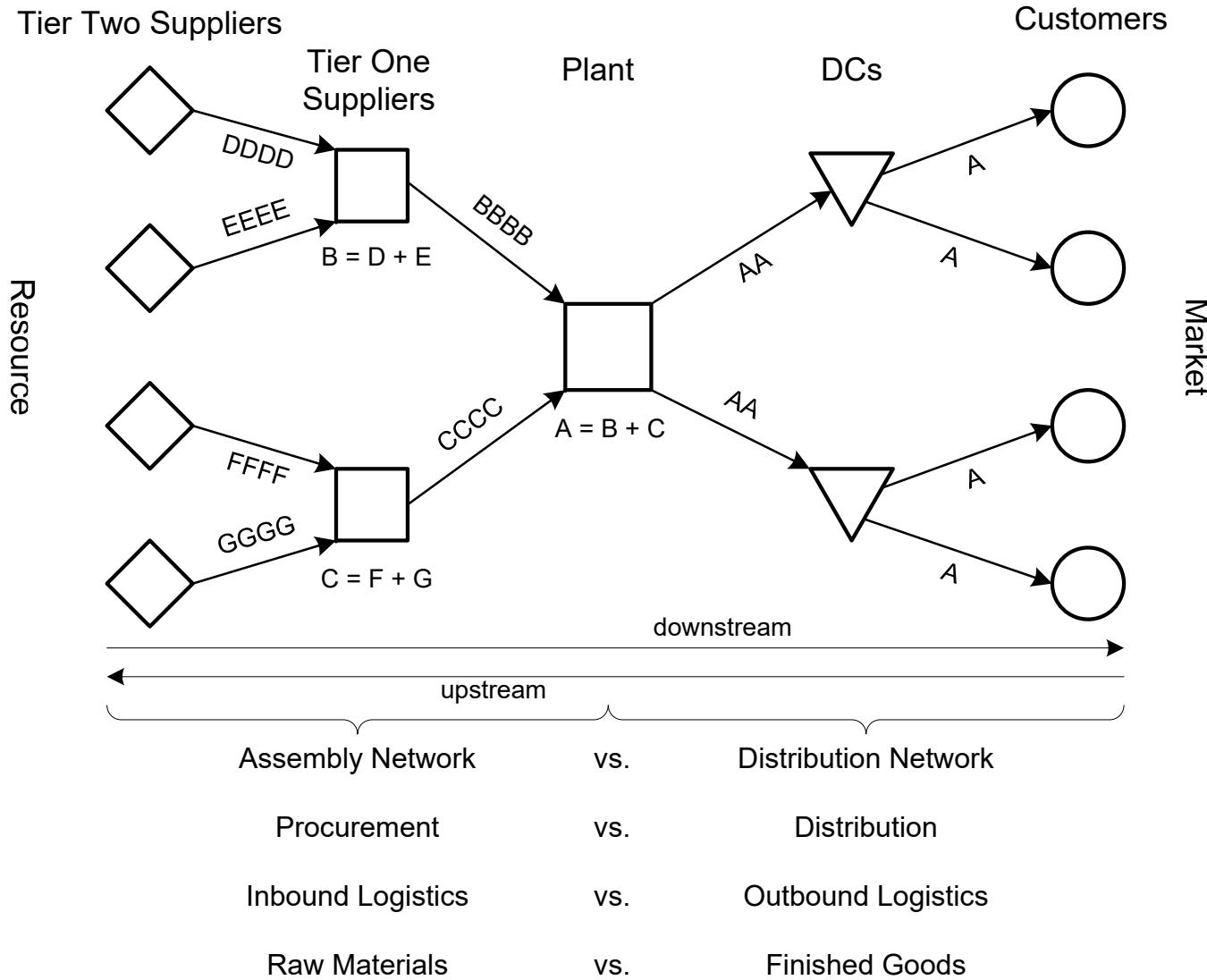
2. Locate  $x$ -dimension of NF at the first  $\text{EF}_j$  where  $\sum_{i=1}^j w_i \geq \frac{W}{2}$ , where  $W = \sum_{i=1}^m w_i$



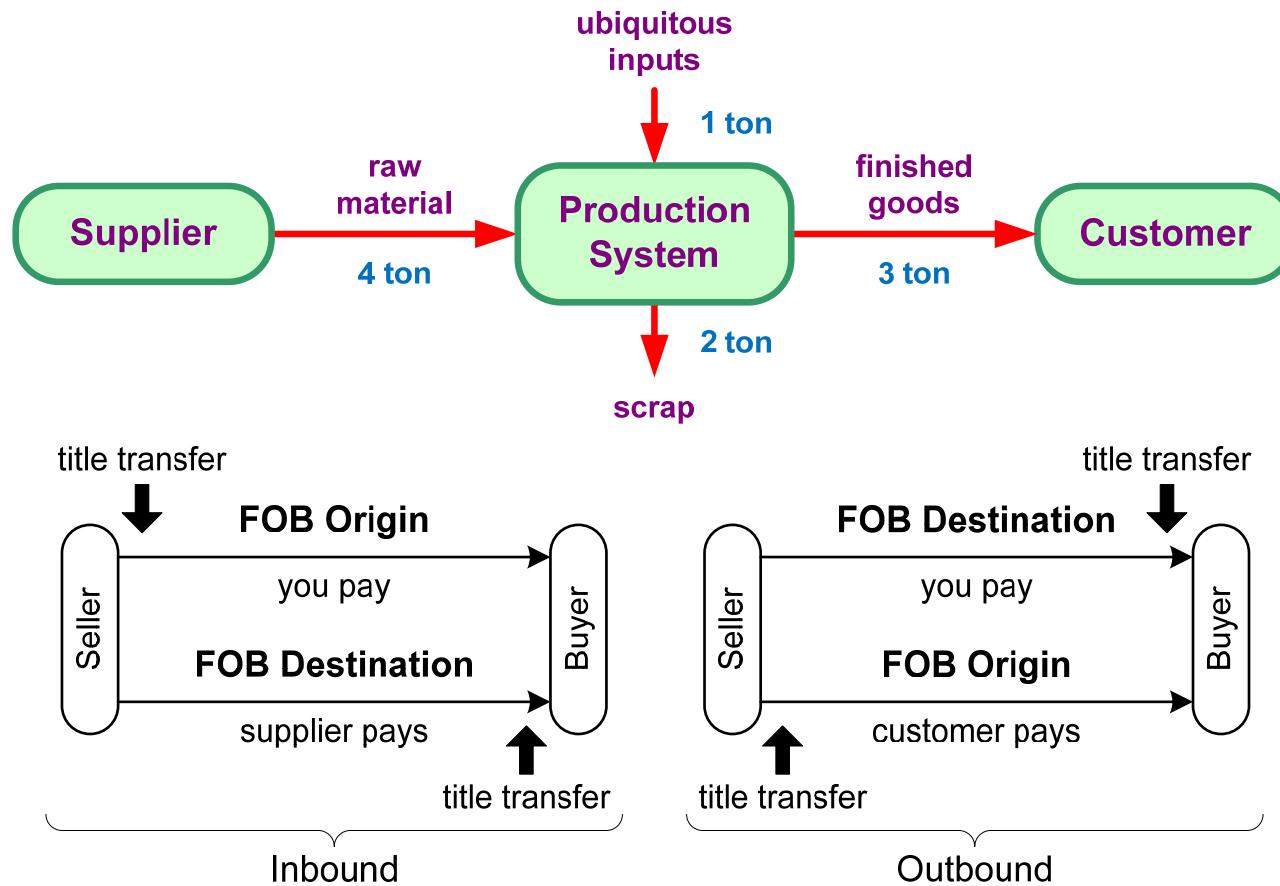
$$d_1(P_1, P_2) = |x_1 - x_2| + |y_1 - y_2|$$

$$d_2(P_1, P_2) = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$$

# Logistics Network for a Plant



# Basic Production System



**FOB (free on board)**

# Monetary vs. Physical Weight

$$\min TC(X) = \sum_{i=1}^m w_i d(X, P_i) = \sum_{i=1}^m \underbrace{f_i r_i}_{w_i} d(X, P_i)$$

where  $TC$  = total transport cost (\$/yr)

$w_i$  = monetary weight (\$/mi-yr)

$f_i$  = physical weight rate (ton/yr)

$r_i$  = transport rate (\$/ton-mi)

$d(X, P_i)$  = distance between NF at  $X$  and EF $_i$  at  $P_i$  (mi)

NF = new facility to be located

EF = existing facility

$m$  = number of EFs

(Monetary) Weight Gaining:  $\Sigma w_{\text{in}} < \Sigma w_{\text{out}}$

Physically Weight Losing:  $\Sigma f_{\text{in}} > \Sigma f_{\text{out}}$

# Minisum Location: TC vs. TD

- Assuming local input costs are
  - same at every location or
  - insignificant as compared to transport costs,the minisum transport-oriented single-facility location problem is to locate NF to minimize TC
- Can minimize total distance (TD) if transport rate is same:

$$\min TD(X) = \sum_{i=1}^m w_i d(X, P_i) = \sum_{i=1}^m \underbrace{f_i r_i}_{w_i} d(X, P_i)$$

where       $TD$  = total transport distance (mi/yr)

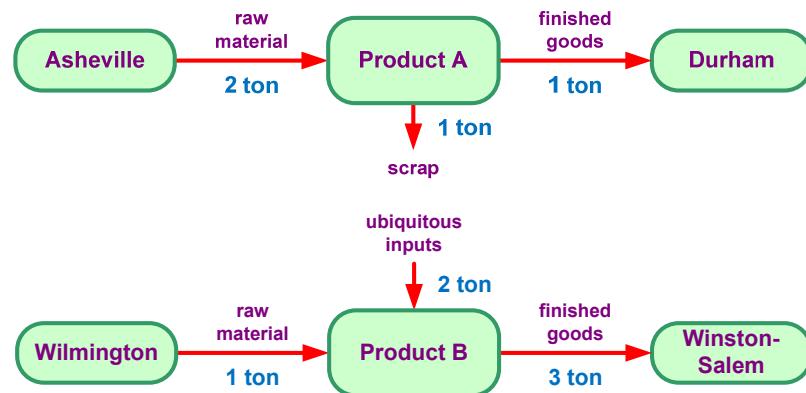
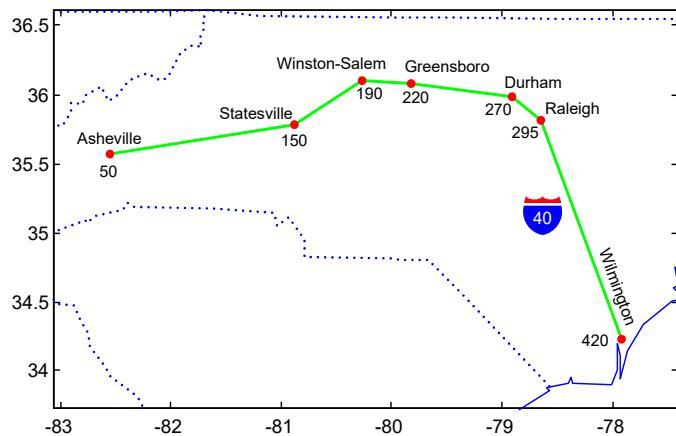
$w_i$  = monetary weight (trip/yr)

$f_i$  = trips per year (trip/yr)

$r_i$  = transport rate = 1

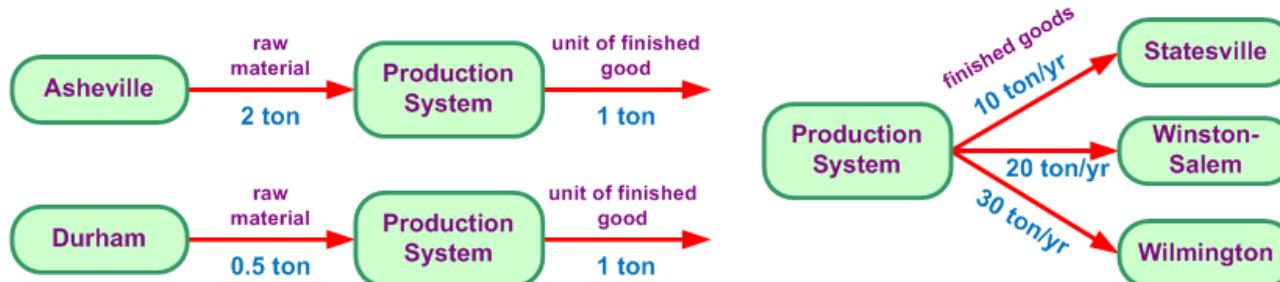
$d(X, P_i)$  = per-trip distance between NF and EF<sub>i</sub> (mi/trip)

# Example: Single Supplier/Customer



- The cost per ton-mile (i.e., the cost to ship one ton, one mile) for both raw materials and finished goods is \$0.10.
  - Where should the plant for each product be located so that procurement and distribution costs (i.e., transportation costs to and from the plant) are minimized?
  - How would the location decision change if the customers paid for distribution costs (FOB Origin) instead of the producer (FOB Destination)?
    - In particular, what would be the impact if there were competitors located along I-40 producing the same product?

# 1-D Location with Procurement and Distribution Costs



Assume: all scrap is disposed of locally

$$r_{in} = \$0.33/\text{ton-mi}$$

$$f_4 = BOM_4 \sum_{i=1}^3 f_i = 2(60) = 120, \quad w_4 = f_4 r_{in} = 40$$

$$f_5 = BOM_5 \sum_{i=1}^3 f_i = 0.5(60) = 30, \quad w_5 = f_5 r_{in} = 10$$

$$TC = \sum_{i=1}^j w_i \times d_i$$

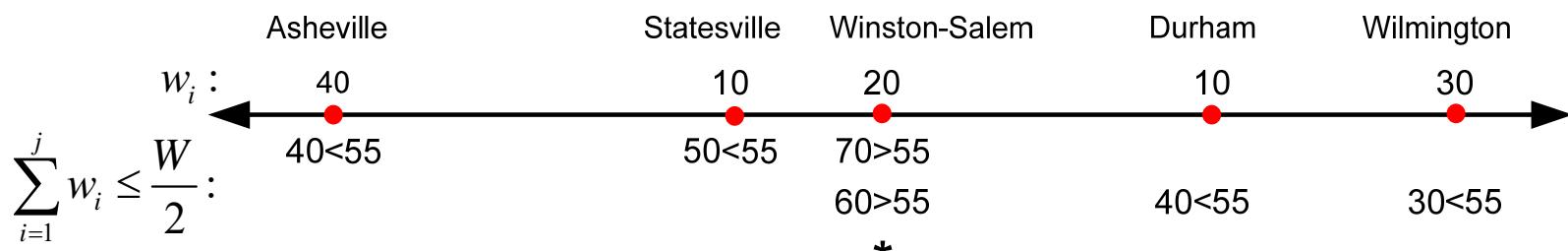
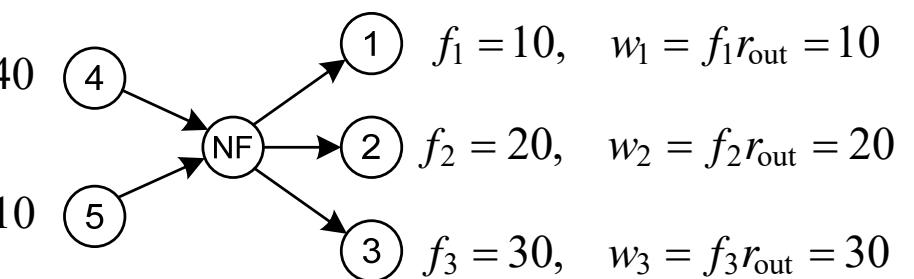
$$\underbrace{w_i}_{\substack{\text{monetary} \\ (\$/\text{mi-yr})}} = \underbrace{f_i}_{\substack{\text{physical} \\ (\text{ton/yr})}} \times \underbrace{r_i}_{(\$/\text{ton-mi})}$$

$$r_{out} = \$1.00/\text{ton-mi}$$

$$f_1 = 10, \quad w_1 = f_1 r_{out} = 10$$

$$f_2 = 20, \quad w_2 = f_2 r_{out} = 20$$

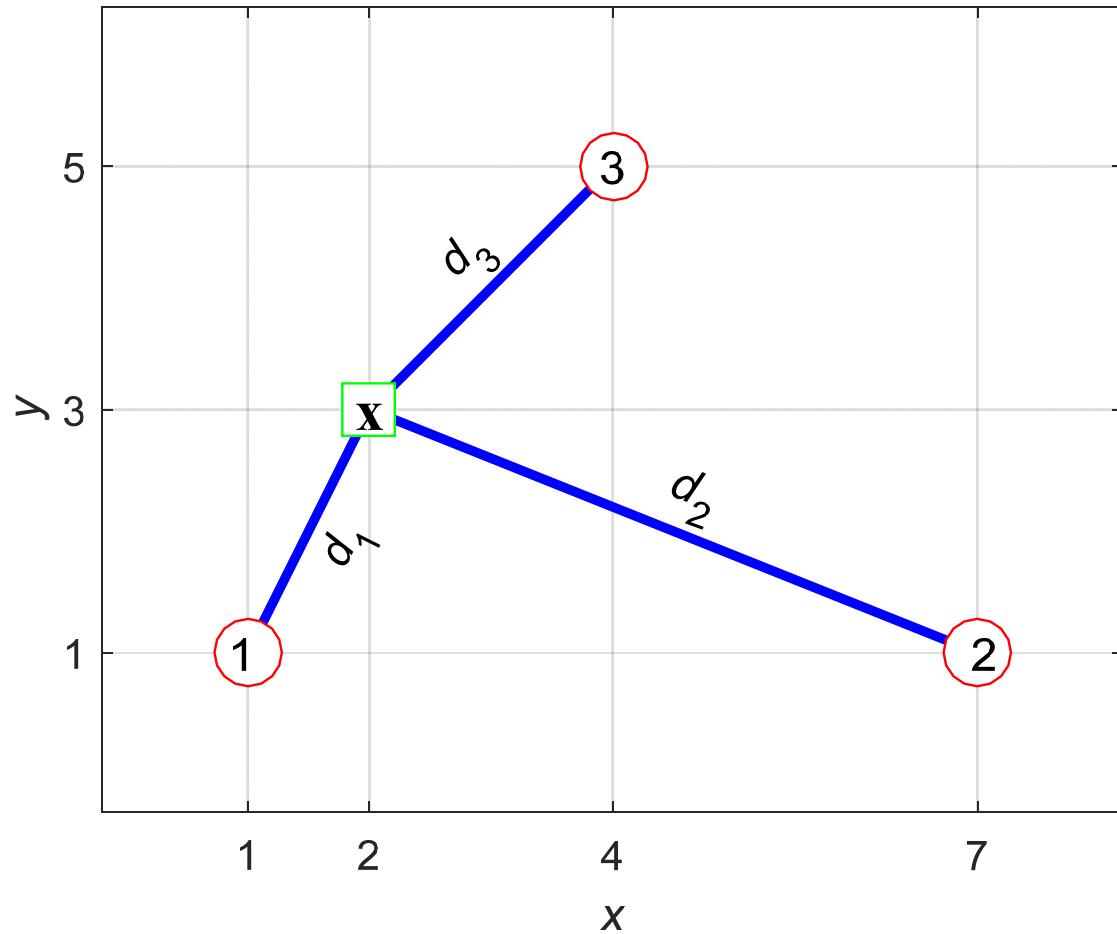
$$f_3 = 30, \quad w_3 = f_3 r_{out} = 30$$



(Monetary) Weight Gaining:  $\Sigma w_{in} = 50 < \Sigma w_{out} = 60$

Physically Weight Losing:  $\Sigma f_{in} = 150 > \Sigma f_{out} = 60$

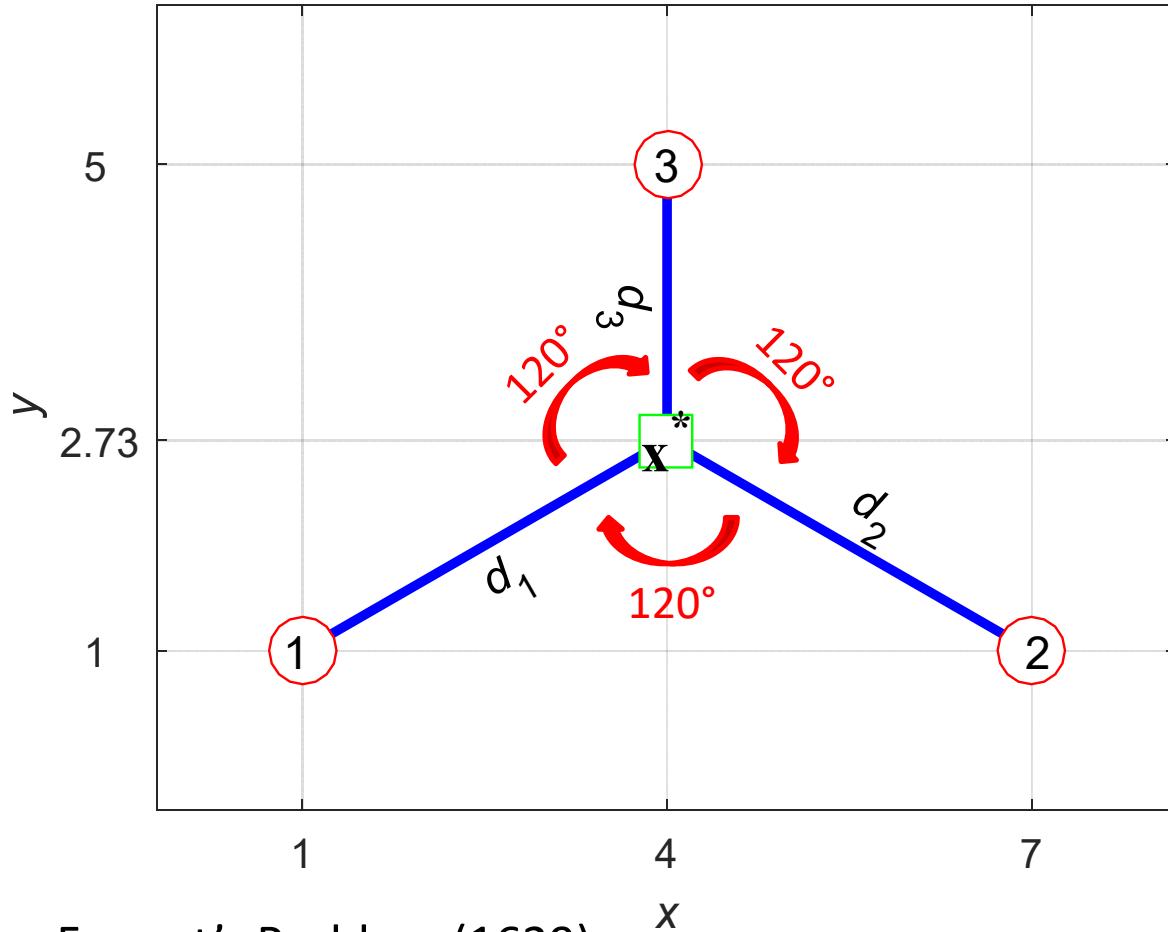
# 2-D Euclidean Distance



$$\mathbf{x} = \begin{bmatrix} 2 & 3 \end{bmatrix}, \quad \mathbf{P} = \begin{bmatrix} 1 & 1 \\ 7 & 1 \\ 4 & 5 \end{bmatrix}$$

$$\mathbf{d} = \begin{bmatrix} d_1 \\ d_2 \\ d_3 \end{bmatrix} = \begin{bmatrix} \sqrt{(x_1 - p_{1,1})^2 + (x_2 - p_{1,2})^2} \\ \sqrt{(x_1 - p_{2,1})^2 + (x_2 - p_{2,2})^2} \\ \sqrt{(x_1 - p_{3,1})^2 + (x_2 - p_{3,2})^2} \end{bmatrix}$$

# Minisum Distance Location



$$\mathbf{P} = \begin{bmatrix} 1 & 1 \\ 7 & 1 \\ 4 & 5 \end{bmatrix}$$

$$d_i(\mathbf{x}) = \sqrt{(x_1 - p_{i,1})^2 + (x_2 - p_{i,2})^2}$$

$$TD(\mathbf{x}) = \sum_{i=1}^3 d_i(\mathbf{x})$$

$$\mathbf{x}^* = \arg \min_{\mathbf{x}} TD(\mathbf{x})$$

$$TD^* = TD(\mathbf{x}^*)$$

Fermat's Problem (1629):

Given three points, find fourth (Steiner point) such that sum to others is minimized

(Solution: Optimal location corresponds to all angles = 120°)

# Minisum Weighted-Distance Location

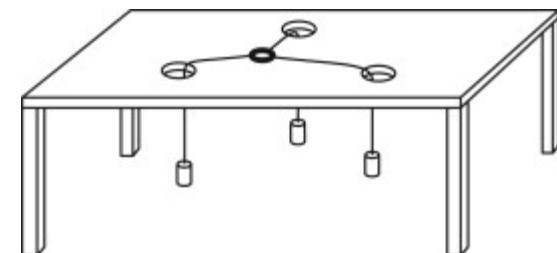
- Solution for 2-D+ and non-rectangular distances:
  - *Majority Theorem*: Locate NF at EF $j$  if  $w_j \geq \frac{W}{2}$ , where  $W = \sum_{i=1}^m w_i$
  - Mechanical (Varigon frame)
  - 2-D rectangular approximation
  - Numerical: nonlinear unconstrained optimization
    - Analytical/estimated derivative (quasi-Newton, fminunc)
    - Direct, derivative-free (Nelder-Mead, fminsearch)

$m$  = number of EFs

$$TC(\mathbf{x}) = \sum_{i=1}^m w_i d_i(\mathbf{x})$$

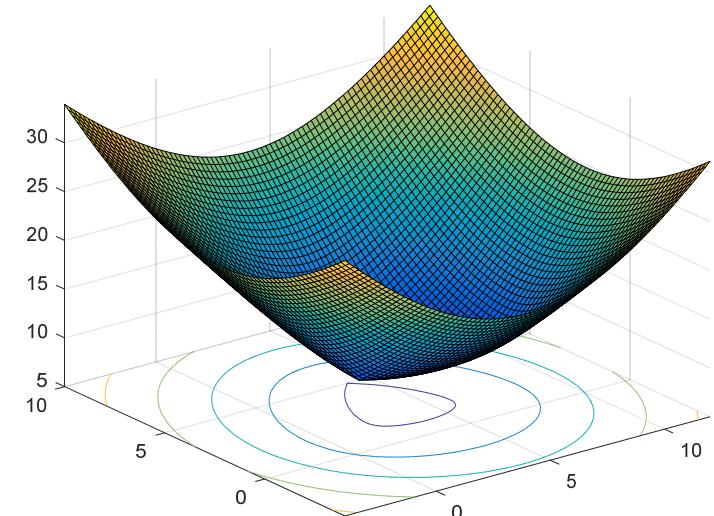
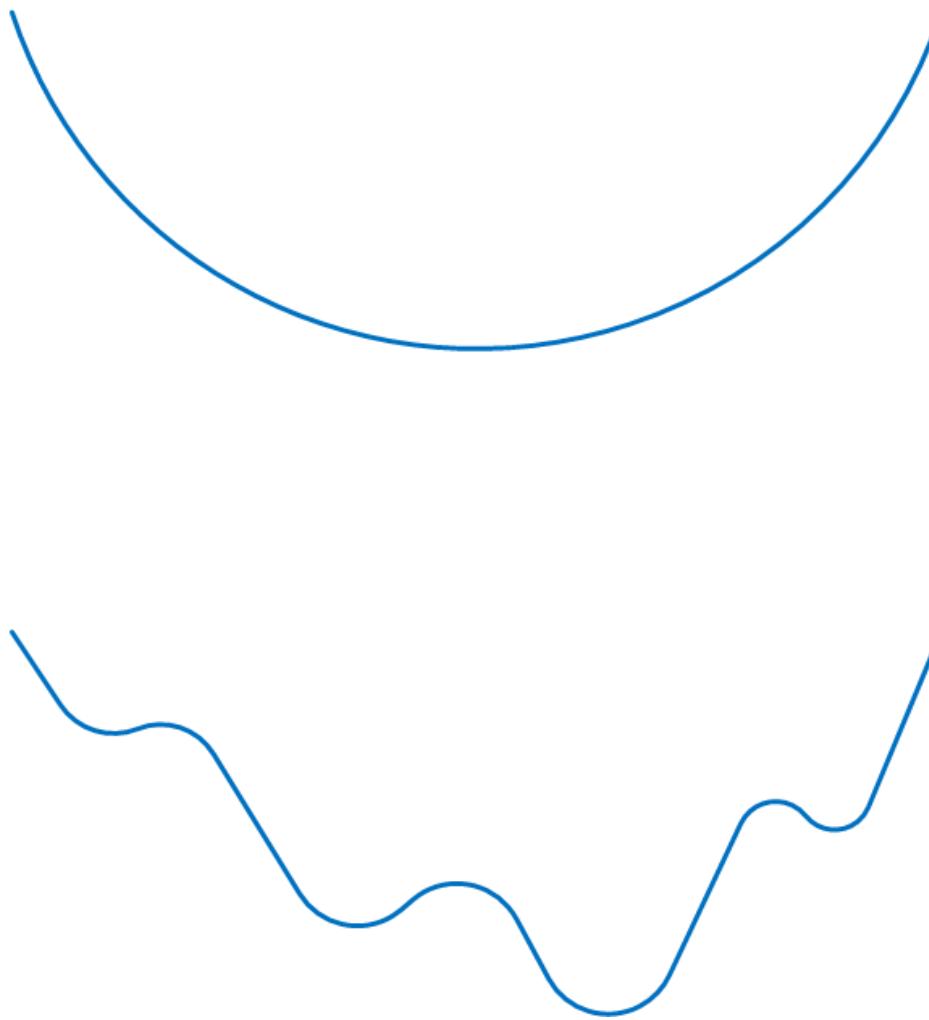
$$\mathbf{x}^* = \arg \min_{\mathbf{x}} TC(\mathbf{x})$$

$$TC^* = TC(\mathbf{x}^*)$$

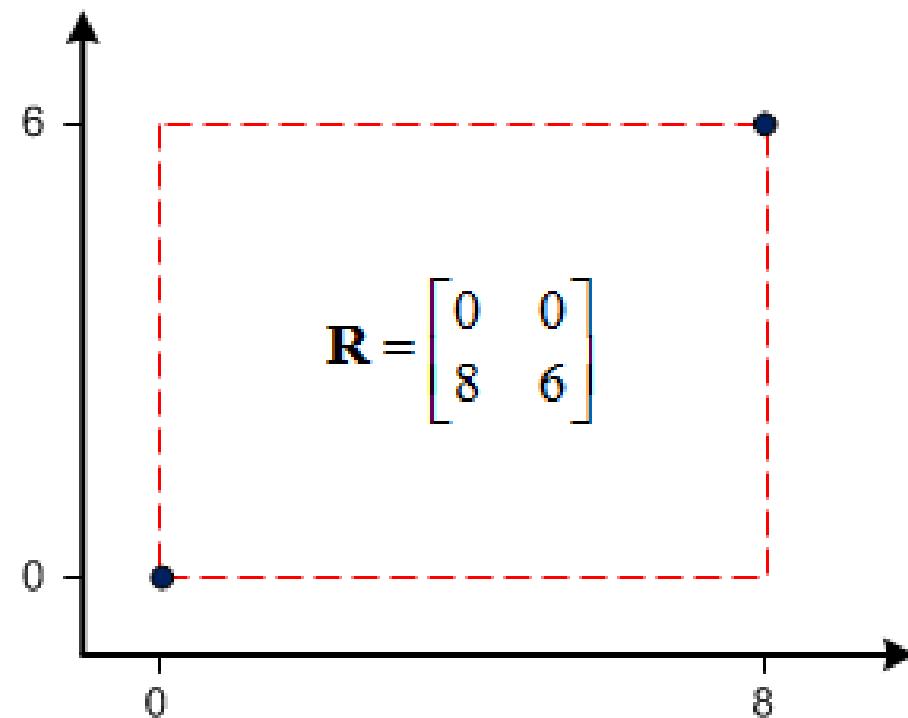


Varignon Frame

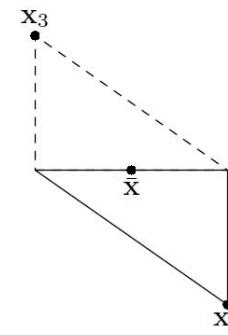
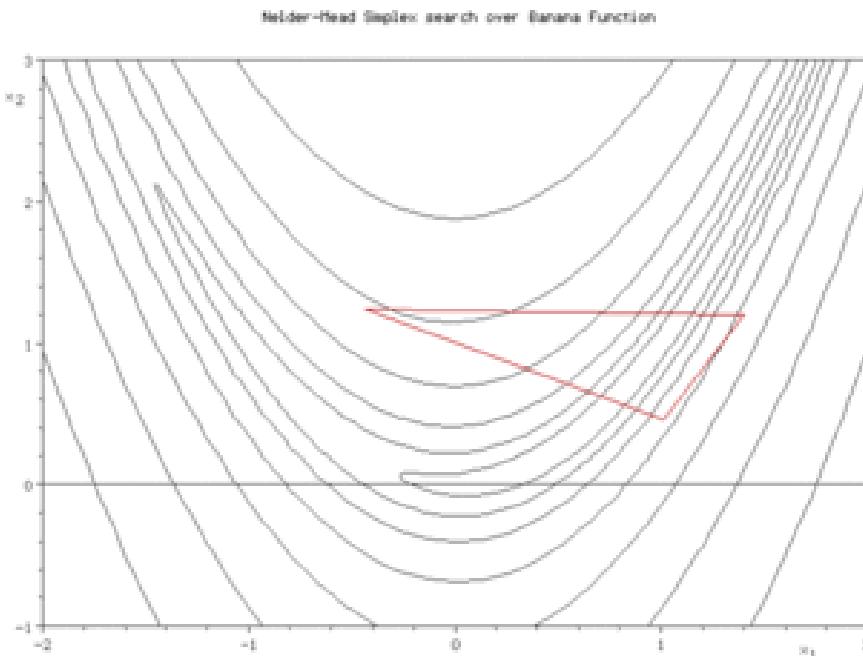
# Convex vs Nonconvex Optimization



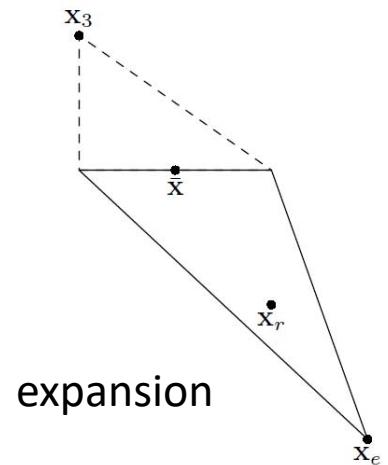
# Feasible Region



# Nelder-Mead Simplex Method

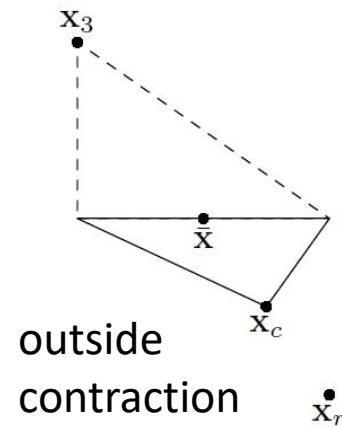


reflection

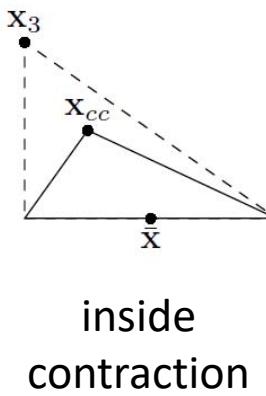


expansion

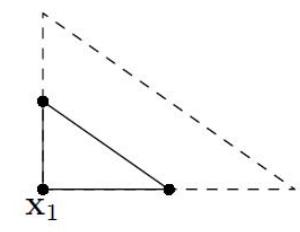
- AKA amoeba method
- Simplex is triangle in 2-D (dashed line in figures)



outside  
contraction



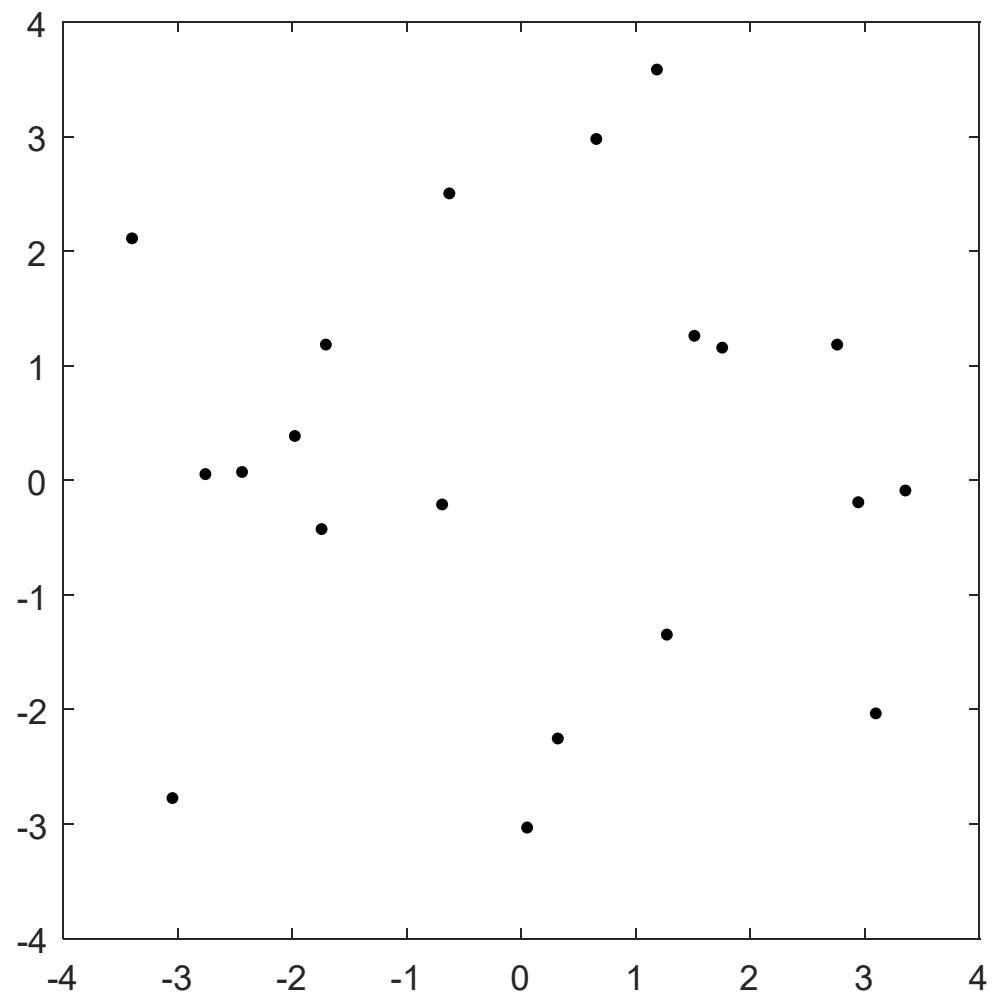
inside  
contraction



a shrink

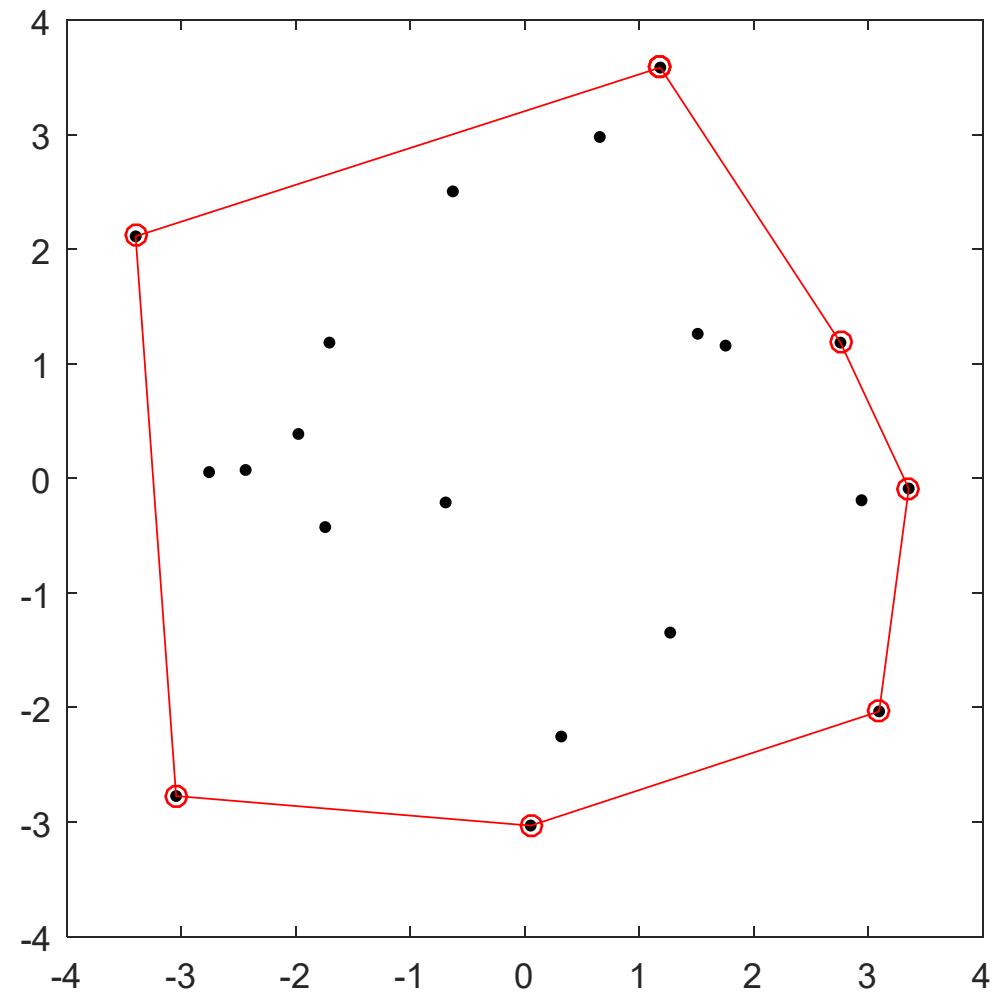
# Computational Geometry

- Design and analysis of algorithms for solving geometric problems
  - Modern study started with Michael Shamos in 1975
- Facility location:
  - geometric data structures used to “simplify” solution procedures



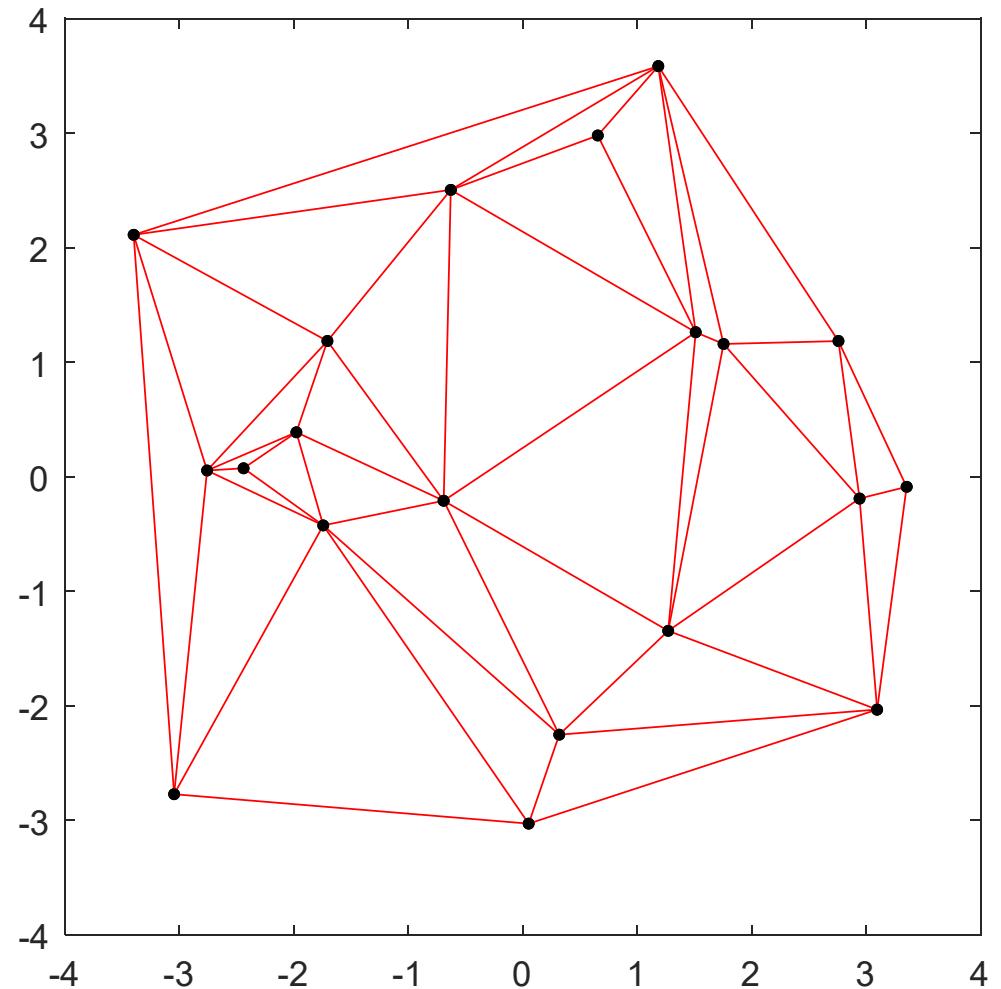
# Convex Hull

- Find the points that enclose all points
  - Most important data structure
  - Calculated, via Graham's scan in  $O(n \log n)$ ,  $n$  points



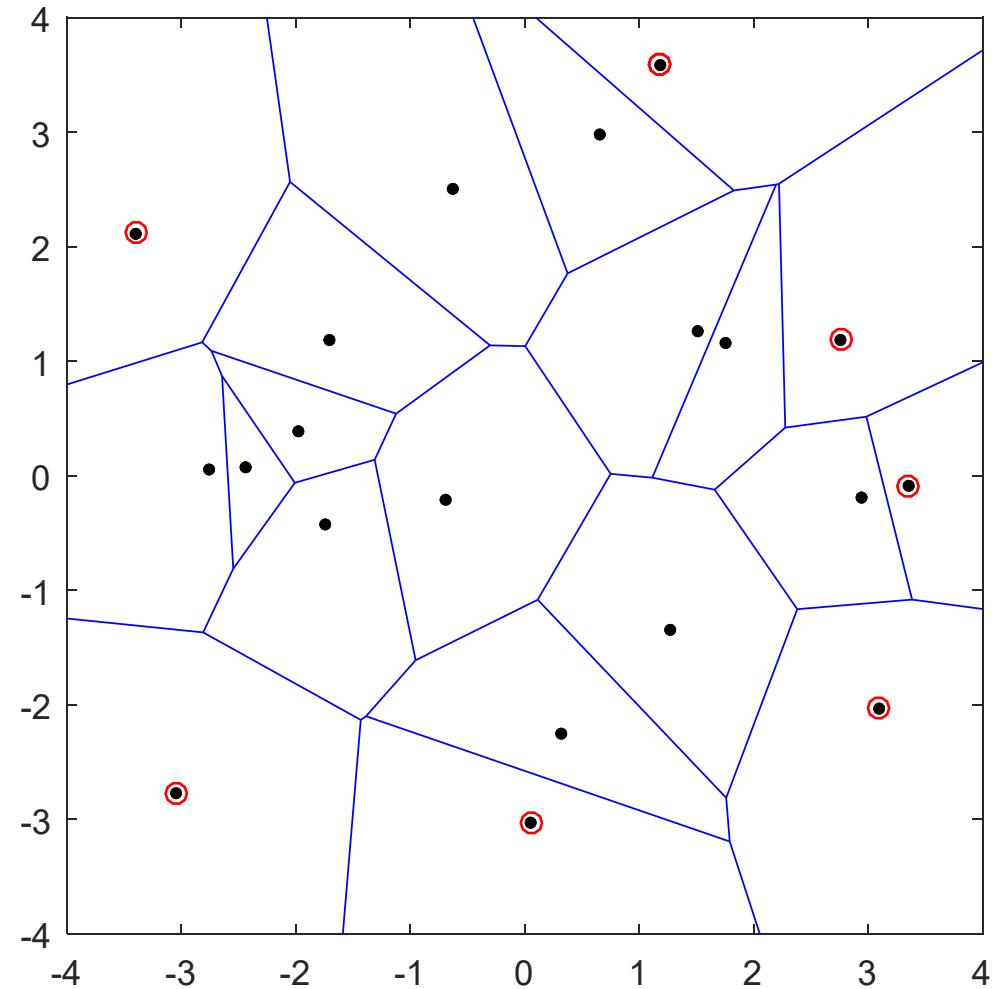
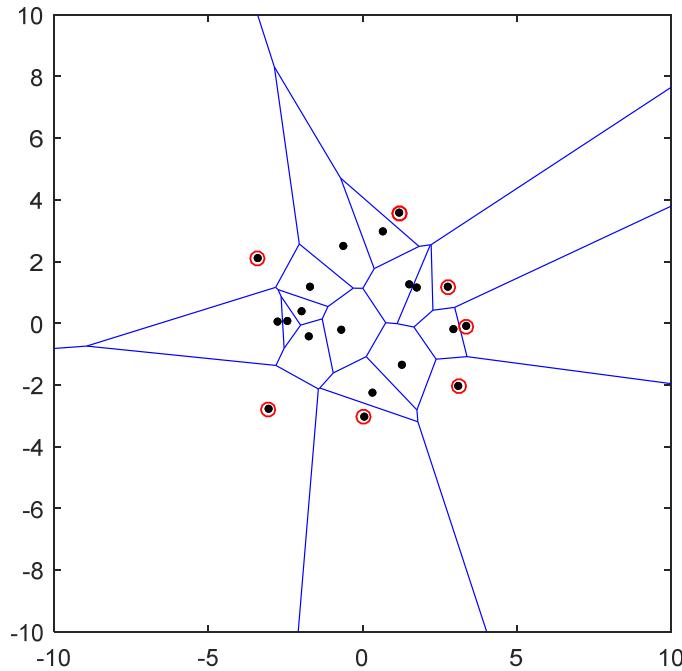
# Delaunay Triangulation

- Find the triangulation of points that maximizes the minimum angle of any triangle
  - Captures proximity relationships
  - Used in 3-D animation
  - Calculated, via divide and conquer, in  $O(n \log n)$ ,  $n$  points



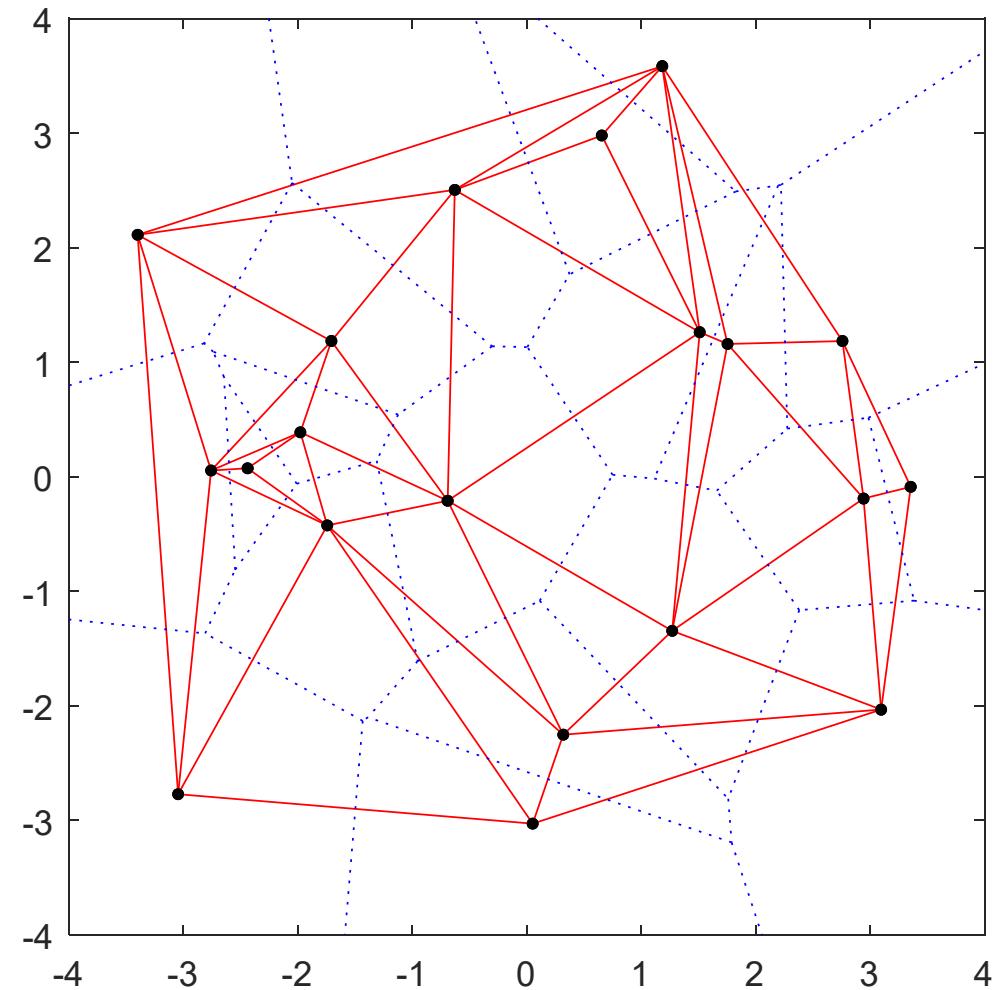
# Voronoi Diagram

- Each region defines area closest to a point
  - Open face regions indicate points in convex hull



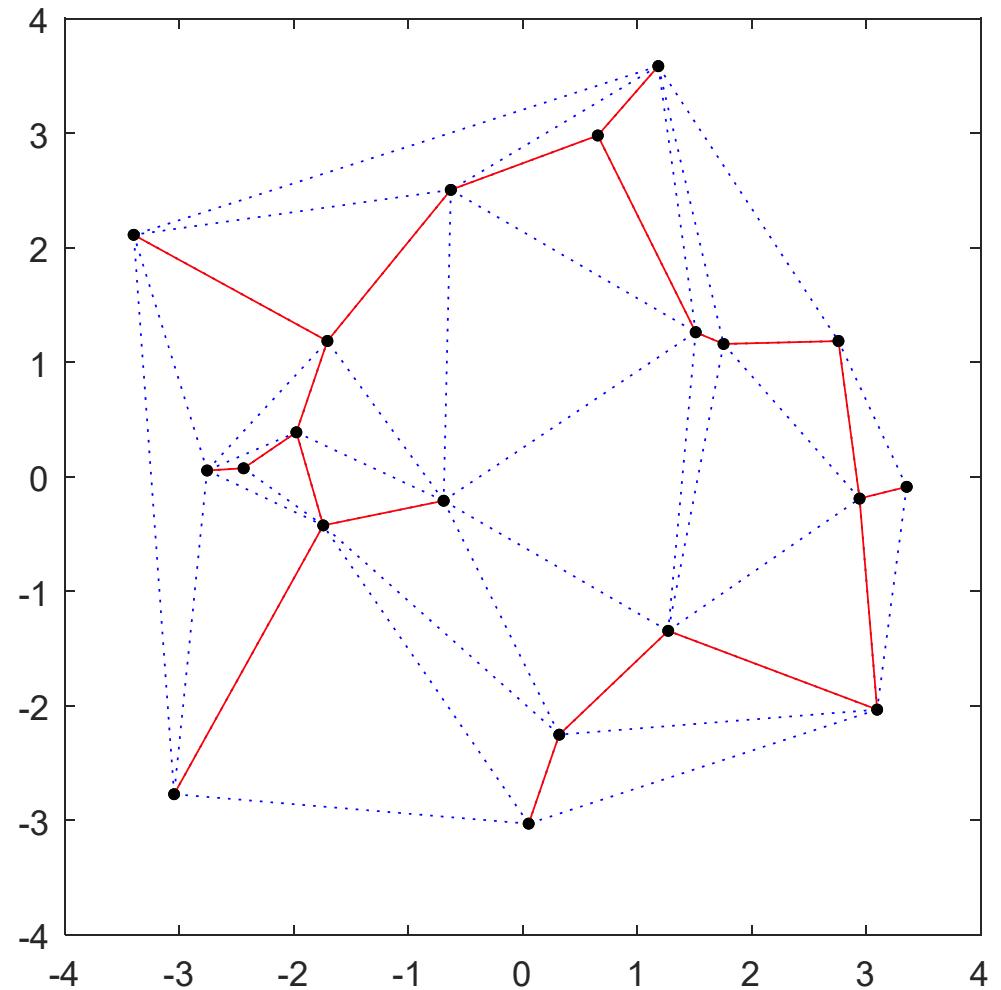
# Delaunay-Voronoi

- Delaunay triangulation is straight-line dual of Voronoi diagram
  - Can easily convert from one to another



# Minimum Spanning Tree

- Find the minimum weight set of arcs that connect all nodes in a graph
  - *Undirected* arcs: calculated, via Kruskal's algorithm,  $O(m \log n)$ ,  $m$  arcs,  $n$  nodes
  - *Directed* arcs: calculated, via Edmond's branching algorithm, in  $O(mn)$ ,  $m$  arcs,  $n$  nodes



# Steiner Network



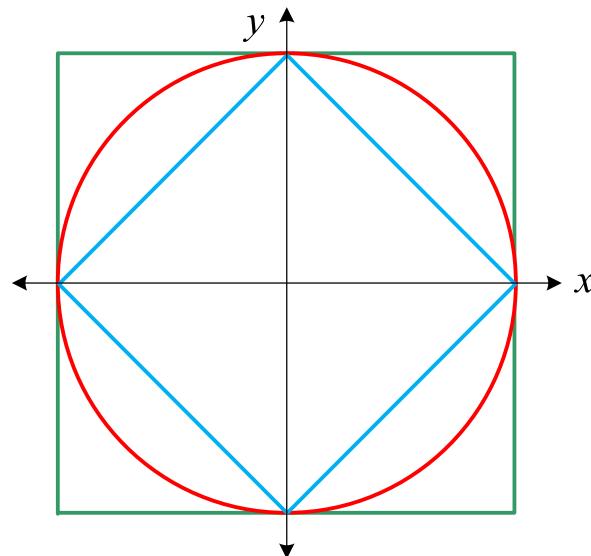
# Metric Distances

General  $\underline{l}_p$ :  $d_p(P_1, P_2) = \left[ |x_1 - x_2|^p + |y_1 - y_2|^p \right]^{\frac{1}{p}}, \quad p \geq 1$

Rectilinear :  $d_1(P_1, P_2) = |x_1 - x_2| + |y_1 - y_2|$   
 $(p=1)$

Euclidean :  $d_2(P_1, P_2) = \sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$   
 $(p=2)$

Chebychev :  $d_\infty(P_1, P_2) = \max_{(p \rightarrow \infty)} \{|x_1 - x_2|, |y_1 - y_2|\}$



# Chebychev Distances

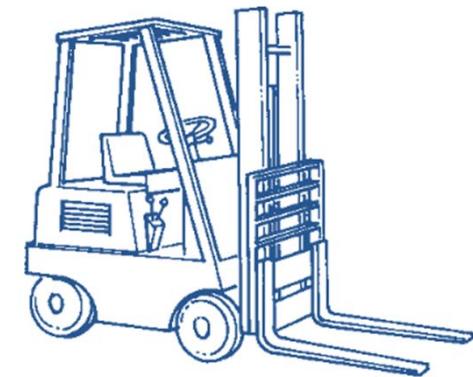
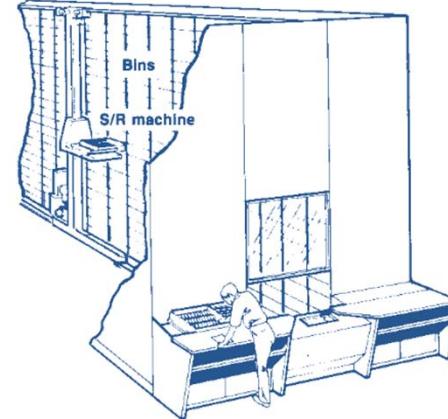
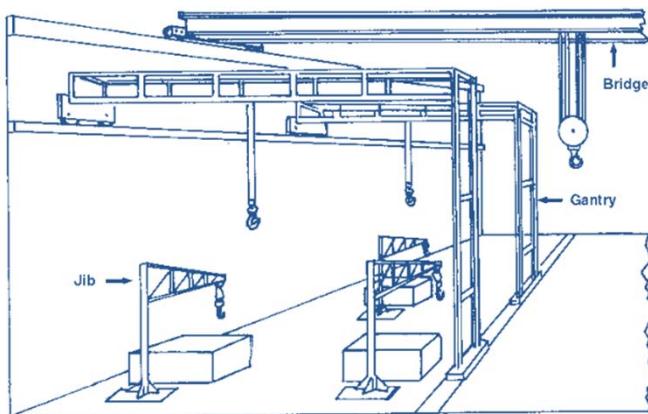
## Proof

Without loss of generality, let  $P_1 = (x, y)$ , for  $x, y \geq 0$ , and  $P_2 = (0, 0)$ . Then  $d_\infty(P_1, P_2) = \max\{x, y\}$  and  $d_p(P_1, P_2) = [x^p + y^p]^{1/p}$ .

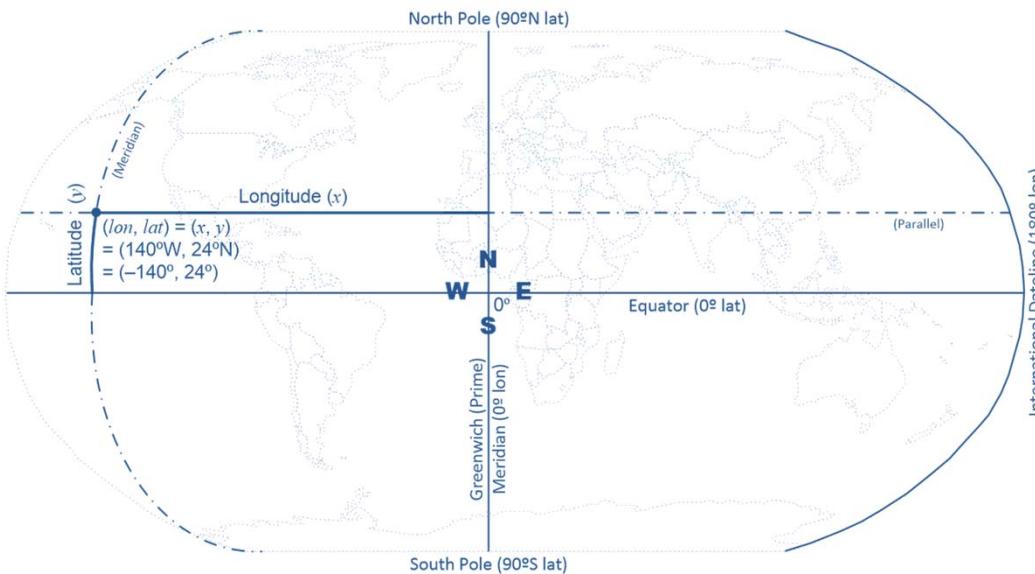
If  $x = y$ , then  $\lim_{p \rightarrow \infty} [x^p + y^p]^{1/p} = \lim_{p \rightarrow \infty} [2x^p]^{1/p} = \lim_{p \rightarrow \infty} [2^{1/p} x] = x$ .

If  $x < y$ , then  $\lim_{p \rightarrow \infty} [x^p + y^p]^{1/p} = \lim_{p \rightarrow \infty} \left[ \left( (x/y)^p + 1 \right) y^p \right]^{1/p} = \lim_{p \rightarrow \infty} \left( (x/y)^p + 1 \right)^{1/p} y = 1 \cdot y = y$ .

A similar argument can be made if  $x > y$ . ■



# Great Circle Distances



$$(lon_1, lat_1) = (x_1, y_1), \quad (lon_2, lat_2) = (x_2, y_2)$$

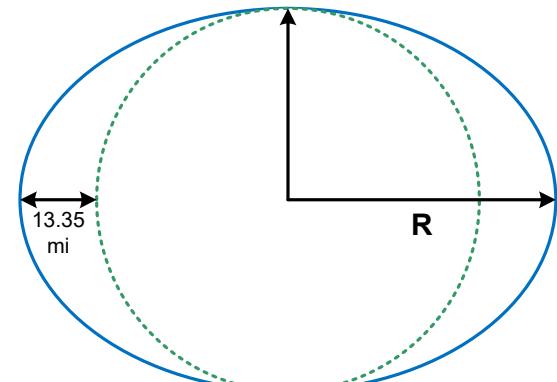
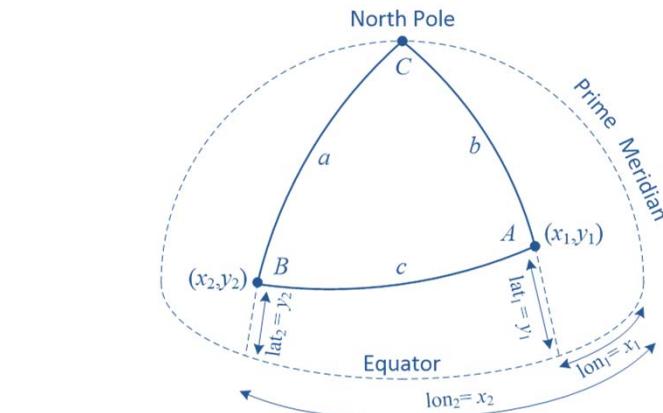
$d_{rad}$  = (great circle distance in radians of a sphere)

$$= \cos^{-1} [\sin y_2 \sin y_1 + \cos y_2 \cos y_1 \cos(x_1 - x_2)]$$

$R$  = (radius of earth at equator) – (bulge from north pole to equator)

$$= 3,963.34 - 13.35 \sin\left(\frac{y_1 + y_2}{2}\right) \text{ mi}, \quad = 6,378.388 - 21.476 \sin\left(\frac{y_1 + y_2}{2}\right) \text{ km}$$

$$d_{GC} = \text{distance } (x_1, y_1) \text{ to } (x_2, y_2) = \boxed{d_{rad} \cdot R}$$



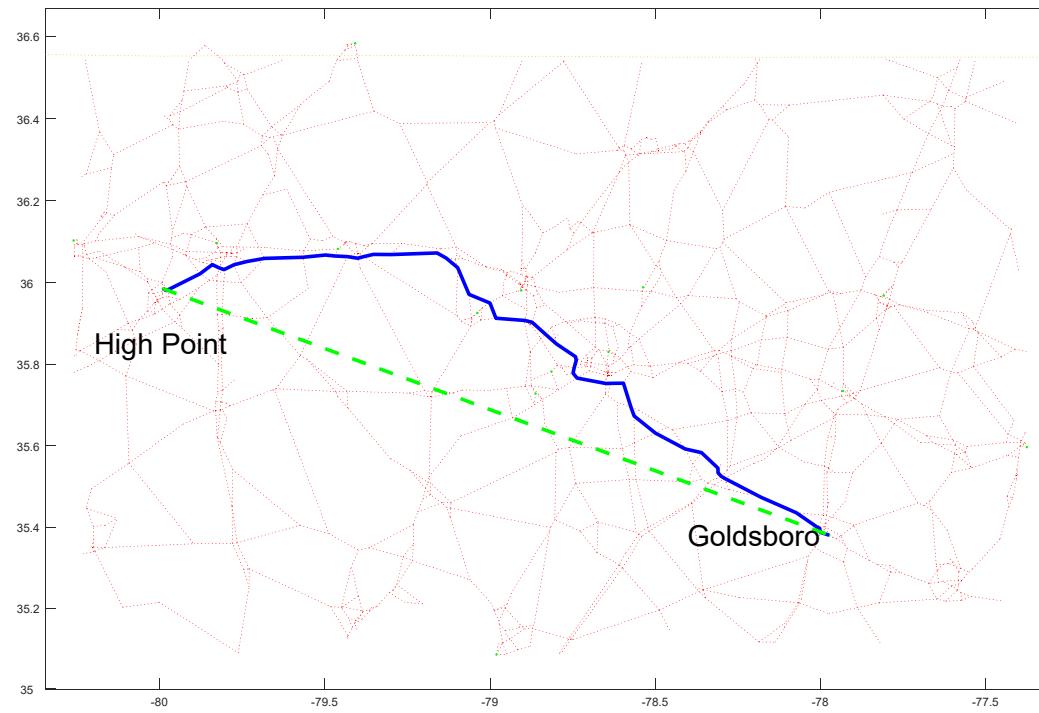
$$x_{deg} = \begin{cases} DD + \frac{MM}{60} + \frac{SS}{3,600}, & \text{if } E \text{ or } N \\ -DD - \frac{MM}{60} - \frac{SS}{3,600}, & \text{if } W \text{ or } S \end{cases}$$

# Circuit Factor

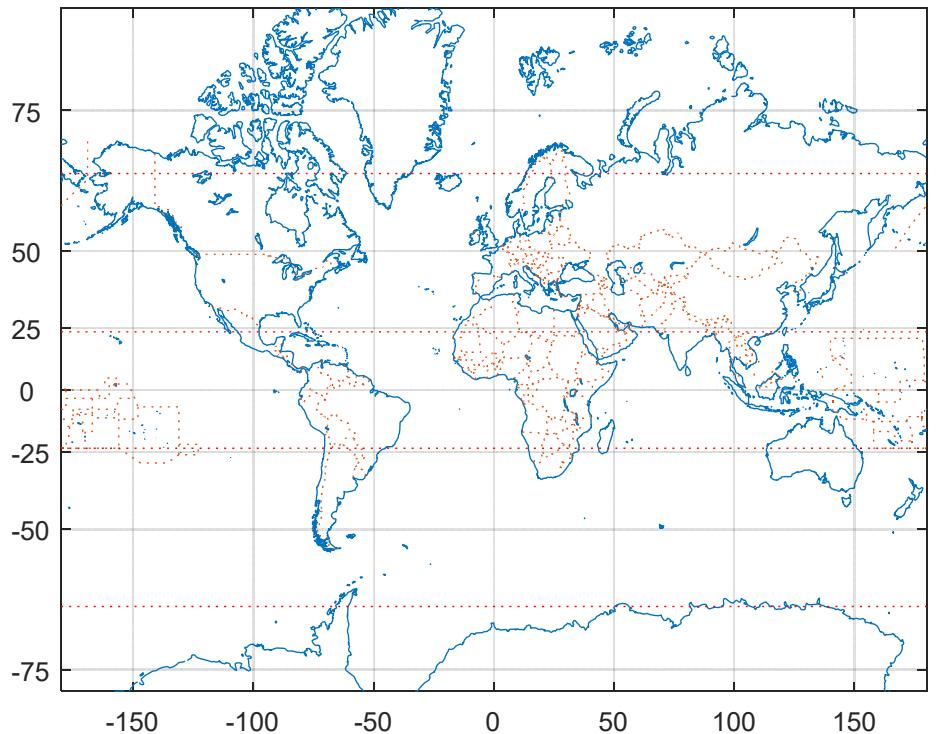
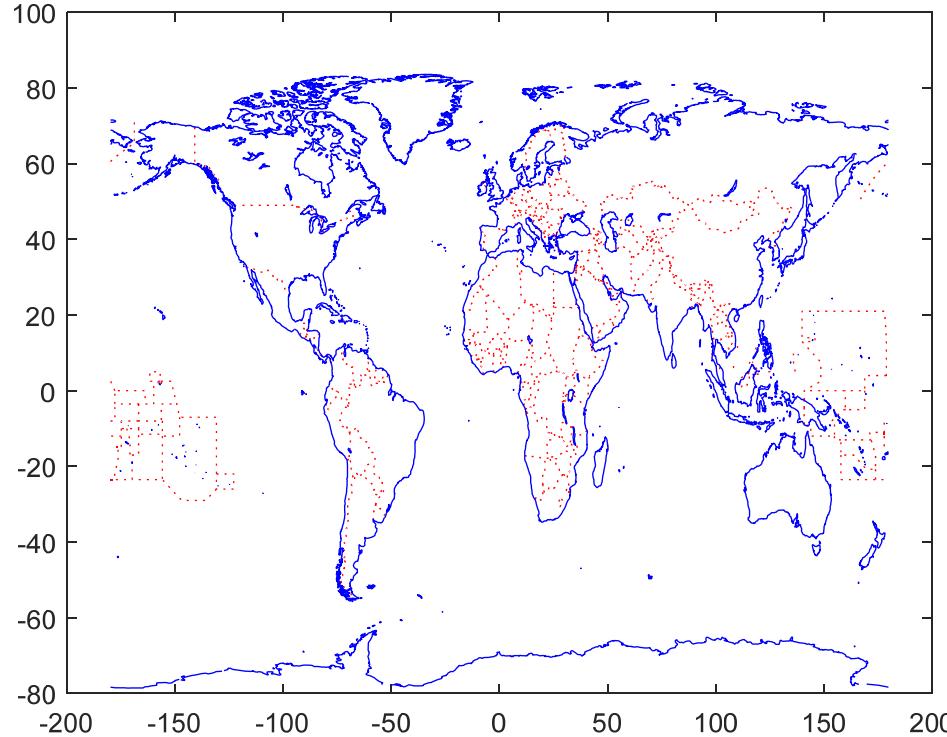
*Circuit Factor:*  $g = \sum \frac{d_{\text{road}_i}}{d_{GC_i}}$ , where usually  $1.15 \leq g \leq 1.5$

$d_{\text{road}} \approx g \cdot d_{GC}(P_1, P_2)$ , estimated road distance from  $P_1$  to  $P_2$

**From High Point to Goldsboro: Road = 143 mi, Great Circle = 121 mi, Circuitry = 1.19**



# Mercator Projection



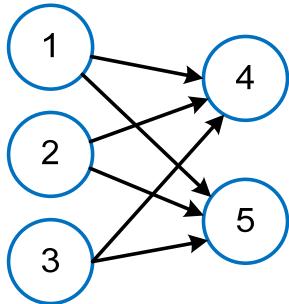
$$x_{\text{rad}} = \frac{x_{\text{deg}}}{180} \pi \quad \text{and} \quad x_{\text{deg}} = \frac{x_{\text{rad}} \cdot 180}{\pi}$$

$$x_{\text{proj}} = x$$

$$y_{\text{proj}} = \sinh^{-1}(\tan y)$$

$$y = \tan^{-1}(\sinh y_{\text{proj}})$$

# Metric Distances using dists



$$\mathbf{D} = \begin{array}{c|cc} & 4 & 5 \\ \hline 1 & \bullet & \bullet \\ 2 & \bullet & \bullet \\ 3 & \bullet & \bullet \\ \hline 3 \times 2 & & \\ n \times m & & \end{array} = \text{dists}(\mathbf{X1}, \mathbf{X2}, p), \quad p = \begin{cases} \text{'mi'} & \text{'km'} \\ 1 & 2 \\ \text{Inf} & \end{cases}$$

$d = 2$

$$\mathbf{X1} = [\bullet \quad \bullet], \mathbf{X2} = [\bullet \quad \bullet] \Rightarrow \mathbf{d} = [\bullet]$$

$$\mathbf{X1} = [\bullet \quad \bullet], \mathbf{X2} = \begin{bmatrix} \bullet & \bullet \\ \bullet & \bullet \\ \bullet & \bullet \end{bmatrix} \Rightarrow \mathbf{d} = [\bullet \quad \bullet \quad \bullet]$$

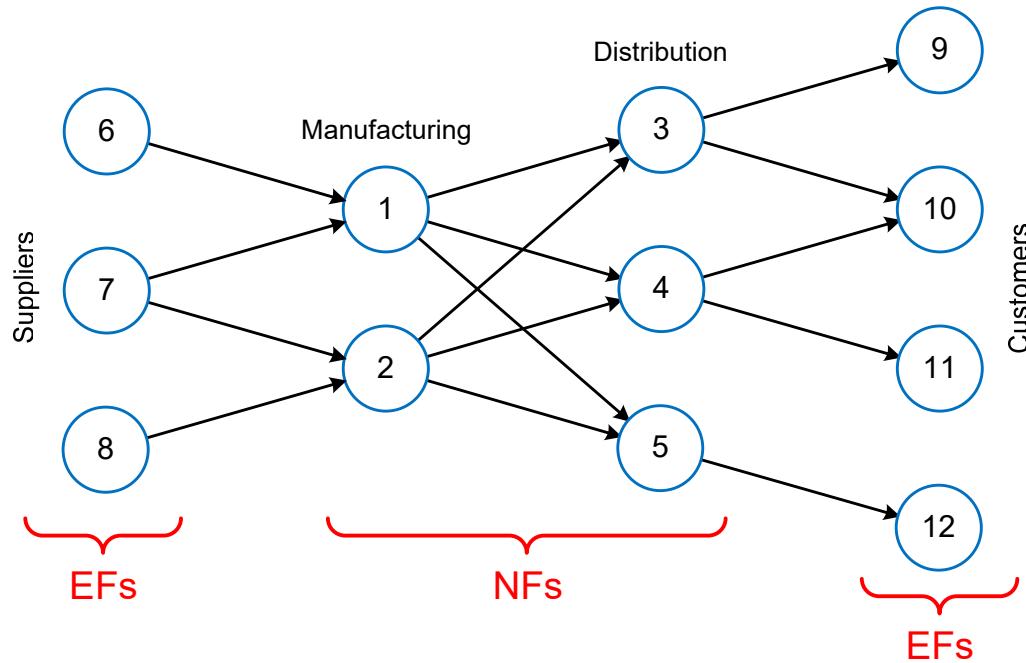
$$\mathbf{X1} = \begin{bmatrix} \bullet & \bullet \\ \bullet & \bullet \\ \bullet & \bullet \end{bmatrix}, \mathbf{X2} = \begin{bmatrix} \bullet & \bullet \\ \bullet & \bullet \\ \bullet & \bullet \end{bmatrix} \Rightarrow \mathbf{D} = \begin{bmatrix} \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \end{bmatrix}$$

$d = 1$

$$\mathbf{X1} = \begin{bmatrix} \bullet \\ \bullet \\ \bullet \end{bmatrix}, \mathbf{X2} = \begin{bmatrix} \bullet \\ \bullet \\ \bullet \end{bmatrix} \Rightarrow \mathbf{D} = \begin{bmatrix} \bullet & \bullet \\ \bullet & \bullet \\ \bullet & \bullet \end{bmatrix}$$

$$\mathbf{X1} = \begin{bmatrix} \bullet \\ \bullet \end{bmatrix}, \mathbf{X2} = [\bullet \quad \bullet] \Rightarrow \text{Error}$$

# Minsum Multifacility Location



$n = \text{no. of NFs}, \quad m = \text{no. of EFs}$

$\mathbf{X}_{n \times d} = \text{NF locations}, \quad \mathbf{P}_{m \times d} = \text{EF locations}$

	1	2	3	4	5	
1	0	0	+	+	+	
2	0	0	+	+	+	
3	0	0	0	0	0	
4	0	0	0	0	0	
5	0	0	0	0	0	

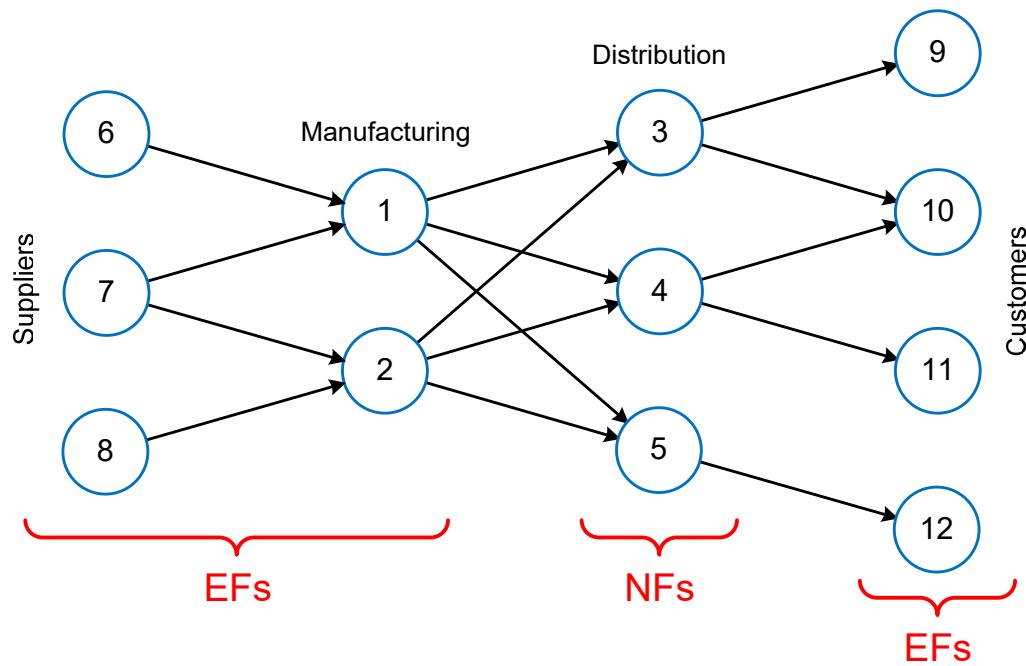
	6	7	8	9	10	11	12	
1	+	+	0	0	0	0	0	
2	0	+	+	0	0	0	0	
3	0	0	0	+	+	0	0	
4	0	0	0	0	+	+	0	
5	0	0	0	0	0	0	+	

$$TC(\mathbf{X}) = \sum_{j=1}^n \sum_{k=1}^n v_{jk} d(\mathbf{X}_j, \mathbf{X}_k) + \sum_{j=1}^n \sum_{i=1}^m w_{ji} d(\mathbf{X}_j, \mathbf{P}_i)$$

$$\mathbf{X}^* = \arg \min_{\mathbf{X}} TC(\mathbf{X})$$

$$TC^* = TC(\mathbf{X}^*)$$

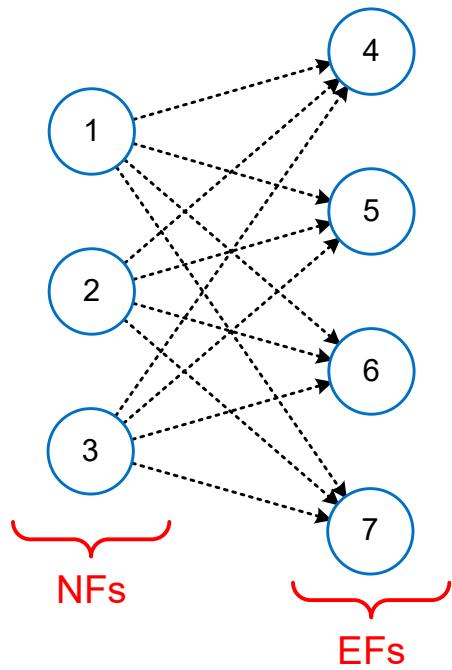
# Multiple Single-Facility Location



$$\begin{aligned}
 TC(\mathbf{X}) &= \sum_{j=1}^n \sum_{k=1}^n v_{jk} d(\mathbf{X}_j, \mathbf{X}_k) + \sum_{j=1}^n \sum_{i=1}^m w_{ji} d(\mathbf{X}_j, \mathbf{P}_i) \\
 &= \sum_{j=1}^n TC(\mathbf{X}_j)
 \end{aligned}$$

# Facility Location–Allocation Problem

- Determine both the location of  $n$  NFs and the allocation of flow requirements of  $m$  EFs that minimize TC



$w_{ji} = r_{ji} f_{ji} = (1)f_{ji}$  = flow between NF $j$  and EF $i$

$w_i$  = total flow requirements of EF $i$

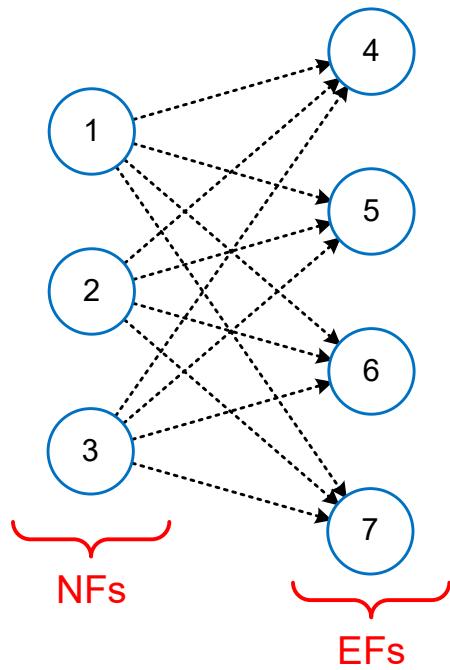
$$TC(\mathbf{X}, \mathbf{W}) = \sum_{j=1}^n \sum_{i=1}^m w_{ji} d(\mathbf{X}_j, \mathbf{P}_i)$$

$$\mathbf{X}^*, \mathbf{W}^* = \arg \min_{\mathbf{X}, \mathbf{W}} \left\{ TC(\mathbf{X}, \mathbf{W}) : \sum_{j=1}^n w_{ji} = w_i, w_{ji} \geq 0 \right\}$$

$$TC^* = TC(\mathbf{X}^*, \mathbf{W}^*)$$

# Integrated Formulation

- If there are no capacity constraints on NFs, it is optimal to always satisfy all the flow requirements of an EF from its closest NF
- Requires search of  $(n \times d)$ -dimensional TC that combines location with allocation



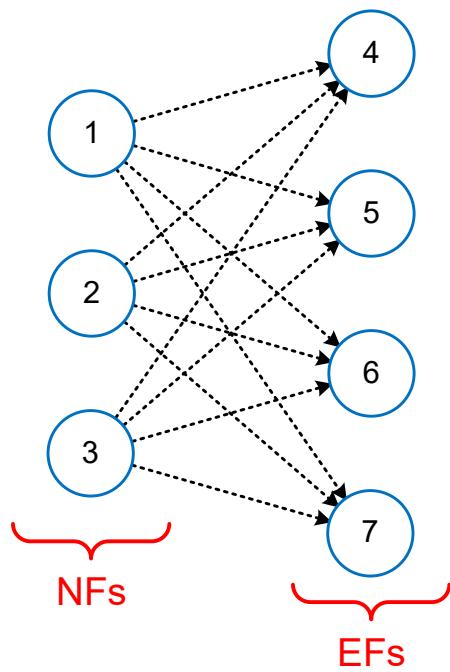
$$\alpha_i(\mathbf{X}) = \arg \min_j d(\mathbf{X}_j, \mathbf{P}_i)$$

$$TC(\mathbf{X}) = \sum_{i=1}^m w_i d(\mathbf{X}_{\alpha_i(\mathbf{X})}, \mathbf{P}_i)$$

$$\mathbf{X}^* = \arg \min_{\mathbf{X}} TC(\mathbf{X})$$

$$TC^* = TC(\mathbf{X}^*)$$

# Alternating Formulation



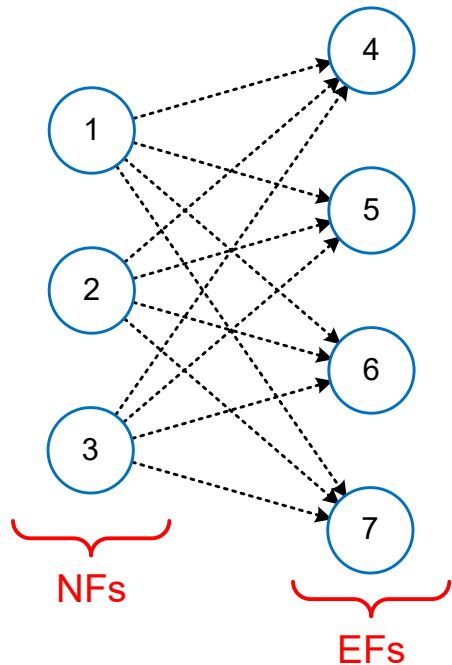
- Alternate between finding locations and finding allocations until no further TC improvement
- Requires  $n$   $d$ -dimensional location searches together with separate allocation procedure
- Separating location from allocation allows other types of location and/or allocation procedures to be used:
  - Allocation with NF with capacity constraints (solved as minimum cost network flow problem)
  - Location with some NFs at fixed locations

$$\text{allocate}(\mathbf{X}) = \begin{bmatrix} w_{ji} \end{bmatrix} = \begin{cases} w_i, & \text{if } \arg \min_k d(\mathbf{X}_k, \mathbf{P}_i) = j \\ 0, & \text{otherwise} \end{cases}$$

$$TC(\mathbf{X}, \mathbf{W}) = \sum_{j=1}^n \sum_{i=1}^m w_{ji} d(\mathbf{X}_j, \mathbf{P}_i)$$

$$\text{locate}(\mathbf{W}, \mathbf{X}) = \arg \min_{\mathbf{X}} TC(\mathbf{X}, \mathbf{W})$$

# ALA: Alternate Location–Allocation



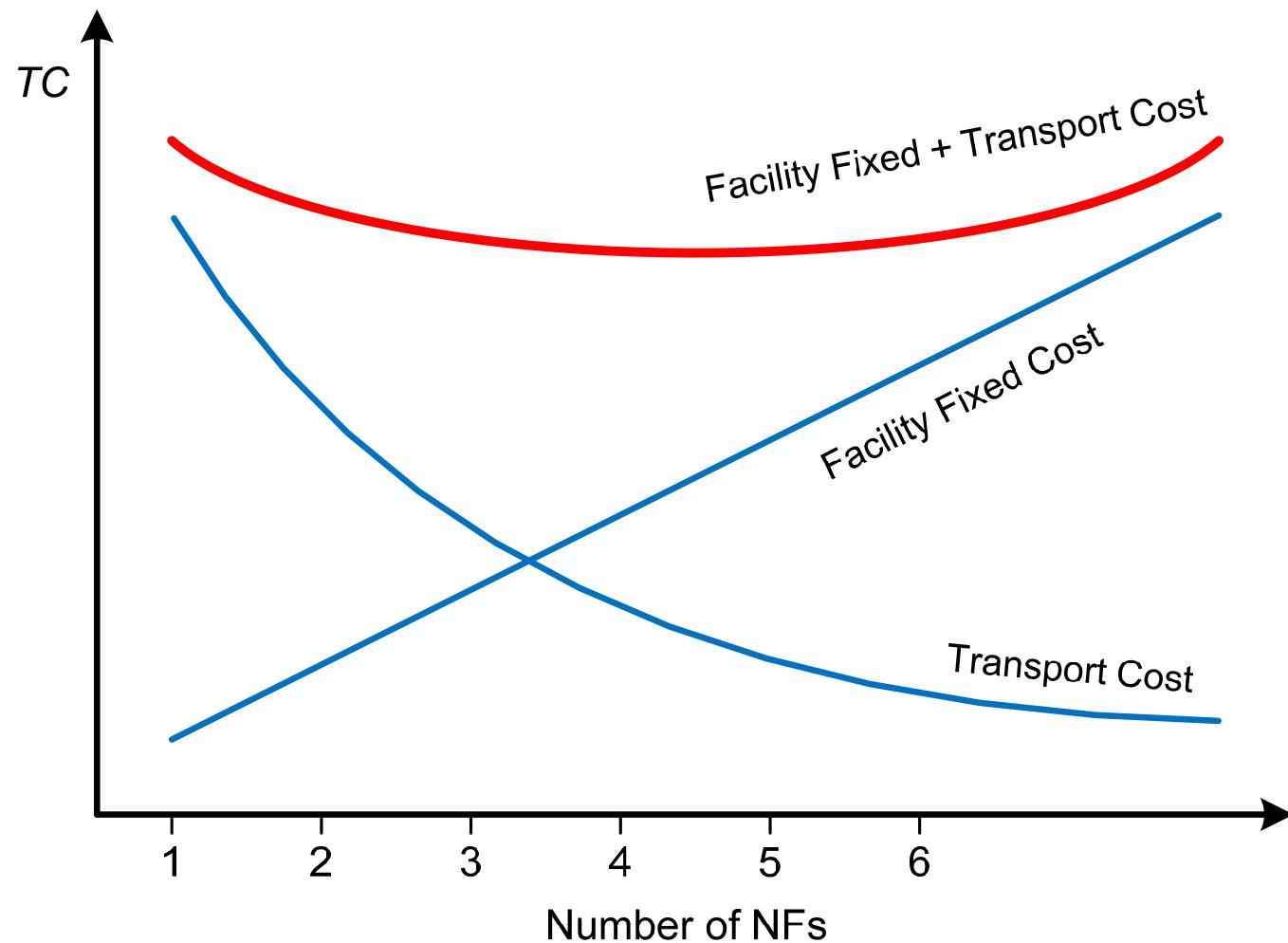
```
procedure ala(X)
    TC  $\leftarrow \infty$ , done  $\leftarrow$  false
repeat
    W'  $\leftarrow$  allocate(X)
    X'  $\leftarrow$  locate(W', X)
    TC'  $\leftarrow$  TC(X', W')
    if TC' < TC
        TC  $\leftarrow$  TC', X  $\leftarrow$  X', W  $\leftarrow$  W'
    else
        done  $\leftarrow$  true
    endif
until done = true
return X, W
```

```
%% ALA Matlab Code
x = randX(P, n);
TC = Inf; done = false;
while ~done
    Wi = alloc_h(X);
    Xi = loc_h(Wi, X);
    TCi = TCh(Wi, Xi);
    if TCi < TC
        TC = TCi; X = Xi; W = Wi;
    else
        done = true;
    end
end
X, W
```

# Best Retail Warehouse Locations

Number of Locations	Average Transit Time (days)	Warehouse Location		
1	2.20	Bloomington, IN		
2	1.48	Ashland, KY	Palmdale, CA	
3	1.29	Allentown, PA	Palmdale, CA	McKenzie, TN
4	1.20	Edison, NJ Meridian, MS	Palmdale; CA	Chicago, IL
5	1.13	Madison, NJ Dallas, TX	Palmdale, CA Macon, GA	Chicago, IL
6	1.08	Madison, NJ Dallas, TX	Pasadena, CA Macon, GA	Chicago, IL Tacoma, WA
7	1.07	Madison, NJ Dallas, TX Lakeland, FL	Pasadena, CA Gainesville, GA	Chicago, IL Gainesville, GA Tacoma, WA
8	1.05	Madison, NJ Dallas, TX Lakeland, FL	Pasadena, CA Gainesville, GA Denver, CO	Chicago, IL Gainesville, GA Tacoma, WA
9	1.04	Madison, NJ Dallas, TX Lakeland. FL	Alhambra, CA Gainesville, GA Denver, CO	Chicago, IL Tacoma, WA Oakland, CA
10	1.04	Newark, NJ <u>Palistine</u> , TX Lakeland, FL Mansfield, OH	Alhambra, CA Gainesville, GA Denver, CO	Rockford, IL Tacoma, WA Oakland. CA

# Optimal Number of NFs



# Uncapacitated Facility Location Problem

$$\text{Min } TC = \sum_{i=1}^n k_i y_i + \sum_{i=1}^n \sum_{j=1}^m c_{ij} x_{ij}$$

subject to

$$\begin{aligned} \sum_{i=1}^n x_{ij} &= 1, \quad j = 1, \dots, m \\ y_i &\geq x_{ij}, \quad i = 1, \dots, n; j = 1, \dots, m \\ 0 \leq x_{ij} &\leq 1, \quad i = 1, \dots, n; j = 1, \dots, m \\ y_i &\in \{0,1\}, \quad i = 1, \dots, n, \end{aligned}$$

where

$k_i$  = fixed cost of establishing a NF at site  $i$

$c_{ij}$  = variable cost to serve all of EF  $j$ 's demand from site  $i$

$y_i$  = 1, if NF established at site  $i$ ; 0, otherwise

$x_{ij}$  = fraction of EF  $j$ 's demand served from NF at site  $i$ .

# UFLADD: Add Construction Procedure

```

procedure ufladd(k, C)
    Y  $\leftarrow \{\}$ 
    TC  $\leftarrow \infty$ , done  $\leftarrow \text{false}$ 
repeat
    TC'  $\leftarrow \infty$ 
    for i'  $\in \{1, \dots, n\} \setminus Y$ 
        
$$TC'' \leftarrow \sum_{h \in Y \cup i'} k_h + \sum_{j=1}^m \min_{h \in Y \cup i'} c_{hj}$$

        if TC'' < TC'
            TC'  $\leftarrow TC'', i \leftarrow i'$ 
        endif
    endfor
    if TC' < TC
        TC  $\leftarrow TC', Y \leftarrow Y \cup i$ 
    else
        done  $\leftarrow \text{true}$ 
    endif
until done = true
return Y, TC

```

```

%% UFLADD Matlab code, given k and C as inputs
y = [];
TC = Inf; done = false;
while ~done
    TC1 = Inf; % Stops if y = all NF,
    for i1 = setdiff(1:size(C, 1), y) % since i1 = []
        TC2 = sum(k([y i1])) + sum(min(C([y i1], :), [], 1));
        if TC2 < TC1
            TC1 = TC2; i = i1;
        end
    end
    if TC1 < TC % not true if y = all NF, since TC1 = Inf
        TC = TC1; y = [y i];
    else
        done = true;
    end
end
y, TC

```

# UFLXCHG: Exchange Improvement Procedure

```

procedure uflxchg(k,C,Y)
     $TC \leftarrow \sum_{i \in Y} k_i + \sum_{j=1}^m \min_{i \in Y} c_{ij}$ 
     $TC' \leftarrow \infty$ , done  $\leftarrow$  false
    while  $|y| > 1$  and done = false
        for  $i' \in y$ 
            for  $j' \in \{1, \dots, n\} \setminus Y$ 
                 $Y' \leftarrow Y \setminus i' \cup j'$ 
                 $TC'' \leftarrow \sum_{i \in Y'} k_i + \sum_{j=1}^m \min_{i \in Y'} c_{ij}$ 
                if  $TC'' < TC'$ 
                     $TC' \leftarrow TC'', i \leftarrow i', j \leftarrow j'$ 
                endif
            endfor
        endfor
        if  $TC' < TC$ 
             $TC \leftarrow TC', Y \leftarrow Y \setminus i \cup j$ 
        else
            done  $\leftarrow$  true
        endif
    endwhile
    return Y, TC

```

```

%% UFLXCHG Matlab code, given k, C, and y as inputs
TC = sum(k(y)) + sum(min(C(y,:),[],1));
TC1 = Inf; done = false;
while length(y) > 1 && ~done
    for i1 = y
        for j1 = setdiff(1:size(C,1),y)
            y1 = [setdiff(y,i1) j1];
            TC2 = sum(k(y1)) + sum(min(C(y1,:),[],1));
            if TC2 < TC1
                TC1 = TC2; i = i1; j = j1;
            end
        end
    end
    if TC1 < TC
        TC = TC1; y = [setdiff(y,i) j];
    else
        done = true;
    end
end
y, TC

```

# Modified UFLADD

```
procedure ufladd(k,C,Y,p)
Y ← {}
TC ← ∞, done ← false
repeat
    TC' ← ∞
    for i' ∈ {1,...,n} \ Y
        TC'' ←  $\sum_{h \in Y \cup i'} k_h + \sum_{j=1}^m \min_{h \in Y \cup i'} c_{hj}$ 
        if TC'' < TC'
            TC' ← TC'', i ← i'
        endif
    endfor
    if (p = {} and TC' < TC) or (p ≠ {} and |Y| < p)
        TC ← TC', Y ← Y ∪ i
    else
        done ← true
    endif
until done = true
return Y, TC
```

# UFL: Hybrid Algorithm

```
procedure ufl(k,C)
Y', TC' ← ufladd(k,C)
done ← false
repeat
    Y, TC ← uflxchg(k,C,Y')
    if Y ≠ Y'
        Y', TC' ← ufladd(k,C,Y)
        Y'', TC'' ← ufldrop(k,C,Y)
        if TC'' < TC'
            TC' ← TC'', Y' ← Y''
        endif
        if TC' ≥ TC
            done ← true
        endif
    else
        done ← true
    endif
until done = true
return Y, TC
```

```
%% UFL Matlab code, given k and C
[y1,TC1] = ufladd(k,C);
done = false;
while ~done
    [y,TC] = uflxchg(k,C,y1);
    if ~isequal(y,y1)
        [y1,TC1] = ufladd(k,C,y);
        [y2,TC2] = ufldrop(k,C,y);
        if TC2 < TC1
            TC1 = TC2; y1 = y2;
        end
        if TC1 >= TC
            done = true;
        end
    else
        done = true;
    end
end
y, TC
```

# PMEDIAN: Discrete Location Algorithm

```
procedure pmedian( $p, \mathbf{C}$ )
 $Y \leftarrow ufladd(0, \mathbf{C}, \{\}, p)$ 
 $Y, TC \leftarrow uflxchg(0, \mathbf{C}, Y)$ 
 $Y' \leftarrow ufldrop(0, \mathbf{C}, \{\}, p)$ 
 $Y', TC' \leftarrow uflxchg(0, \mathbf{C}, Y')$ 
if  $TC' < TC$ 
     $TC \leftarrow TC', Y \leftarrow Y'$ 
endif
return  $Y, TC$ 
```

# Aggregate Demand Point Data Sources

- Aggregate demand point: centroid of population
  - Good rule of thumb: use 100x number of NFs ( $\approx$  1000 pts provides good coverage for locating  $\approx$  10 NFs)
1. City data: **ONLY USE FOR LABELING!**, not as demand points
  2. 3-digit ZIP codes:  $\approx$  1000 pts covering U.S., = 20 pts NC
  3. County data:  $\approx$  3000 pts covering U.S., = 100 pts NC
    - Grouped by state or MSA (Metropolitan Statistical Area)
    - MSA = defined by set of counties (382 MSAs in U.S.)
    - FIPS code = 5-digit state-county FIPS code  
= 2-digit state code + 3-digit county code  
= 37183 = 37 NC FIPS + 183 Wake FIPS
    - MSA List: [www.census.gov/population/metro/files/lists/2009/List1.txt](http://www.census.gov/population/metro/files/lists/2009/List1.txt)
  4. 5-digit ZIP codes: > 35K pts U.S.,  $\approx$  1000 pts NC
  5. Census Block Group: > 220K pts U.S.,  $\approx$  1000 pts Raleigh-Durham MSA
    - Grouped by state, county, or MSA

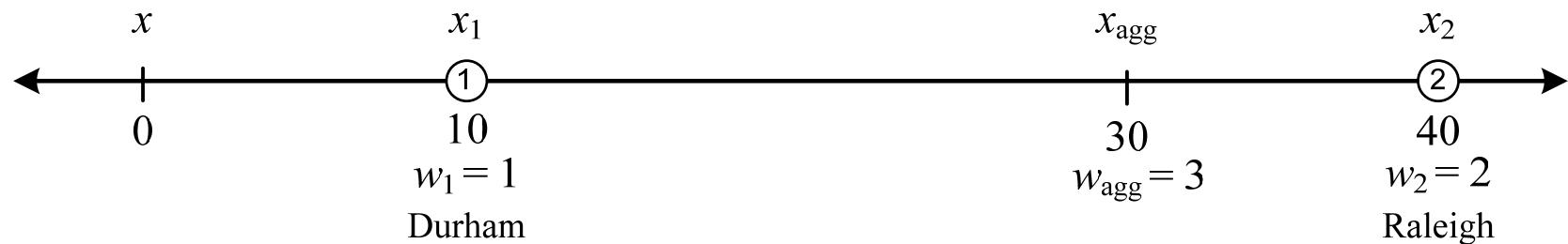
# City vs MSA Population Data

Rank	City; State	2010 population	2012 population
1	New York City; New York	8,175,133	<b>8,336,697</b>
2	Los Angeles; California	3,792,621	<b>3,857,799</b>
3	Chicago; Illinois	2,695,598	<b>2,714,856</b>
4	Houston; Texas	2,099,451	<b>2,160,821</b>
5	Philadelphia; Pennsylvania	1,526,006	<b>1,547,607</b>
6	Phoenix; Arizona	1,445,632	<b>1,488,750</b>
7	San Antonio; Texas	1,327,407	<b>1,382,951</b>
8	San Diego; California	1,307,402	<b>1,338,348</b>
9	Dallas; Texas	1,197,816	<b>1,241,162</b>
10	San Jose; California	945,942	<b>982,765</b>
11	Austin; Texas	790,390	<b>842,592</b>
12	Jacksonville; Florida	821,784	<b>836,507</b>
13	Indianapolis; Indiana	820,445	<b>834,852</b>
14	San Francisco; California	805,235	<b>825,863</b>
15	Columbus; Ohio	787,033	<b>809,798</b>
16	Fort Worth; Texas	741,206	<b>777,992</b>
17	Charlotte; North Carolina	731,424	<b>775,202</b>
18	Detroit; Michigan	713,777	<b>701,475</b>
19	El Paso; Texas	649,121	<b>672,538</b>
20	Memphis; Tennessee	646,889	<b>655,155</b>

Metropolitan Area	2010 Population	City
New York-Northern NJ-Long Island, NY-NJ-PA	18,897,109	New York
Los Angeles-Long Beach-Santa Ana, CA	12,828,837	Los Angeles
Chicago-Joliet-Naperville, IL-IN-WI	9,461,105	Chicago
Dallas-Fort Worth-Arlington, TX	6,371,773	Dallas
Philadelphia-Camden-Wilmington, PA-NJ-DE-MD	5,965,343	Philadelphia
Houston-Sugar Land-Baytown, TX	5,946,800	Houston
Washington-Arlington-Alexandria, DC-VA-MD-WV	5,582,170	Washington
Miami-Fort Lauderdale-Pompano Beach, FL	5,564,635	Miami
Atlanta-Sandy Springs-Marietta, GA	5,268,860	Atlanta
Boston-Cambridge-Quincy, MA-NH	4,552,402	Boston
San Francisco-Oakland-Fremont, CA	4,335,391	San Francisco
Detroit-Warren-Livonia, MI	4,296,250	Detroit
Riverside-San Bernardino-Ontario, CA	4,224,851	Riverside
Phoenix-Mesa-Glendale, AZ	4,192,887	Phoenix
Seattle-Tacoma-Bellevue, WA	3,439,809	Seattle
Minneapolis-St. Paul-Bloomington, MN-WI	3,279,833	Minneapolis
San Diego-Carlsbad-San Marcos, CA	3,095,313	San Diego
St. Louis, MO-IL	2,812,896	St. Louis
Tampa-St. Petersburg-Clearwater, FL	2,783,243	Tampa
Baltimore-Towson, MD	2,710,489	Baltimore

# Demand Point Aggregation

- *Existing facility (EF)*: actual physical location of demand source
- *Aggregate demand point*: single location representing multiple demand sources



- For minisum location, would like for any location  $x$ :

$$(w_1 + w_2)d(x, x_{\text{agg}}) = w_1d(x, x_1) + w_2d(x, x_2), \quad \text{w.l.o.g., let } x = 0, x_1, x_2 > 0$$

$$(w_1 + w_2)x_{\text{agg}} = w_1x_1 + w_2x_2$$

$$x_{\text{agg}} = \frac{w_1x_1 + w_2x_2}{w_1 + w_2} \Rightarrow \text{centroid}$$

# Bottom-Up vs Top-Down Analysis

- Bottom-Up: HW 3 Q 3

$\mathbf{P}_{3 \times 2}$  = lon-lat of EFs

$$\mathbf{f} = [45, 25, 35] \text{ (TL/yr)}$$

$$r = 2 \text{ ($/TL-mi)}$$

$$g = \frac{1}{3} \left[ \frac{d_{RD}(\mathbf{P}_1, \mathbf{P}_2)}{d_{GC}(\mathbf{P}_1, \mathbf{P}_2)} + \frac{d_{RD}(\mathbf{P}_2, \mathbf{P}_3)}{d_{GC}(\mathbf{P}_2, \mathbf{P}_3)} + \frac{d_{RD}(\mathbf{P}_3, \mathbf{P}_1)}{d_{GC}(\mathbf{P}_3, \mathbf{P}_1)} \right]$$

$$TC(\mathbf{x}) = \sum_{i=1}^3 f_i r g d_{GC}(\mathbf{x}, \mathbf{P}_i)$$

$$\mathbf{x}^* = \arg \min_{\mathbf{x}} TC(\mathbf{x})$$

$$TC^* = TC(\mathbf{x}^*)$$

$\mathbf{x}^{\text{cary}}$  = lon-lat of Cary

$$TC^{\text{cary}} = TC(\mathbf{x}^{\text{cary}})$$

$$\Delta TC = TC^{\text{cary}} - TC^*$$

- Top-Down: estimate  $r$  (circuit factor cancels, so not needed)

$TC^{\text{cary}}$  = current known  $TC$

$$\mathbf{f} = [450, 250, 350] \text{ (ton/yr)}$$

$$r_{\text{nom}} = \frac{TC^{\text{cary}}}{\sum_{i=1}^3 f_i d_{GC}(\mathbf{x}^{\text{cary}}, \mathbf{P}_i)} \text{ ($/ton-mi$)}$$

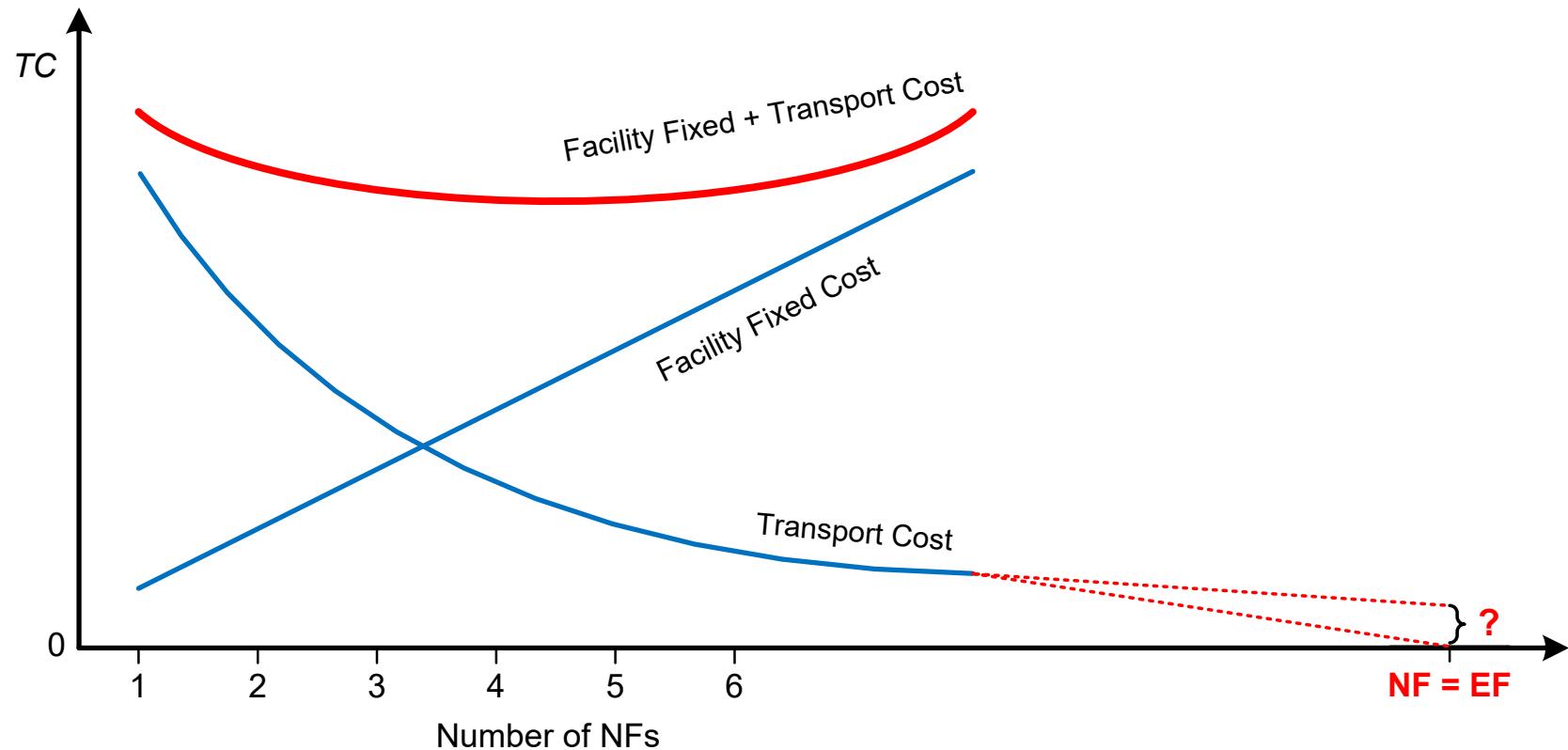
$$TC(\mathbf{x}) = \sum_{i=1}^3 f_i r_{\text{nom}} d_{GC}(\mathbf{x}, \mathbf{P}_i)$$

$$\mathbf{x}^* = \arg \min_{\mathbf{x}} TC(\mathbf{x})$$

$$TC^* = TC(\mathbf{x}^*)$$

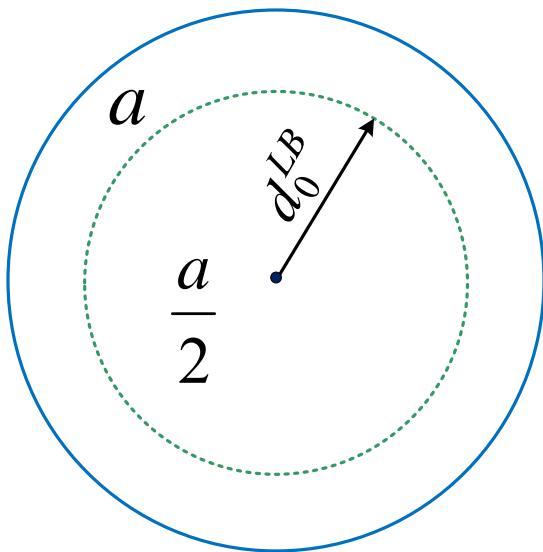
$$\Delta TC = TC^{\text{cary}} - TC^*$$

# Transport Cost if NF at every EF



$$TC = \underbrace{\sum_{i=1}^n k_i y_i}_{\text{fixed cost}} + \underbrace{\sum_{i=1}^n \sum_{j=1}^m c_{ij} x_{ij}}_{\text{transport cost}}$$

# Area Adjustment for Aggregate Data Distances

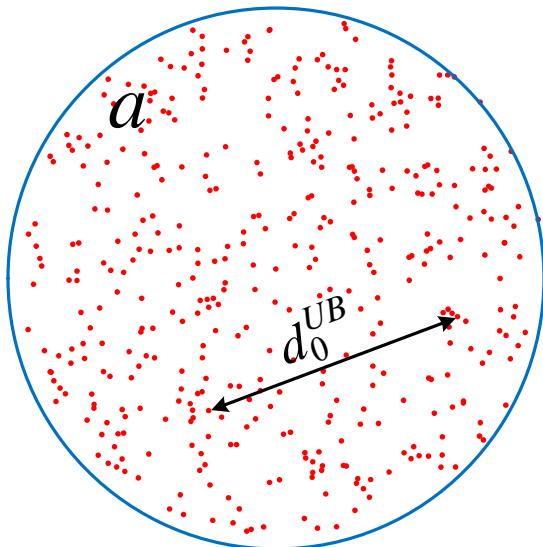


- $LB$ : average distance from center to all points in area
- $UB$ : average distance between all random pairs of points

$$\frac{a}{2} = \pi (d_0^{LB})^2$$

$$d_0^{LB} = \sqrt{\frac{a}{2\pi}} \approx 0.40\sqrt{a}$$

$$d_0^{UB} = \frac{32}{15} \frac{\sqrt{a}}{\pi \Gamma(5/2)} \approx 0.51\sqrt{a}$$

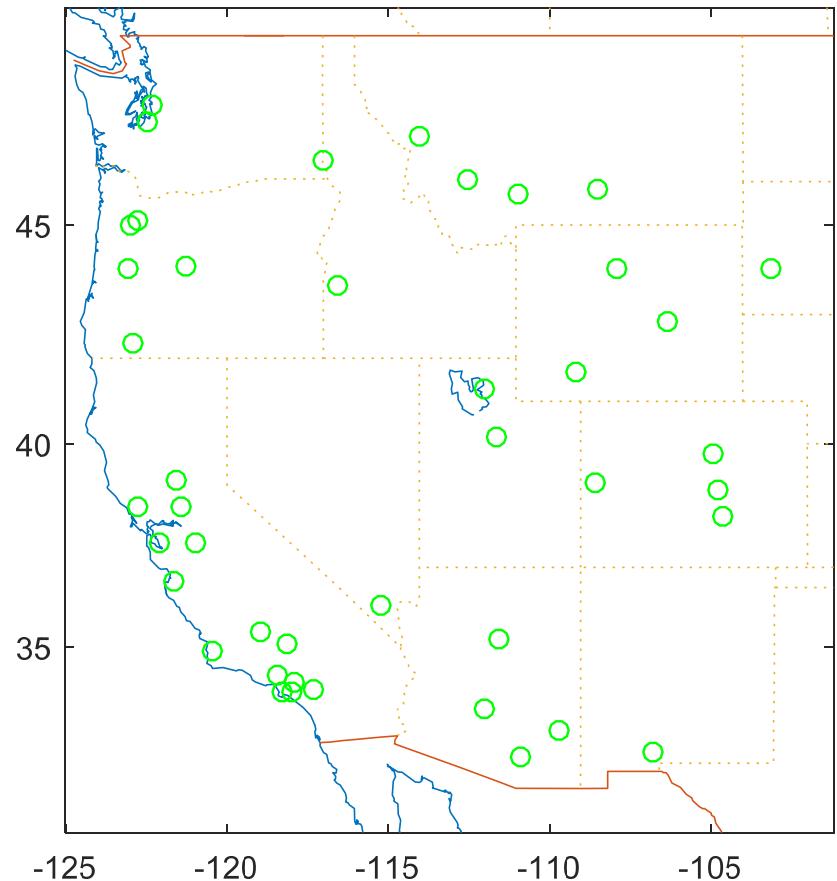


$$d_0 = \sqrt{d_0^{LB} d_0^{UB}} \approx 0.45\sqrt{a}$$

$$d_a(\mathbf{X}_1, \mathbf{X}_2) = \max \left\{ d_{GC}(\mathbf{X}_1, \mathbf{X}_2), 0.45 \max \left\{ \sqrt{a_1}, \sqrt{a_2} \right\} \right\}$$

Mathai, A.M.,  
*An Intro to Geo  
 Prob*, p. 207 (2.3.68)

# Popco Bottling Company Example



- **Problem:** Popco currently has 43 bottling plants across the western U.S. and wants to know if they should consider reducing or adding plants to improve their profitability.
- **Solution:** Formulate as an UFL to determine the number of plants that minimize Popco's production, procurement, and distribution costs.

# Popco Bottling Company Example

- Following representative information is available for each of  $N$  current plants (DC)  $i$ :

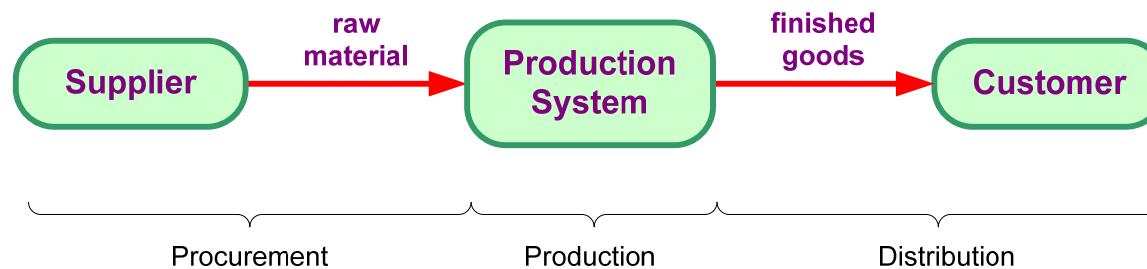
$xy_i$  = location

$f_i^{DC}$  = aggregate production (tons)

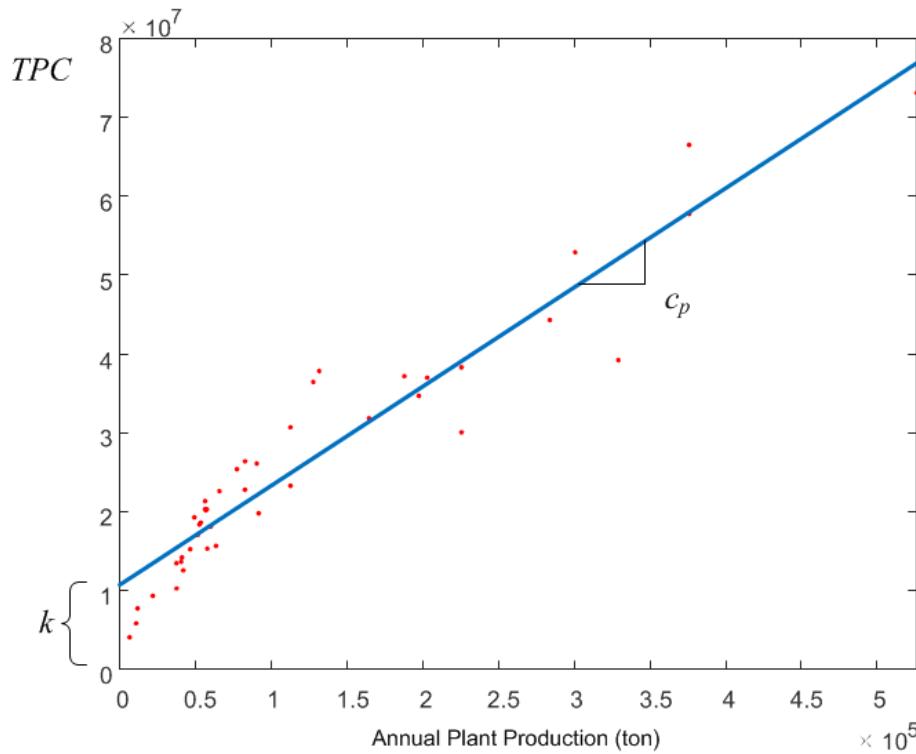
$TPC_i$  = total production and procurement cost

$TDC_i$  = total distribution cost

- Assuming plants are (monetarily) weight gaining, so UFL can ignore inbound procurement costs related to location



# Popco Bottling Company Example

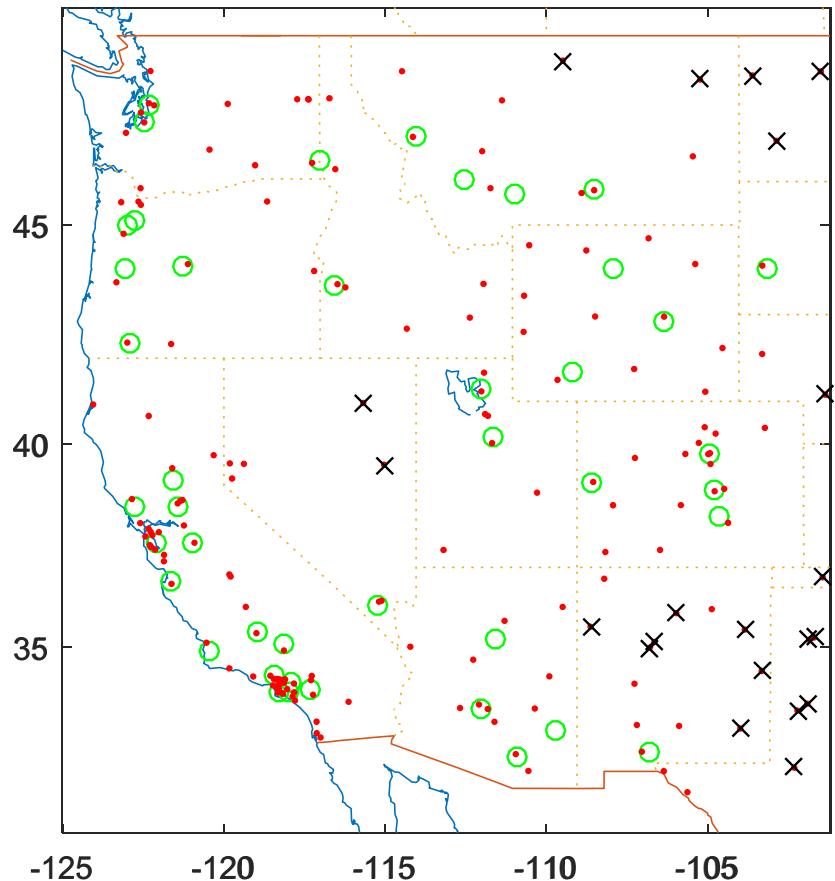


- Difficult to estimate fixed cost of each new facility because this cost must not include any cost related to quantity of product produced at facility.
1. Use plant (DC) production costs to find UFL fixed costs via linear regression
    - variable production costs  $c_p$  do not change and can be cut

$$TPC = \sum_{i \in N} TPC_i = \sum_{i \in N} (k + c_p f_i^{DC})$$

(only keep  $k$  for UFL)

# Popco Bottling Company Example



2. Allocate all 3-digit ZIP codes to closest plant (up to 200 mi max) to serve as aggregate customer demand points.

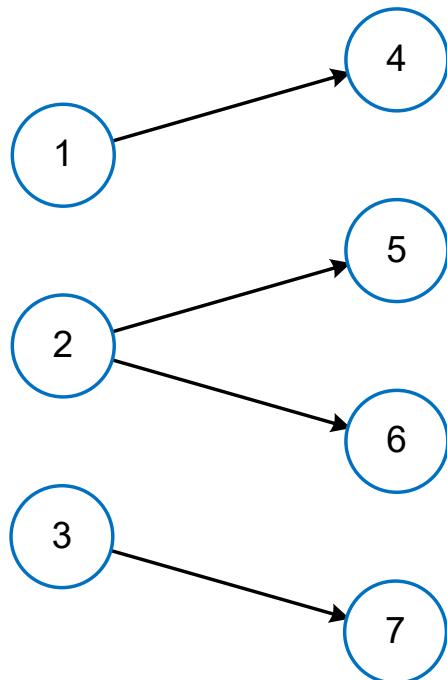
$$M_i = \left\{ j : \arg \min_h d_{hj}^a = i \text{ and } d_{ij}^a \leq d_{\max} \right\}$$

$$d_{\max} = 200 \text{ mi}$$

$$M = \bigcup_{i \in N} M_i$$

# Popco Bottling Company Example

3. Allocate each plant's demand (tons of product) to each of its customers based on its population.



$$f_{j \in M_i} = f_i^{DC} \frac{q_j}{\sum_{h \in M_i} q_h}$$

$q_j$  = population of EF $j$

$$f_5 = f_2^{DC} \frac{q_5}{q_5 + q_6}$$

# Popco Bottling Company Example

4. Estimate a nominal transport rate (\$/ton-mi) using the ratio of total distribution cost (\$) to the sum of the product of the demand (ton) at each customer and its distance to its plant (mi).

$$r_{\text{nom}} = \frac{\sum_{i \in N} TDC_i}{\sum_{i \in N} \sum_{j \in M_j} f_j d_{ij}^a}$$

# Popco Bottling Company Example

5. Calculate UFL variable transportation cost  $c_{ij}$  (\$) for each possible NF site  $i$  (all customer and plant locations) and EF site  $j$  (all customer locations) as the product of customer  $j$  demand (ton), distance from site  $i$  to  $j$  (mi), and the nominal transport rate (\$/ton-mi).

$$\mathbf{C} = \left[ c_{ij} \right]_{\substack{i \in M \cup N \\ j \in M}} = \left[ r_{\text{nom}} f_j d_{ij}^a \right]_{\substack{i \in M \cup N \\ j \in M}}$$

6. Solve as UFL, where  $TC$  returned includes all new distribution costs and the fixed portion of production costs.

$$TC = \underbrace{\sum_{i=1}^n k y_i}_{\text{fixed cost}} + \underbrace{\sum_{i=1}^n \sum_{j=1}^m c_{ij} x_{ij}}_{\text{transport cost}}$$

$$n = |M \cup N|, \quad \text{number of potential NF sites}$$

$$m = |M|, \quad \text{number of EF sites}$$

# MILP

$$\begin{aligned} \text{LP: } & \max \mathbf{c}' \mathbf{x} \\ \text{s.t. } & \mathbf{A}\mathbf{x} \leq \mathbf{b} \\ & \mathbf{x} \geq 0 \end{aligned}$$

MILP: some  $x_i$  integer

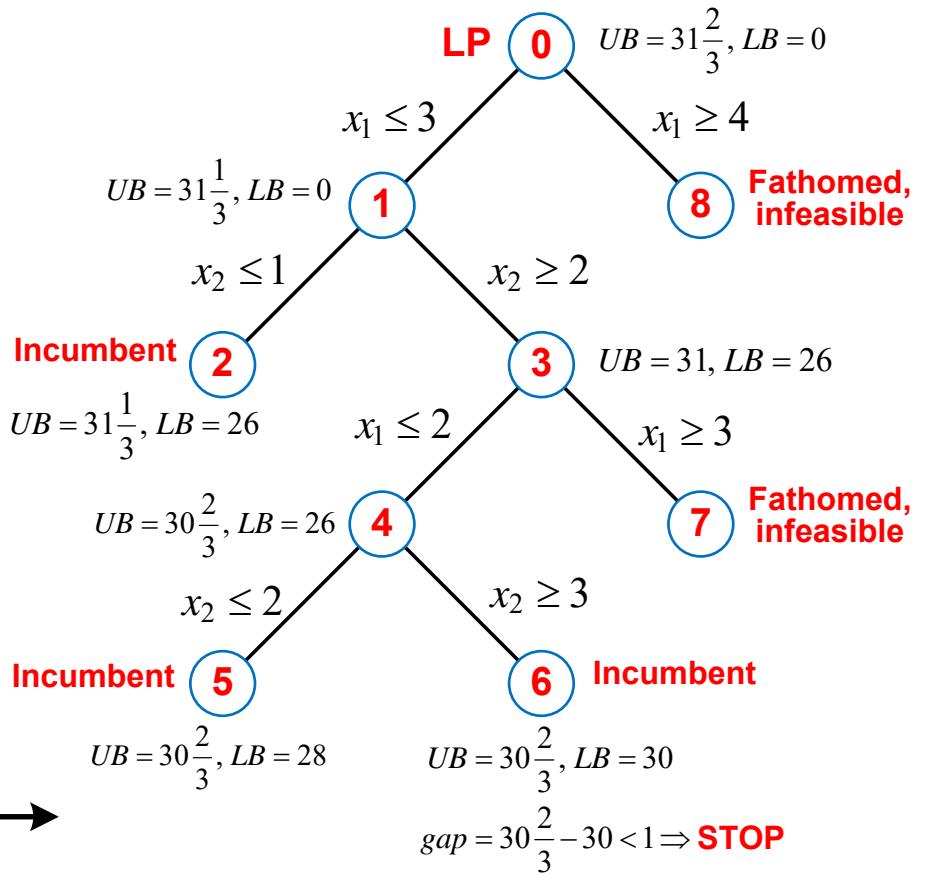
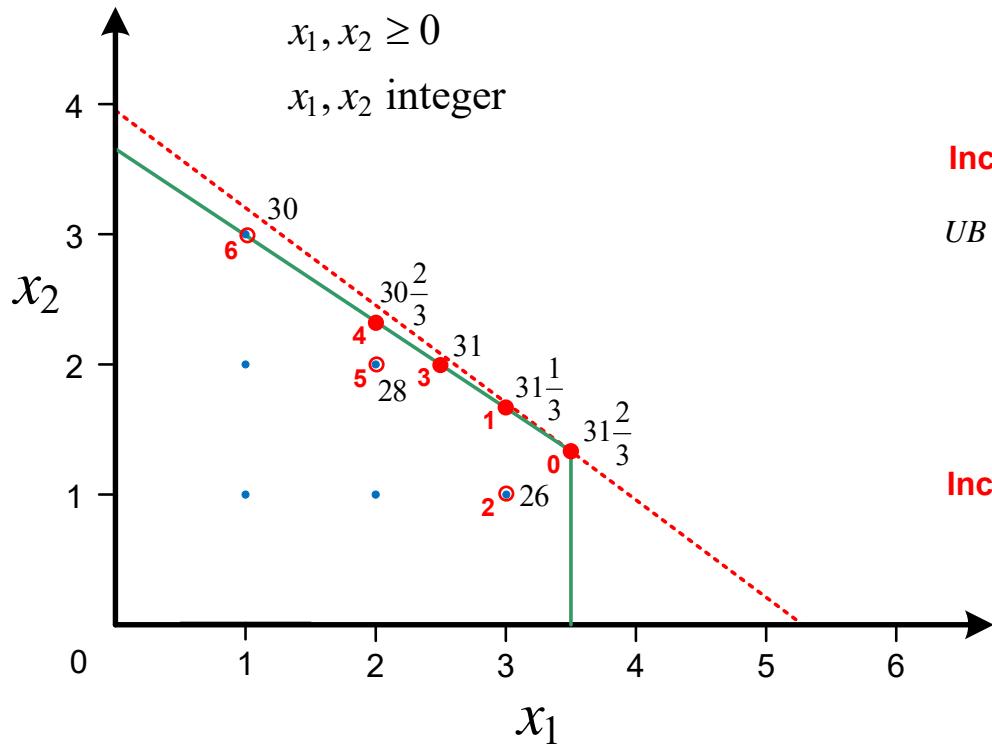
ILP:  $\mathbf{x}$  integer

BLP:  $\mathbf{x} \in \{0,1\}$

# Branch and Bound

$$\begin{aligned}
 \max \quad & 6x_1 + 8x_2 \\
 \text{s.t.} \quad & 2x_1 + 3x_2 \leq 11 \\
 & 2x_1 \leq 7 \\
 & x_1, x_2 \geq 0 \\
 & x_1, x_2 \text{ integer}
 \end{aligned}$$

$\mathbf{c} = [6 \quad 8]$   
 $\mathbf{A} = \begin{bmatrix} 2 & 3 \\ 2 & 0 \end{bmatrix}, \mathbf{b} = \begin{bmatrix} 11 \\ 7 \end{bmatrix}$



# MILP Solvers

LP:  $\max \mathbf{c}'\mathbf{x}$   
 s.t.  $\mathbf{A}\mathbf{x} \leq \mathbf{b}$   
 $\mathbf{x} \geq 0$

intlinprog:  $\min \mathbf{c}$  ( $\max -\mathbf{c}$ )  
 s.t.  $\mathbf{A}_{lt} \leq \mathbf{b}_{lt}$   
 $\mathbf{A}_{eq} = \mathbf{b}_{eq}$

MILP: some  $x_i$  integer

$LB \leq \mathbf{x} \leq UB$

ILP:  $\mathbf{x}$  integer

integer variable indices

BLP:  $\mathbf{x} \in \{0,1\}$

gurobi:  $\mathbf{c}$  (modelsense *min* or *max*)

s.t.  $\mathbf{A} \begin{cases} < \\ = \\ > \end{cases} \mathbf{b}$

$LB \leq \mathbf{x} \leq UB$

variable:  $\begin{cases} C & \text{continuous} \\ B & \text{binary} \\ I & \text{general integer} \end{cases}$

cplex:  $\mathbf{c}$  (sense *min* or *max*)

s.t.  $lhs \leq \mathbf{A} \leq rhs$

$LB \leq \mathbf{x} \leq UB$

variable:  $\begin{cases} C & \text{continuous} \\ B & \text{binary} \\ I & \text{general integer} \end{cases}$

$lhs \quad rhs$

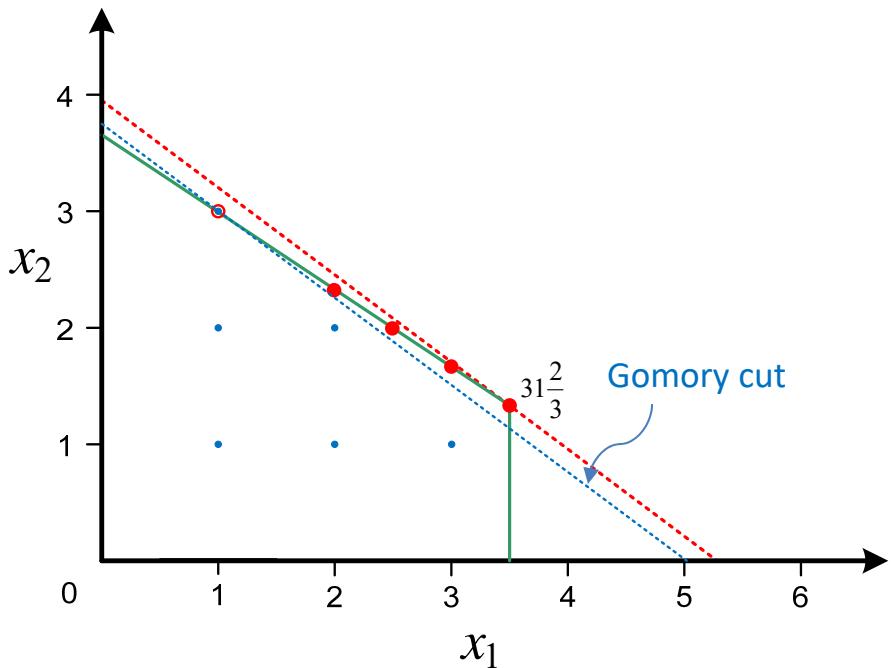
$-\infty \quad \mathbf{b} \quad \Rightarrow \quad \leq$

$\mathbf{b} \quad \infty \quad \Rightarrow \quad \geq$

$\mathbf{b} \quad \mathbf{b} \quad \Rightarrow \quad =$

# MILP Solvers

- **Presolve:** eliminate variables
$$2x_1 + 2x_2 \leq 1, x_1, x_2 \geq 0 \text{ and integer}$$
$$\Rightarrow x_1 = x_2 = 0$$
- **Cutting planes:** keeps all integer solutions and cuts off LP solutions (Gomory cut)
- **Heuristics:** find good initial incumbent solution
- **Parallel:** use separate cores to solve nodes in B&B tree
- **Speedup from 1990-2014:**
  - 320,000× computer speed
  - 580,000× algorithm improvements



# MILP Formulation of UFL

$$\begin{aligned}
 \text{min} \quad & \sum_{i \in N} k_i y_i + \sum_{i \in N} \sum_{j \in M} c_{ij} x_{ij} \\
 \text{s.t.} \quad & \sum_{i \in N} x_{ij} = 1, \quad j \in M \\
 & my_i \geq \sum_{j \in M} x_{ij}, \quad i \in N \\
 & 0 \leq x_{ij} \leq 1, \quad i \in N, j \in M \\
 & y_i \in \{0,1\}, \quad i \in N
 \end{aligned}$$

where

$k_i$  = fixed cost of NF at site  $i \in N = \{1, \dots, n\}$

$c_{ij}$  = variable cost from  $i$  to serve EF  $j \in M = \{1, \dots, m\}$

$y_i = \begin{cases} 1, & \text{if NF established at site } i \\ 0, & \text{otherwise} \end{cases}$

$x_{ij}$  = fraction of EF  $j$  demand served from NF at site  $i$ .

```

%% UFL MILP Matlab code, given k and C
mp.addobj('min', k, C)
for j = M
    mp.addcstr(0, {':', j}, '=', 1)
end
for i = N
    mp.addcstr({m, {i}}, '>=', {i, ':'})
end
mp.addub(1, 1)
mp.addctype('B', 'C')

```

$$y_i \geq x_{ij}, \quad i \in N, j \in M$$

# Set Covering

$M = \{1, \dots, m\}$ , objects to be covered

$M_i \subseteq M, i \in N = \{1, \dots, n\}$ , subsets of  $M$

$c_i$  = cost of using  $M_i$  in cover

$I^* = \arg \min_I \left\{ \sum_{i \in I} c_i : \bigcup_{i \in I} M_i = M \right\}$ , min cost covering of  $M$

$$\min \sum_{i \in N} c_i x_i$$

$$\text{s.t. } \sum_{i \in N} a_{ji} x_i \geq 1, \quad j \in M$$

$$x_i \in \{0, 1\}, \quad i \in N$$

```
%% Set Covering BLP Matlab code,  
% given c and A  
mp = Milp('Set Cover')  
mp.addobj('min', c)  
mp.addcstr(A, '>=' , 1)  
mp.addctype('B')
```

where

$$x_i = \begin{cases} 1, & \text{if } M_i \text{ is in cover} \\ 0, & \text{otherwise} \end{cases}$$

$$a_{ji} = \begin{cases} 1, & \text{if } j \in M_i \\ 0, & \text{otherwise.} \end{cases}$$

# Bin Packing

$M = \{1, \dots, m\}$ , objects to be packed

$v_j$  = volume of object  $j$

$V$  = volume of each bin  $B_i$  ( $\max v_j \leq V$ )

$B^* = \arg \min_B \left\{ |B| : \sum_{j \in B_i} v_j \leq V, \bigcup_{B_i \in B} B_i = M \right\}, \text{ min packing of } M$

$$\min \quad \sum_{i \in M} y_i$$

$$\text{s.t.} \quad Vy_i \geq \sum_{j \in M} v_j x_{ij}, \quad i \in M$$

$$\sum_{i \in M} x_{ij} = 1, \quad j \in M$$

$$y_i \in \{0, 1\}, \quad i \in M$$

$$x_{ij} \in \{0, 1\}, \quad i \in M, j \in M$$

where

$$y_i = \begin{cases} 1, & \text{if bin } B_i \text{ is used in packing} \\ 0, & \text{otherwise} \end{cases}$$

$$x_{ij} = \begin{cases} 1, & \text{if object } j \text{ packed into bin } B_i \\ 0, & \text{otherwise.} \end{cases}$$

```
%% Bin Packing BLP Matlab code,
% given v and V
mp = Milp('Bin Packing')
mp.addobj('min', ones(1, m), zeros(m))
for i = M
    mp.addcstr({v, {i}}, '>=', {v, {i}, ':'})
end
for j = M
    mp.addcstr(0, {':', j}, '=', 1)
end
mp.addctype('B', 'B')
```

# Topics

1. Introduction
2. Facility location
- 3. Freight transport**
  - Midterm exam
4. Network models
5. Routing
6. Warehousing
  - Final project
  - Final exam

# Logistics Engineering Design Constants

1. Circuit Factor: **1.2** ( $g$ )
  - $1.2 \times \text{GC distance} \approx \text{actual road distance}$
2. Local vs. Intercity Transport:
  - Local: < **50 mi**  $\Rightarrow$  use actual road distances
  - Intercity: > 50 mi  $\Rightarrow$  can estimate road distances
    - 50-250 mi  $\Rightarrow$  return possible (11 HOS)
    - > 250 mi  $\Rightarrow$  always one-way transport
    - > 500-750 mi  $\Rightarrow$  intermodal rail possible
3. Inventory Carrying Cost ( $h$ ) = funds + storage + obsolescence
  - **20%** average (no product information)
  - 10% low-value product (construction)
  - 25-30% general durable manufactured goods
  - 50% computer equipment
  - >> 100% perishable goods (produce)

# Logistics Engineering Design Constants

4.  $\frac{\text{Value}}{\text{Transport Cost}} \gg 1: \$1 \text{ ft}^3 \approx \frac{\$2,620 \text{ Shanghai-LA/LB shipping cost}}{2,400 \text{ ft}^3 40' \text{ ISO container capacity}}$

5. TL Weight Capacity: **25 tons** ( $K_{wt}$ )

- (40 ton max per regulation) –  
(15 ton tare for tractor-trailer)  
= 25 ton max payload
- Weight capacity = 100% of physical capacity



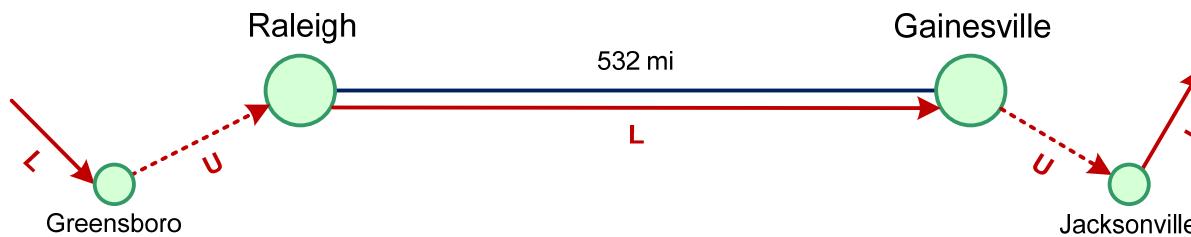
6. TL Cube Capacity: **2,750 ft<sup>3</sup>** ( $K_{cu}$ )

- Trailer physical capacity = 3,332 ft<sup>3</sup>
- Effective capacity =  
 $3,332 \times 0.80 \approx 2,750 \text{ ft}^3$
- Cube capacity = 80% of physical capacity



# Logistics Engineering Design Constants

7. TL Revenue per Loaded Truck-Mile: \$2/mi in 2004 (  $r$  )
  - TL revenue for the carrier is your TL cost as a shipper



15%, average deadhead travel

\$1.60, cost per mile in 2004

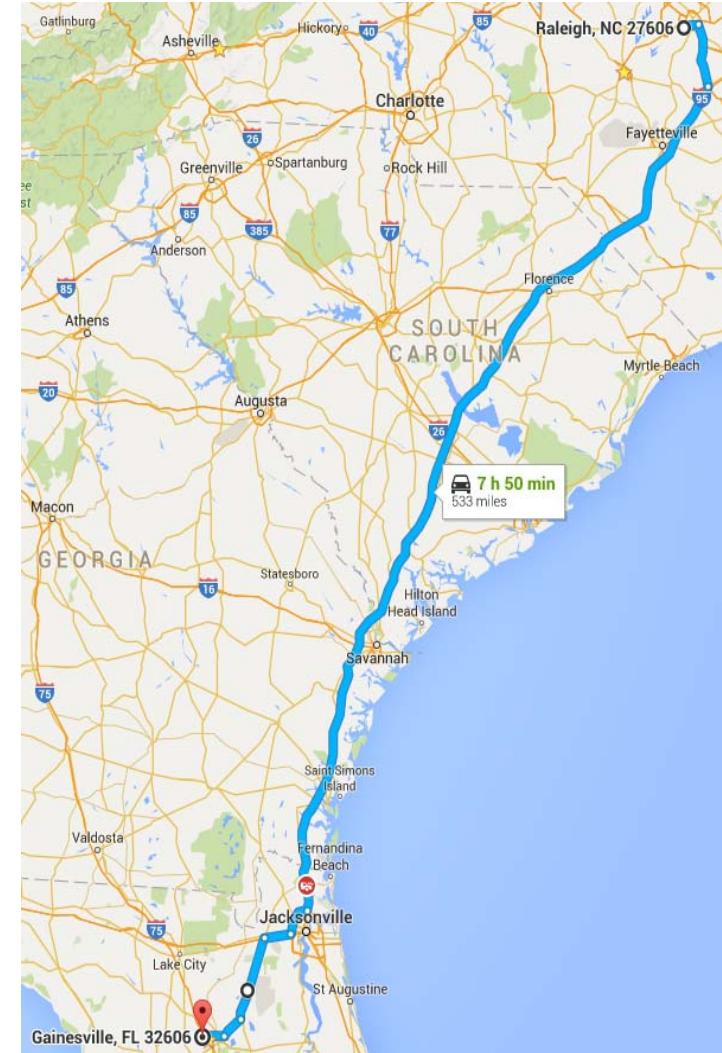
$$\frac{\$1.60}{1 - 0.15} = \$1.88, \text{ cost per loaded-mile}$$

6.35%, average operating margin for trucking

$$\frac{\$1.88}{1 - 0.0635} \approx \$2.00, \text{ revenue per loaded-mile}$$

# Truck Shipment Example

- Product is to be shipped in cartons from Raleigh, NC (27606) to Gainesville, FL (32606). Each unit weighs 40 lb and occupies 9 ft<sup>3</sup>, and units can be stacked on top of each other in a trailer.
- **One-Time Shipments** (operational decision):
  - Know when and how much to ship, need to determine if TL and/or LTL to be used
  - Need to know much it will cost
- **Periodic Shipments** (tactical decision):
  - Need to determine how often and how much to ship



# Truck Shipment Example: One-Time

- Assuming that the product is to be shipped P2P TL, what is the maximum payload for each trailer used for the shipment?

$$q_{\max}^{wt} = K_{wt} = 25 \text{ ton}$$

$$K_{cu} = 2750 \text{ ft}^3$$

$$s = \frac{40 \text{ lb/unit}}{9 \text{ ft}^3/\text{unit}} = 4.4444 \text{ lb/ft}^3$$

$$K_{cu} = \frac{q_{\max}^{cu}}{\left(\frac{s}{2000}\right)} \Rightarrow q_{\max}^{cu} = \frac{sK_{cu}}{2000}$$

$$q_{\max} = \min \left\{ q_{\max}^{wt}, q_{\max}^{cu} \right\} = \min \left\{ K_{wt}, \frac{sK_{cu}}{2000} \right\}$$

$$= \min \left\{ 25, \frac{4.4444(2750)}{2000} \right\} = 6.1111 \text{ ton}$$

# Truck Shipment Example: One-Time

2. Next Monday, 350 units of the product are to be shipped.  
How many truckloads are required for this shipment?

$$q = 350 \frac{40}{2000} = 7 \text{ ton}, \quad \left\lceil \frac{q}{q_{\max}} \right\rceil = \left\lceil \frac{7}{6.1111} \right\rceil = 2 \text{ truckloads}$$

3. Using the most recent rate estimate available, what is the TL transport charge for this shipment?
  - Ratio of most recent Producer Price Index (PPI) for TL (see course website for link) to PPI in 2004 used to inflate known \$2/mi 2004 value

$$d = 532 \text{ mi}$$

$$r = \frac{PPI_{TL}^{\text{Jan 2017}}}{PPI_{TL}^{2004}} \times r_{2004} = \frac{124.3}{102.7} \times \$2.00 / \text{mi} = \$2.4206 / \text{mi}$$

$$c_{TL} = \left\lceil \frac{q}{q_{\max}} \right\rceil r d = \left\lceil \frac{7}{6.1111} \right\rceil (2.4206)(532) = \$2,575.56$$

# Truck Shipment Example: One-Time

4. Using the most recent LTL rate estimate, what is transport charge to ship the fractional portion of the shipment LTL (i.e., the last partially full truckload portion)?

$$q_{\text{frac}} = q - q_{\text{max}} = 7 - 6.1111 = 0.8889 \text{ ton}$$

$$\begin{aligned} r_{LTL} &= PPI_{LTL}^{\text{Jan 2017}} \left[ \frac{\frac{s^2}{8} + 14}{\left( q_{\text{frac}}^{\frac{1}{7}} d^{\frac{15}{29}} - \frac{7}{2} \right) (s^2 + 2s + 14)} \right] \\ &= 168.2 \left[ \frac{\frac{4.49^2}{8} + 14}{\left( 7^{\frac{1}{7}} 532^{\frac{15}{29}} - \frac{7}{2} \right) (4.49^2 + 2(4.49) + 14)} \right] = \$2.9837 / \text{ton-mi} \end{aligned}$$

$$c_{LTL} = r_{LTL} q_{\text{frac}} d = 2.9837(0.8889)(532) = \$1,410.95$$

# Truck Shipment Example: One-Time

5. What is the change in total charge associated with the combining TL and LTL as compared to just using TL?

$$\begin{aligned}\Delta c &= c_{TL} - \left( \frac{c_{TL}}{2} + c_{LTL} \right) \\ &= \left\lceil \frac{q}{q_{\max}} \right\rceil r d - \left( \left\lfloor \frac{q}{q_{\max}} \right\rfloor r d + r_{LTL} q_{\text{frac}} d \right) \\ &= 2,575.56 - (1,287.78 + 1,410.95) \\ &= 2,575.56 - 2,698.73 = -\$123.17\end{aligned}$$

# Truck Shipment Example: One-Time

6. What would the fractional portion have to be so that the TL and LTL chargers are equal?

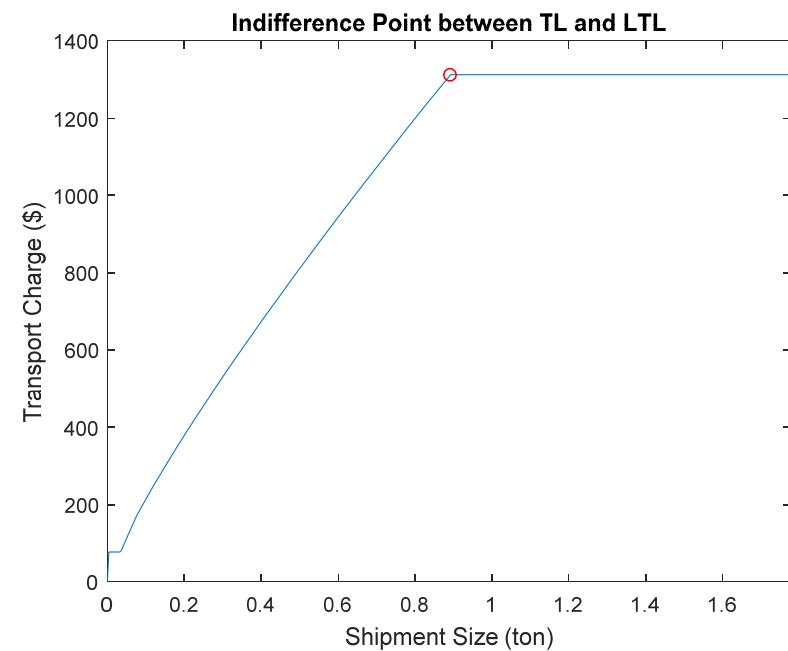
$$c_{TL}(q) = \left\lceil \frac{q}{q_{\max}} \right\rceil r d$$

$$r_{LTL}(q) = PPI_{LTL} \left[ \frac{\frac{s^2}{8} + 14}{\left( q^{\frac{1}{7}} d^{\frac{15}{29}} - \frac{7}{2} \right) (s^2 + 2s + 14)} \right]$$

$$c_{LTL}(q) = r_{LTL}(q) q d$$

$$q_I = \arg \min_q \left( \| c_{TL}(q) - c_{LTL}(q) \| \right)$$

$$= 0.7967 \text{ ton}$$

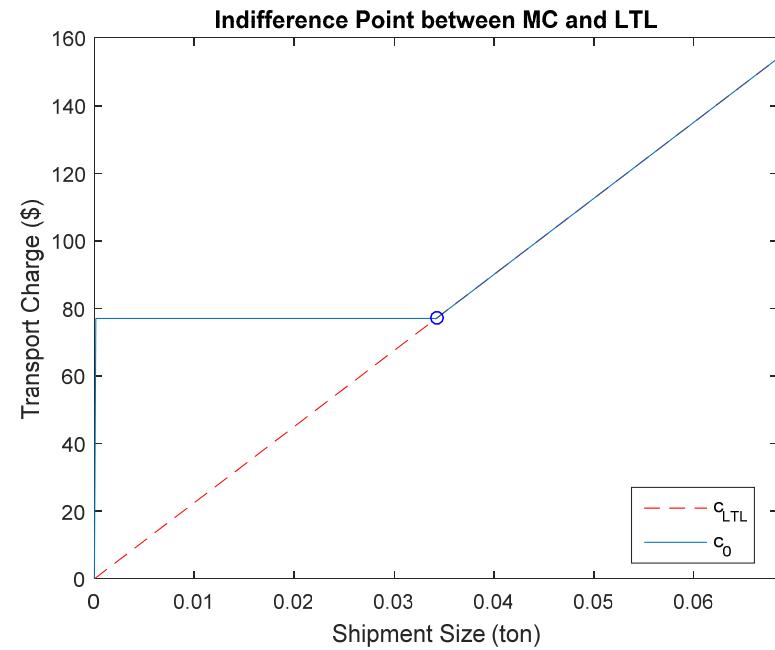


# Truck Shipment Example: One-Time

7. What are the TL and LTL minimum charges?

$$MC_{TL} = \left( \frac{r}{2} \right) 45 = \left( \frac{2.4206}{2} \right) 45 = \$54.46$$

$$MC_{LTL} = \left( \frac{PPI_{LTL}}{104.2} \right) \left( 45 + \frac{d^{\frac{28}{19}}}{1625} \right)$$
$$= \left( \frac{168.2}{104.2} \right) \left( 45 + \frac{532^{\frac{28}{19}}}{1625} \right) = \$82.97$$

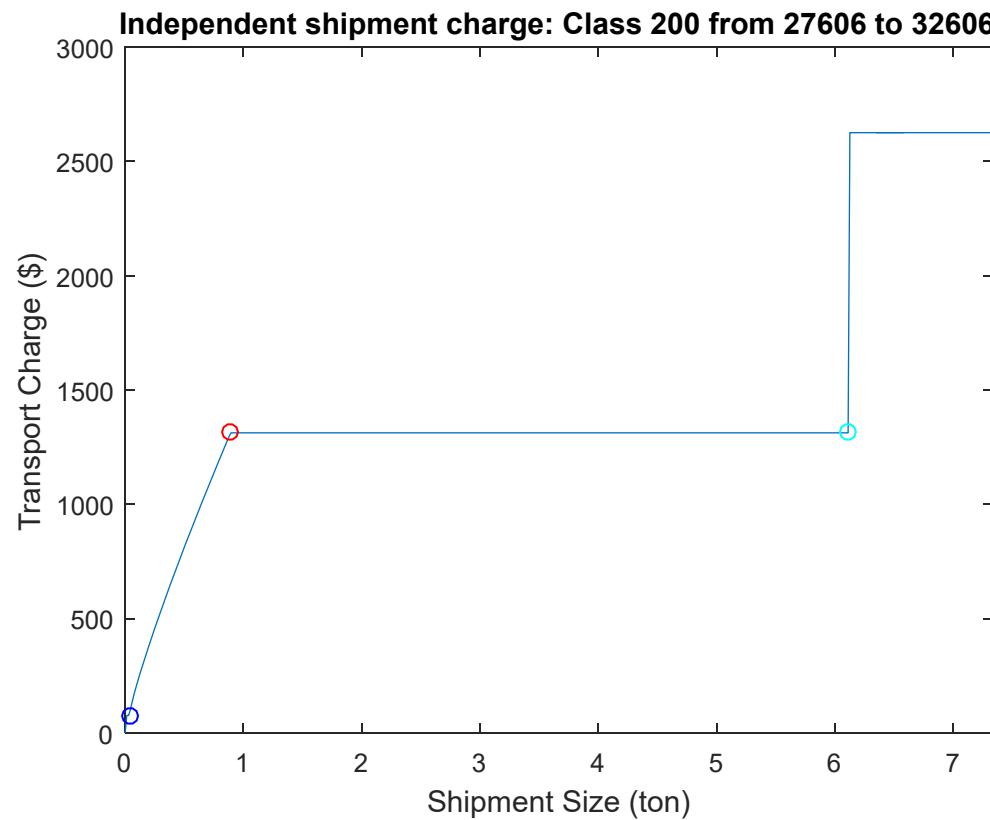


- Why do these charges not depend on the size of the shipment?
- Why does only the LTL minimum charge depend on the distance of the shipment?

# Truck Shipment Example: One-Time

- Independent Transport Charge (\$):

$$c_0(q) = \min \left\{ \max \left\{ c_{TL}(q), MC_{TL} \right\}, \max \left\{ c_{LTL}(q), MC_{LTL} \right\} \right\}$$



# Truck Shipment Example: One-Time

8. Using the same LTL shipment, find online one-time (spot) LTL rate quotes using the FedEx LTL website

$$q_{\text{frac}} = 0.8889 \text{ ton} \Rightarrow 0.8889(2000) = 1778 \text{ lb}$$

$$\Rightarrow \frac{1778}{40} \approx 44 \text{ cartons}$$

- Most likely freight class:

$$s = \frac{40 \text{ lb/unit}}{9 \text{ ft}^3/\text{unit}} \\ = 4.4444 \text{ lb/ft}^3$$

$\Rightarrow$  Class 200

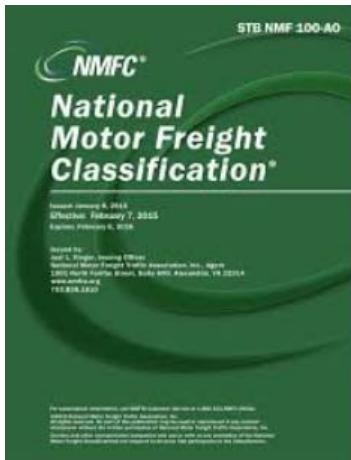
- What is the rate quote for the reverse trip from Gainesville (32606) to Raleigh (27606)?

**Class-Density Relationship**

Class	Load Density (lb/ft <sup>3</sup> )		Max Physical Weight (tons)	Max Effective Cube (ft <sup>3</sup> )
	Minimum	Average		
500	—	0.52	0.72	2,750
400	1	1.49	2.06	2,750
300	2	2.49	3.43	2,750
250	3	3.49	4.80	2,750
200	4	4.49	6.17	2,750
175	5	5.49	7.55	2,750
150	6	6.49	8.92	2,750
125	7	7.49	10.30	2,750
110	8	8.49	11.67	2,750
100	9	9.72	13.37	2,750
92.5	10.5	11.22	15.43	2,750
85	12	12.72	17.49	2,750
77.5	13.5	14.22	19.55	2,750
70	15	18.01	24.76	2,750
65	22.5	25.50	25	1,961
60	30	32.16	25	1,555
55	35	39.68	25	1,260
50	50	56.18	25	890

# Truck Shipment Example: One-Time

- The *National Motor Freight Classification* (NMFC) can be used to determine the product class
- Based on:
  - Load density
  - Special handling
  - Stowability
  - Liability



Item	Description	Class	NMFC	Sub
Abietic Acid	Abietic Acid, in drums	55	42605	-
Accordions	Accordions, in boxes	125	138820	-
Acetonitrile	Acetonitrile, in boxes or drums. See item 60000 for class dependent upon released value	85	42645	-
Acetylene	in steel cylinders	70	85520	-
Acid Fish Scrap	Fish Scrap, NOI, dry, not ground, pulverized nor screened, or Acid Fish Scrap, in bags	77.5	69980	-
Aircraft Parts	metal, struts, skins, panels	200	11790	01
Aluminum Channel	U channel	60	13340	-
Aluminum Table Set	aluminum table SU	200	82105	01
Ambulance Stretcher	stretcher	200	56920	06
Arches Support	Iron Steel	60	52460	-
Architectural Details	6 - 8 lbs per cubic foot	125	56290	05
Architectural Details	2 - 4 lbs per cubic ft	250	56290	03
Assembled Furniture	Bathroom cabinet set up	300	39220	01
Assembled Furniture	Highboys, dressers, wooden set up	125	80120	01
Assembled Furniture	Wood furniture 4-6 Lbs per cu ft	150	82270	04
Assembled Furniture	Chairs wooden setup w/out upholstery	300	80770	01
Assembled Furniture	Chairs wooden setup w/out upholstery KD	125	80770	03
Assembled Furniture	Couch w/ back & arms put together	175	80865	03
Assembled Furniture	Chairs put together w/ upholstery	200	79255	01
Assembled Furniture	Metal cabinets in boxes	110	39270	06
Assembled Furniture	18 gauge steel cabinet	70	39340	-
Assembled Furniture	Benches, cabinets, tables for workstations	125	23410	-
Assembled Furniture	Buffets, china cabinets put together	125	80080	-
Assembled Furniture	Cabinets of metal or plastic for storage	92.5	39235	-
Assembled Furniture	Tanning bed	150	109050	-
Assembled Furniture	Mattresses, in packages or boxes	200	79550	-
Athletic / Sporting Goods	Gym equipment, playground, sports items. Density Item			
Attachments: Backhoe	NOI: Attachments, backhoe (Backhoes), tractor or truck, on lift truck skids or pallets:	175	114217	01
Attachments: Backhoe	Attachments, backhoe (Backhoes), tractor or truck, on lift truck skids or pallets: Each shipped with all components secured to a single pallet, platform or skid, weighing 1100 pounds or more and having a density of 8 pounds or greater per cubic foot	100	114217	02

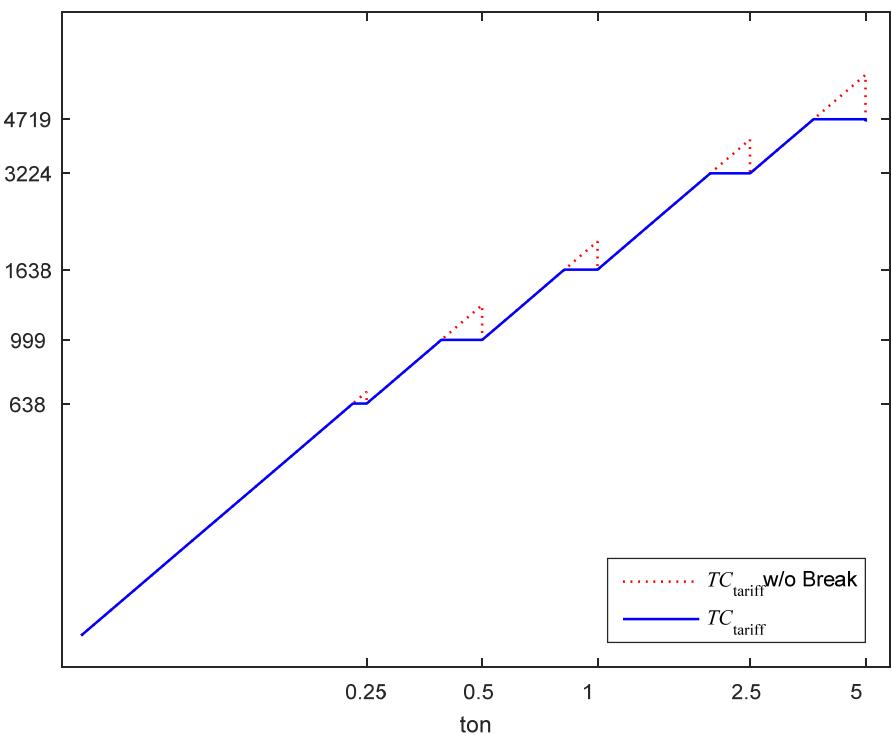
# Truck Shipment Example: One-Time

- CzarLite tariff table for O-D pair 27606-32606

$$cwt = \text{hundredweight} = 100 \text{ lb} = \frac{100}{2000} = \frac{1}{20} \text{ ton}$$

**Tariff (in \$/cwt) from Raleigh, NC (27606) to Gainesville, FL (32606)  
(532 mi, CzarLite DEMOCZ02 04-01-2000, minimum charge = \$95.23)**

Freight Class	Rate Breaks ( <i>i</i> )								
	1	2	3	4	5	6	7	8	9&10
500	341.42	314.14	245.80	201.48	158.60	112.37	55.66	55.66	55.66
400	273.88	251.99	197.19	161.61	127.22	91.12	45.10	45.10	45.10
300	206.34	189.85	148.56	121.76	95.85	69.47	34.43	34.43	34.43
250	172.56	158.77	124.23	101.83	80.15	58.03	28.79	28.79	28.79
200	138.78	127.69	99.92	81.89	64.47	47.19	23.40	23.40	23.40
175	121.37	111.68	87.39	71.62	56.38	41.27	20.39	20.39	20.39
150	104.49	96.13	75.22	61.66	48.53	35.96	17.75	17.75	17.75
125	87.59	80.60	63.07	51.69	40.69	30.24	15.00	15.00	15.00
110	77.57	71.37	55.85	45.77	36.04	28.61	14.40	14.40	14.40
100	71.23	65.55	51.29	42.04	33.09	27.58	14.03	10.80	9.90
92	66.48	61.18	47.88	39.24	30.89	25.75	13.68	10.52	9.66
85	61.74	56.80	44.45	36.43	28.68	23.91	13.20	10.15	9.32
77	56.99	52.44	41.04	33.63	26.48	22.07	12.60	9.68	8.89
70	52.77	48.55	37.99	31.14	24.51	20.43	12.00	9.23	8.47
65	50.07	46.08	36.05	29.56	23.04	19.39	11.87	9.14	8.39
60	47.44	43.64	34.15	28.00	21.82	18.37	11.76	9.04	8.30
55	44.75	41.17	32.22	26.40	20.59	17.32	11.64	8.96	8.22
50	41.57	38.26	29.94	24.54	19.12	16.10	11.52	8.85	8.14
Tons ( $q_i^b$ )	0.25	0.5	1	2.5	5	10	15	20	$\infty$



# Truck Shipment Example: One-Time

9. Using the same LTL shipment, what is the transport cost found using the undiscounted CzarLite tariff?

$q = 0.8889, \text{ class} = 200$	Freight Class	Rate Breaks ( $i$ )								
		1	2	3	4	5	6	7	8	9&10
$disc = 0, MC = 95.23$	500	341.42	314.14	245.80	201.48	158.60	112.37	55.66	55.66	55.66
	400	273.88	251.99	197.19	161.61	127.22	91.12	45.10	45.10	45.10
	300	206.34	189.85	148.56	121.76	95.85	69.47	34.43	34.43	34.43
	250	172.56	158.77	124.23	101.83	80.15	58.03	28.79	28.79	28.79
	200	138.78	127.69	99.92	81.89	64.47	47.19	23.40	23.40	23.40
	175	121.37	111.68	87.39	71.62	56.38	41.27	20.39	20.39	20.39
$i = \arg \left\{ q_i^B \mid q_{i-1}^B \leq q < q_i^B \right\}$	50	41.57	38.26	29.94	24.54	19.12	16.10	11.52	8.85	8.14
$= \arg \left\{ q_3^B \mid q_2^B \leq q < q_3^B \right\}$	Tons ( $q_i^B$ )	0.25	0.5	1	2.5	5	10	15	20	$\infty$

$$= \arg \left\{ q_3^B \mid 0.5 \leq 0.8889 < 1 \right\} = 3$$

$$\begin{aligned}
 c_{\text{tariff}} &= (1 - disc) \max \left\{ MC, \min \left\{ OD(\text{class}, i) 20q, OD(\text{class}, i+1) 20q_i^B \right\} \right\} \\
 &= (1 - 0) \max \left\{ 95.23, \min \left\{ OD(200, 3) 20(0.8889), OD(200, 4) 20(1) \right\} \right\} \\
 &= \max \left\{ 95.23, \min \left\{ (99.92) 20(0.8889), (81.89) 20(1) \right\} \right\} \\
 &= \max \left\{ 95.23, \min \left\{ 1,776.23, 1,637.80 \right\} \right\} = \$1,637.80
 \end{aligned}$$

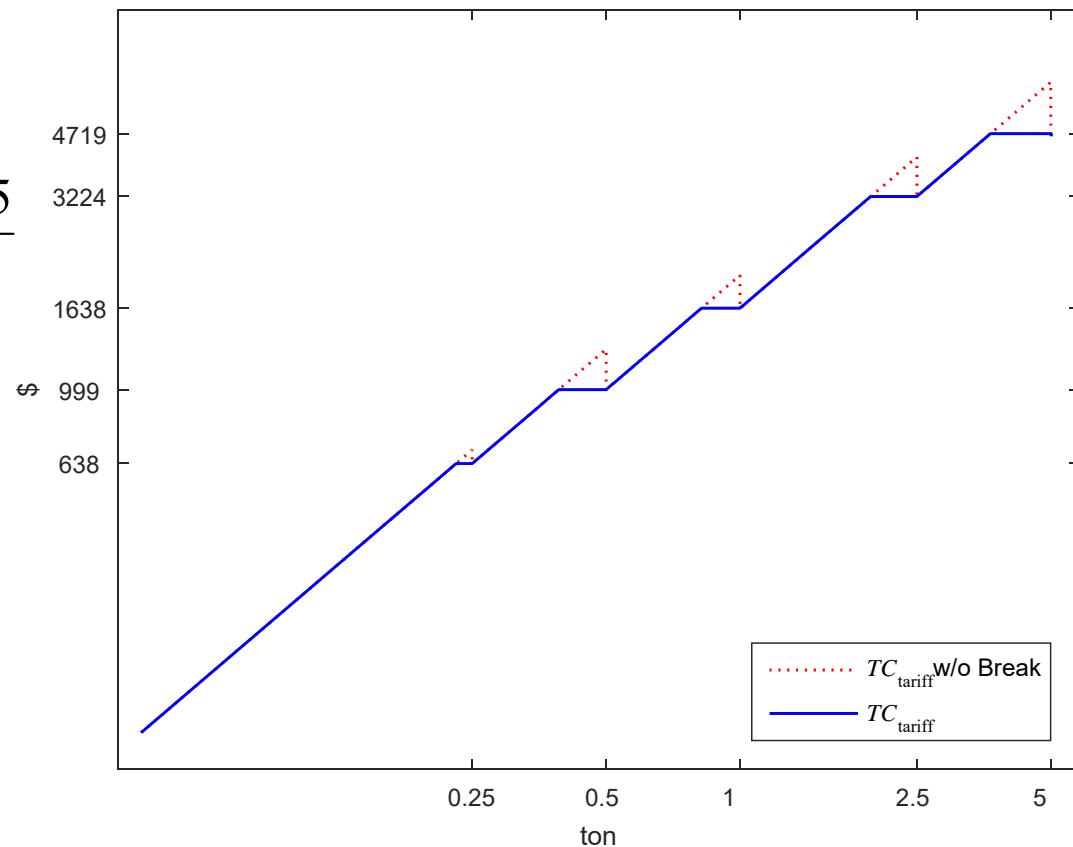
# Truck Shipment Example: One-Time

10. What is the implied discount of the estimated charge from the CzarLite tariff cost?

$$\begin{aligned}disc &= \frac{c_{\text{tariff}} - c_{LTL}}{c_{\text{tariff}}} \\&= \frac{1,637.80 - 1,410.95}{1,637.80} \\&= 14\%\end{aligned}$$

- What is the weight break between the rate breaks?

$$\begin{aligned}q_i^W &= \frac{OD(\text{class}, i+1)}{OD(\text{class}, i)} q_i^B \\&= \frac{81.89}{99.92}(1) = 0.8196 \text{ ton}\end{aligned}$$



# Truck Shipment Example: Periodic

11. Continuing with the example: assuming a constant annual demand for the product of 20 tons, what is the number of full truckloads per year?

$$f = 20 \text{ ton/yr}$$

$$q = q_{\max} = 6.1111 \text{ ton/ TL} \quad (\text{full truckload} \Rightarrow q \equiv q_{\max})$$

$$n = \frac{f}{q} = \frac{20}{6.1111} = 3.2727 \text{ TL/yr}, \quad \text{average shipment frequency}$$

- Why should this number not be rounded to an integer value?

# Truck Shipment Example: Periodic

12. What is the shipment interval?

$$t = \frac{1}{n} = \frac{q}{f} = \frac{6.1111}{20} = 0.3056 \text{ yr/TL}, \text{ average shipment interval}$$

- How many days are there between shipments?

365.25 day/yr

$$t \times 365.25 = \frac{365.25}{n} = 111.6042 \text{ day/TL}$$

# Truck Shipment Example: Periodic

13. What is the annual full-truckload transport cost?

$$d = 532 \text{ mi}, \quad r = \$2.4206 / \text{mi}$$

$$r_{FTL} = \frac{r}{q_{\max}} = \frac{2.4206}{6.1111} = \$0.3961 / \text{ton-mi}$$

$$\begin{aligned} TC_{FTL} &= f r_{FTL} d = n r d \quad (= w d, w = \text{monetary weight in } \$/\text{mi}) \\ &= 3.2727(2.4206)532 = \$4,214.56/\text{yr} \end{aligned}$$

- What would be the cost if the shipments were to be made at least every three months?

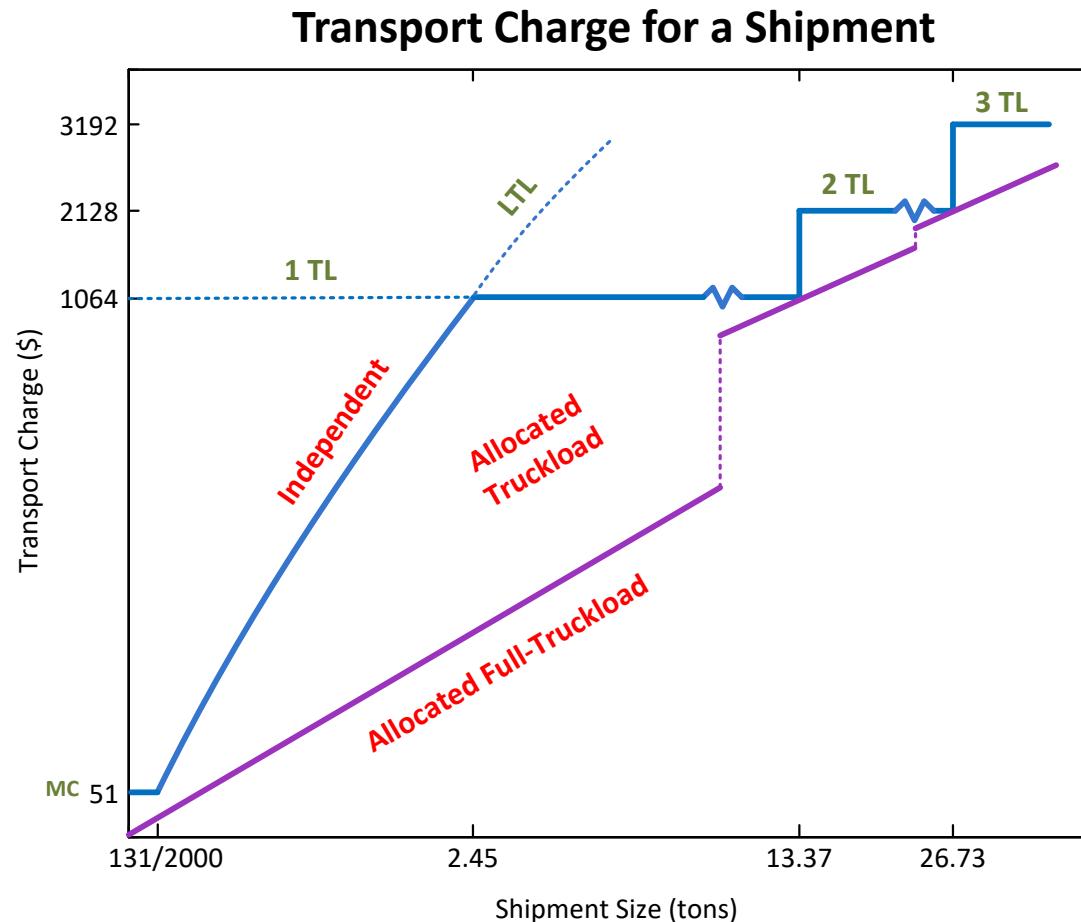
$$t_{\max} = \frac{3}{12} \text{ yr/TL} \quad \Rightarrow \quad n_{\min} = \frac{1}{t_{\max}} = 4 \text{ TL/yr}$$

$$\begin{aligned} TC'_{FTL} &= \max \{n, n_{\min}\} r d \\ &= \max \{3.2727, 4\} 2.4206(532) = \$5,151.13/\text{yr} \end{aligned}$$

# Truck Shipment Example: Periodic

- Independent and allocated full-truckload charges:

$$q \leq q_{\max} \Rightarrow [c_{UB}, c_{LB}] = [c_0(q), qr_{FTL} d]$$



# Truck Shipment Example: Periodic

- *Total Logistics Cost* (TLC) includes all costs that could change as a result of a logistics-related decision

$$TLC = TC + IC + PC$$

$TC$  = transport cost

$IC$  = inventory cost

$$= IC_{\text{cycle}} + IC_{\text{pipeline}} + IC_{\text{safety}}$$

$PC$  = purchase cost

- *Cycle inventory*: held to allow cheaper large shipments
- *Pipeline inventory*: goods in transit or awaiting transshipment
- *Safety stock*: held due to transport uncertainty
- *Purchase cost*: can be different for different suppliers

# Truck Shipment Example: Periodic

14. Since demand is constant throughout the year, one half of a shipment is stored at the customer, on average. Assuming that the production rate is also constant, one half of a shipment will also be stored at the supplier, on average. Assuming each ton of the product is valued at \$25,000, what is a “reasonable estimate” for the total annual cost for this cycle inventory?

$$\begin{aligned} IC_{\text{cycle}} &= (\text{annual cost of holding one ton})(\text{average annual inventory level}) \\ &= (vh)(\alpha q) \end{aligned}$$

$v$  = unit value of shipment (\$/ton)

$h$  = inventory carrying rate, the cost per dollar of inventory per year (1/yr)

$\alpha$  = average inter-shipment inventory fraction at Origin and Destination

$q$  = shipment size (ton)

# Truck Shipment Example: Periodic

Rate ( $h$ ) = interest + warehousing + obsolescence rate

- Interest: 4% real interest rate (prime + 2% – inflation rate)
- Warehousing: 6% per Total U.S. Logistics Costs
- Obsolescence: default rate  $h_{\text{annual}} = 0.3 \Rightarrow h_{\text{obsol}} = 0.2$

– Estimate  $h_{\text{obsol}}$  using “fifth-of-value interval” method:  
given  $t$  when product worth 20% of its original value  $v$ , find  $h_{\text{obsol}}$

$$th_{\text{obsol}}v = \frac{4}{5}v \Rightarrow h_{\text{obsol}} = \frac{4v}{5tv} = \frac{4}{5t}, \quad t = \frac{4}{5h_{\text{obsol}}}$$

– Example: If a product loses 80% of its value after 5 years:

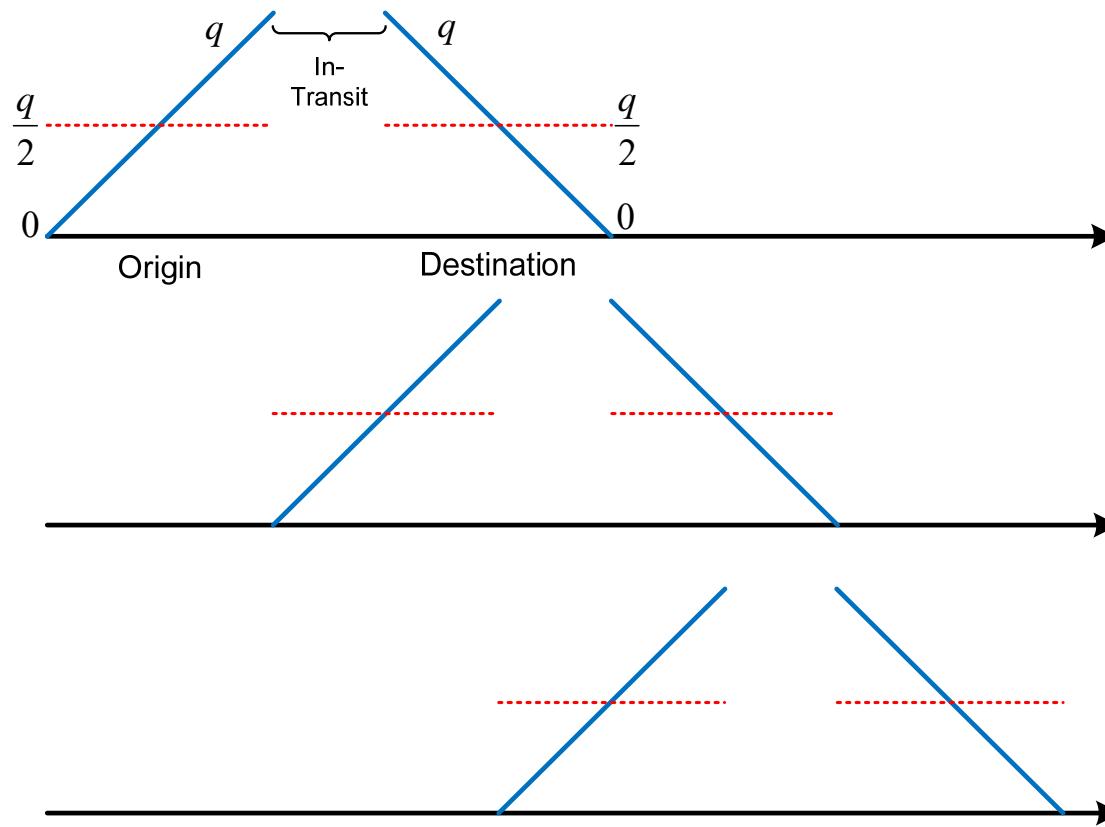
$$t = 5 \Rightarrow h_{\text{obsol}} = \frac{4}{5t} = \frac{4}{5(5)} = 0.16 \quad (16\%)$$

– Example: If a product loses 80% of its value after two weeks:

$$t = \frac{14}{365.25} \Rightarrow h_{\text{obsol}} = \frac{4}{5t} = \frac{4}{5(0.0383)} = 20.87 \quad (2,087\%)$$

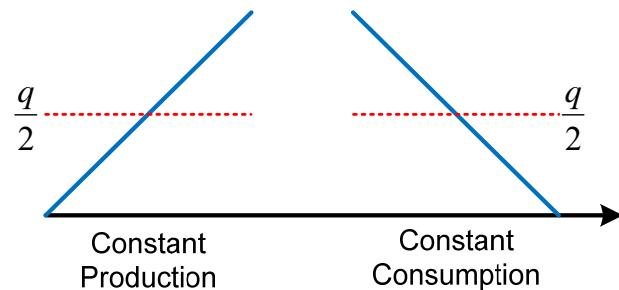
# Truck Shipment Example: Periodic

- Average annual inventory level  $= \frac{q}{2} + \frac{q}{2} = \left(\frac{1}{2} + \frac{1}{2}\right)q = (1)q \Rightarrow \alpha = 1$

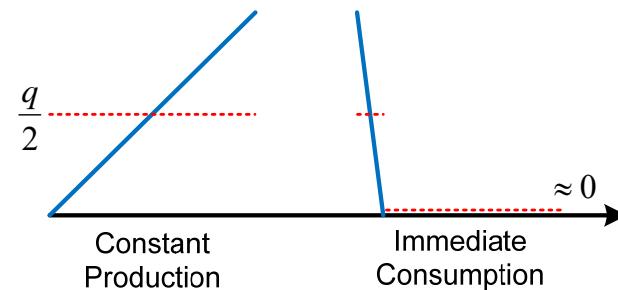


# Truck Shipment Example: Periodic

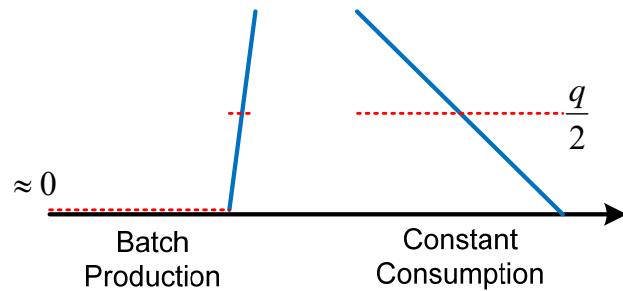
- Inter-shipment inventory fraction alternatives:  $\alpha = \alpha_O + \alpha_D$



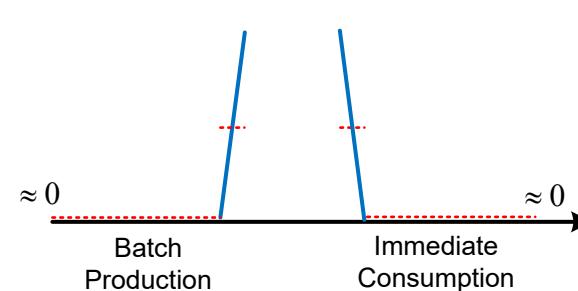
$$\alpha = \frac{1}{2} + \frac{1}{2} = 1$$



$$\alpha = \frac{1}{2} + 0 = \frac{1}{2}$$



$$\alpha = 0 + \frac{1}{2} = \frac{1}{2}$$



$$\alpha = 0 + 0 = 0$$

# Truck Shipment Example: Periodic

- “Reasonable estimate” for the total annual cost for the cycle inventory:

$$\begin{aligned} IC_{\text{cycle}} &= \alpha v h q \\ &= (1)(25,000)(0.3)6.1111 \\ &= \$45,833.33 / \text{yr} \end{aligned}$$

where

$$\alpha = \frac{1}{2} \text{ at Origin} + \frac{1}{2} \text{ at Destination} = 1$$

$$v = \$25,000 = \text{unit value of shipment (\$/ton)}$$

$$h = 0.3 = \text{estimated carrying rate for manufactured products (1/yr)}$$

$$q = q_{\max} = 6.111 = \text{FTL shipment size (ton)}$$

# Truck Shipment Example: Periodic

15. What is the annual total logistics cost (TLC) for these full-truckload TL shipments?

$$\begin{aligned} TLC_{FTL} &= TC_{FTL} + IC_{cycle} \\ &= n rd + \alpha v h q \\ &= 3.2727(2.4206)532 + (1)(25,000)(0.3)6.1111 \\ &= 4,214.56 + 45,833.33 \\ &= \$50,047.89 / yr \end{aligned}$$

# Truck Shipment Example: Periodic

16. What is minimum possible annual total logistics cost for TL shipments, where the shipment size can now be less than a full truckload?

$$TLC_{TL}(q) = TC_{TL}(q) + IC(q) = \frac{f}{q} c_{TL}(q) + \alpha vhq = \frac{f}{q} rd + \alpha vhq$$

$$\frac{dTLC_{TL}(q)}{dq} = 0 \Rightarrow q_{TL}^* = \sqrt{\frac{frd}{\alpha vh}} = \sqrt{\frac{20(2.4206)532}{(1)25000(0.3)}} = 1.8531 \text{ ton}$$

$$\begin{aligned} TLC_{TL}(q_{TL}^*) &= \frac{f}{q_{TL}^*} rd + \alpha vhq_{TL}^* \\ &= \frac{20}{1.8553} (2.4206)532 + (1)25000(0.3)1.8553 \\ &= 13,898.46 + 13,898.46 \\ &= \$27,796.93 / \text{yr} \end{aligned}$$

# Truck Shipment Example: Periodic

- Including the minimum charge and maximum payload restrictions:

$$q_{TL}^* = \min \left\{ \sqrt{\frac{f \max \{rd, MC_{TL}\}}{\alpha vh}}, q_{\max} \right\} \approx \sqrt{\frac{frd}{\alpha vh}}$$

# Truck Shipment Example: Periodic

17. What is the optimal LTL shipment size?

$$TLC_{LTL}(q) = TC_{LTL}(q) + IC(q) = \frac{f}{q} c_{LTL}(q) + \alpha v h q$$

$$q_{LTL}^* = \arg \min_q TLC_{LTL}(q) = 0.7288 \text{ ton}$$

- Must be careful in picking starting point for optimization since random choice likely to be infeasible:

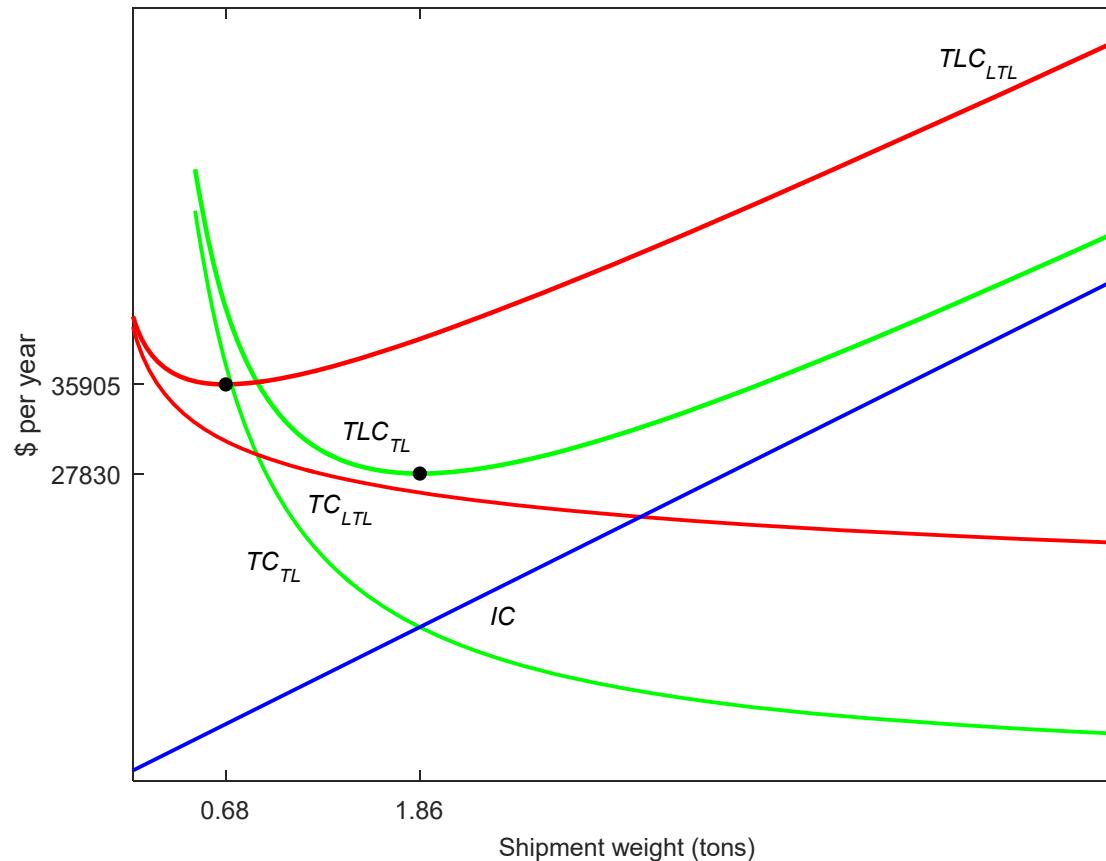
$$r_{LTL} = PPI_{LTL} \left[ \frac{\frac{s^2}{8} + 14}{\left( q^{\frac{1}{7}} d^{\frac{15}{29}} - \frac{7}{2} \right) (s^2 + 2s + 14)} \right]$$

$$\frac{150}{2,000} \leq q \leq \frac{10,000}{2,000}, \quad 37 \leq d \leq 3354, \quad 2000 \frac{q}{s} \leq 650 \text{ ft}^3$$

# Truck Shipment Example: Periodic

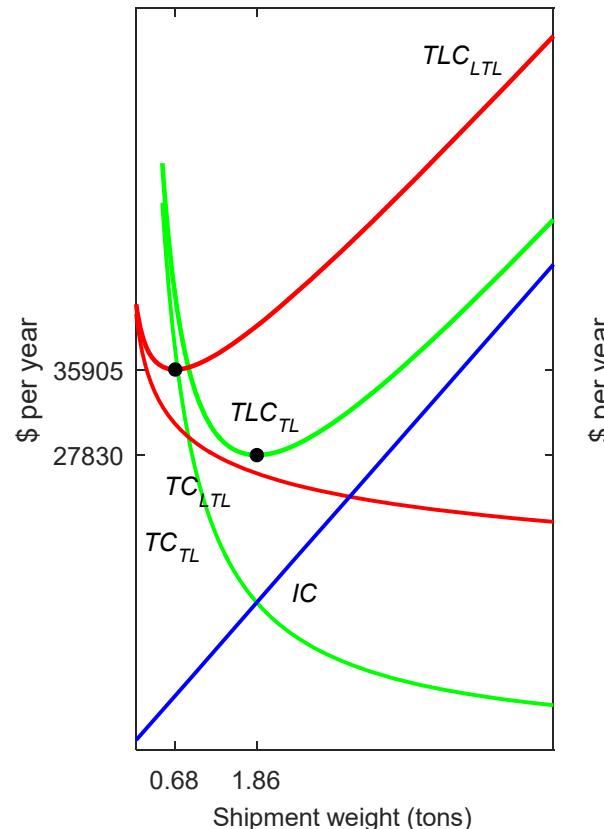
18. Should the product be shipped TL or LTL?

$$TLC_{LTL}(q^*_{LTL}) = TC_{LTL}(q^*_{LTL}) + IC(q^*_{LTL}) = 32,811.57 + 5,466.06 = \$38,277.63 / \text{yr}$$

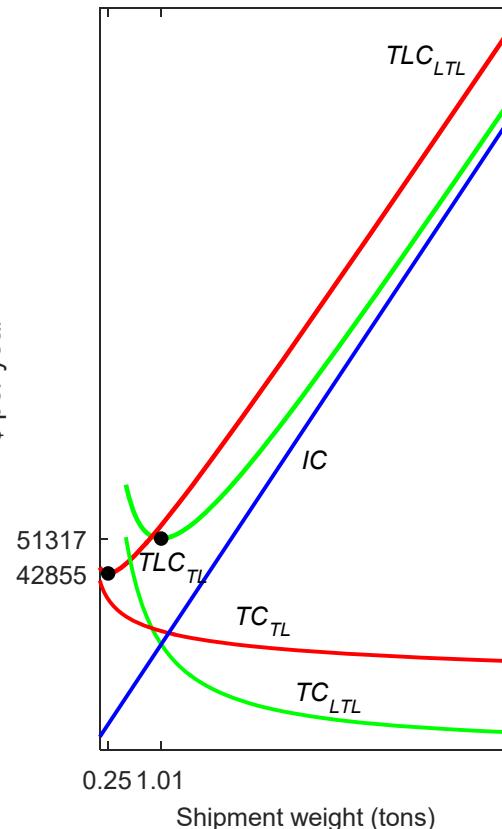


# Truck Shipment Example: Periodic

19. If the value of the product increased to \$85,000 per ton, should the product be shipped TL or LTL?



(a) \$25000 value per ton



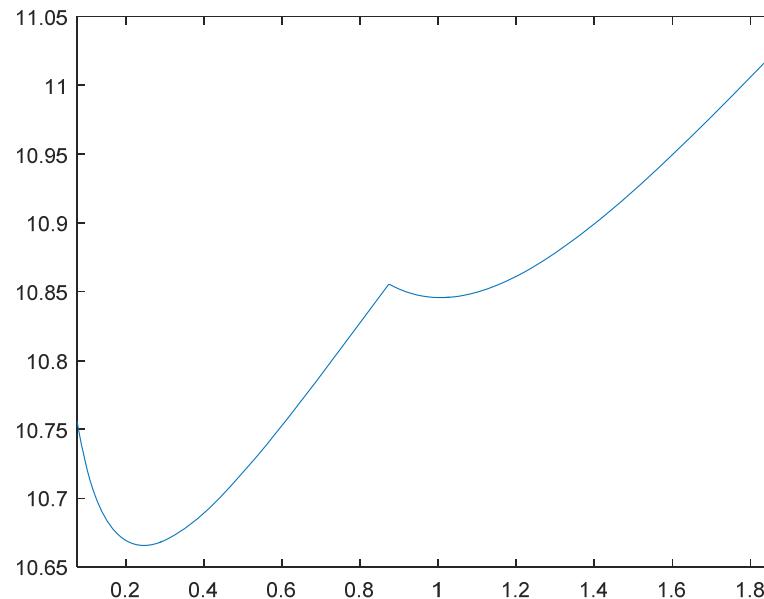
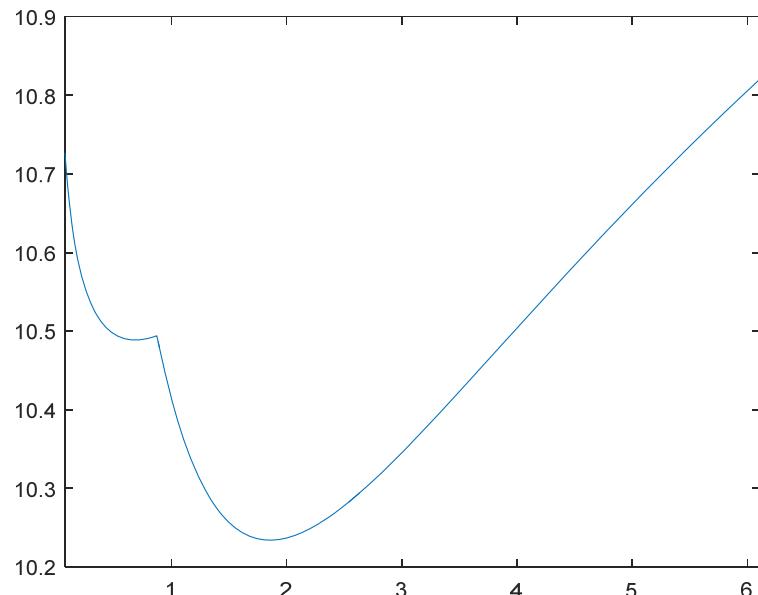
(b) \$85000 value per ton

# Truck Shipment Example: Periodic

- Better to pick from separate optimal TL and LTL because independent charge has two local minima:

$$q_0^* = \arg \min_q \{TLC_{TL}(q), TLC_{LTL}(q)\}$$

$$q_0^* \stackrel{!}{=} \arg \min_q \left\{ \frac{f}{q} c_0(q) + \alpha v h q \right\}$$



# Truck Shipment Example: Periodic

20. What is optimal independent shipment size to ship 80 tons per year of a Class 60 product valued at \$5000 per ton between Raleigh and Gainesville?

$$s = 32.16 \text{ lb/ft}^3$$

$$q_0^* = \arg \min_q \{TLC_{TL}(q), TLC_{LTL}(q)\} = 8.29 \text{ ton}$$

$$TLC_{TL}(q_0^*) = \$24,862.33 / \text{yr} < TLC_{LTL}(q_0^*)$$

# Truck Shipment Example: Periodic

21. What is the optimal shipment size if both shipments will always be shipped together on the same truck (with same shipment interval)?

$$d_1 = d_2, \quad h_1 = h_2, \quad \alpha_1 = \alpha_2$$

$$f_{\text{agg}} = f_1 + f_2 = 20 + 80 = 100 \text{ ton}$$

$$s_{\text{agg}} = \frac{(\text{aggregate weight, in lb})}{(\text{aggregate cube, in ft}^3)} = \frac{f_{\text{agg}}}{\frac{f_1}{s_1} + \frac{f_2}{s_2}} = \frac{100}{\frac{20}{4.44} + \frac{80}{32.16}} = 14.31 \text{ lb/ft}^3$$

$$\nu_{\text{agg}} = \frac{f_1}{f_{\text{agg}}} \nu_1 + \frac{f_2}{f_{\text{agg}}} \nu_2 = \frac{20}{100} 85,000 + \frac{80}{100} 5000 = \$21,000 / \text{ton}$$

$$q_{TL}^* = \sqrt{\frac{f_{\text{agg}}rd}{\alpha\nu_{\text{agg}}h}} = \sqrt{\frac{100(2.4206)532}{(1)21000(0.3)}} = 4.5212 \text{ ton}$$

# Truck Shipment Example: Periodic

- Summary of results:

:	f	s	v	qmax	TLC	q	t
1:	20	4.44	85,000	6.11	45,675.80	0.26	4.78
2:	80	32.16	5,000	25.00	24,862.33	8.29	37.84
1+2:					70,538.13		
Aggregate:	100	14.31	21,000	19.68	56,966.75	4.52	16.51

# Location and Transport Costs

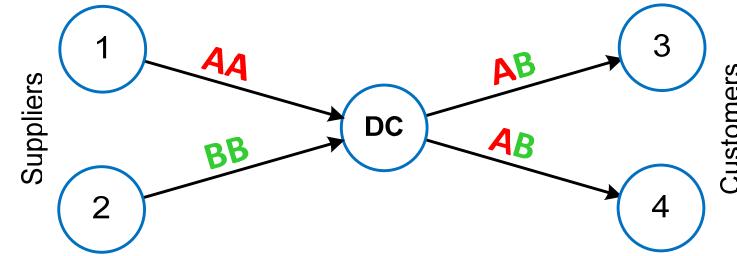
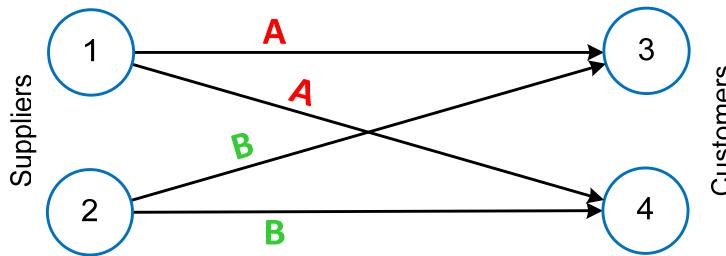
- Monetary weights  $w$  used for location are, in general, a function of the location of a NF
  - Distance  $d$  appears in optimal TL size formula
  - TC & IC functions of location  $\Rightarrow$  Need to minimize TLC instead of TC
  - FTL (since size is fixed at max payload) results in only constant weights for location  $\Rightarrow$  Need to only minimize TC since IC is constant in TLC

$$\begin{aligned}
 TLC_{TL}(\mathbf{x}) &= \sum_{i=1}^m w_i(\mathbf{x})d_i(\mathbf{x}) + \alpha v h q_i(\mathbf{x}) = \sum_{i=1}^m \frac{f_i}{q_i(\mathbf{x})} r d_i(\mathbf{x}) + \alpha v h q_i(\mathbf{x}) \\
 &= \sum_{i=1}^m \frac{f_i}{\sqrt{\frac{f_i r d_i(\mathbf{x})}{\alpha v h}}} r d_i(\mathbf{x}) + \alpha v h \sqrt{\frac{f_i r d_i(\mathbf{x})}{\alpha v h}} = \sum_{i=1}^m \sqrt{\frac{f_i r d_i(\mathbf{x})}{\alpha v h}} + \sqrt{f_i r d_i(\mathbf{x})} \sqrt{\alpha v h}
 \end{aligned}$$

$$TLC_{FTL}(\mathbf{x}) = \sum_{i=1}^m \frac{f_i}{q_{\max}} r d_i(\mathbf{x}) + \alpha v h q_{\max} = \sum_{i=1}^m w_i d_i(\mathbf{x}) + \alpha v h q_{\max} = TC_{FTL}(\mathbf{x}) + \text{constant}$$

# Transshipment

- *Direct*: P2P shipments from Suppliers to Customers



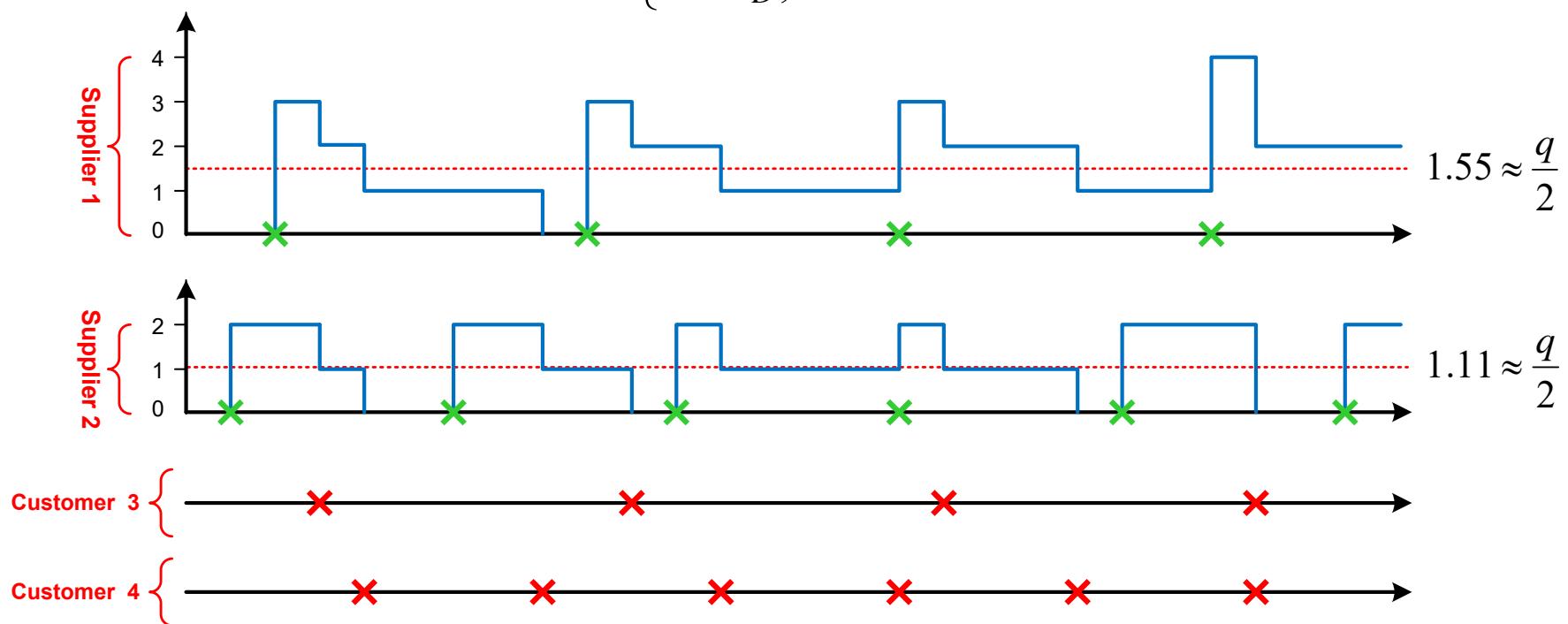
- *Transshipment*: use DC to consolidate outbound shipments
  - *Uncoordinated*: determine separately each optimal inbound and outbound shipment  $\Rightarrow$  hold inventory at DC
  - *Cross-dock*: use single shipment interval for all inbound and outbound shipments  $\Rightarrow$  no inventory at DC

# Uncoordinated Inventory

- Average pipeline inventory level at DC:

$$\alpha = \alpha_O + \alpha_D$$

$$= \begin{cases} \alpha_O + \frac{1}{2}, & \text{inbound} \\ 0 + \alpha_D, & \text{outbound} \end{cases}$$



# TLC with Transshipment

- Uncoordinated:  $TLC_i = TLC$  of supplier/customer  $i$

$$q_i^* = \arg \min_q TLC_i(q)$$

$$TLC^* = \sum TLC_i(q_i^*)$$

- Cross-docking:  $t = \frac{q}{f}$ , shipment interval

$$TLC_i(t) = \frac{c_0(t)}{t} + \alpha v h f t$$

$c_0(t)$  = independent transport charge as function of  $t$

$$\alpha = \begin{cases} \alpha_O + 0, & \text{inbound} \\ 0 + \alpha_D, & \text{outbound} \end{cases}$$

$$t^* = \arg \min_t \sum TLC_i(t)$$

$$TLC^* = \sum TLC_i(t^*)$$

# TLC and Location

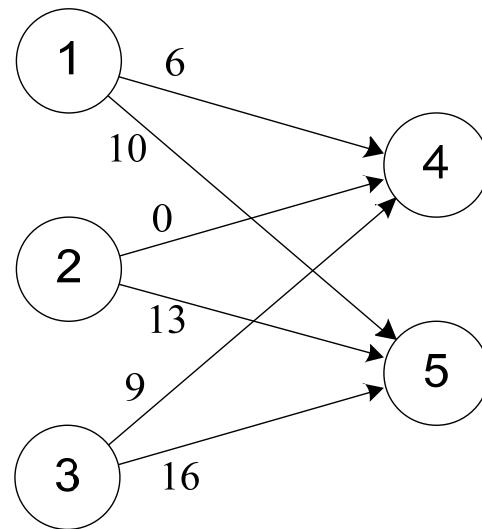
- TLC should include all logistics-related costs
  - ⇒ TLC can be used as sole objective for network design (incl. location)
- Assume public warehouses to be used for all DCs
  - ⇒ Pay only for time each unit spends in WH ⇒ No fixed cost at DC
- Transport fixed costs:
  - Costs that are independent of shipment size (e.g., \$/mi), that make it worthwhile to incur the inventory cost associated with larger shipment sizes in order to spread out the fixed cost
- Example: Optimal Number DCs for Lowe's
  - Mix of top-down (COGS) and bottom-up (typical load/TL parameters)
  - FTL for all inbound and outbound shipments
  - ALA used to determine TC for given number of DCs
  - $IC = \alpha vhq_{\max} \times (\text{number of suppliers} \times \text{number of DCs} + \text{number stores})$
  - Ignoring max DC-to-store distance constraints, consolidation, etc.

# Topics

1. Introduction
2. Facility location
3. Freight transport
  - Midterm exam
- 4. Network models**
5. Routing
6. Warehousing
  - Final project
  - Final exam

# Graph Representations

- Complete bipartite directed (or digraph):
  - Suppliers to multiple DCs, single mode of transport



C:	1	2
- : -----		
1:	6	10
2:	0	13
3:	9	16

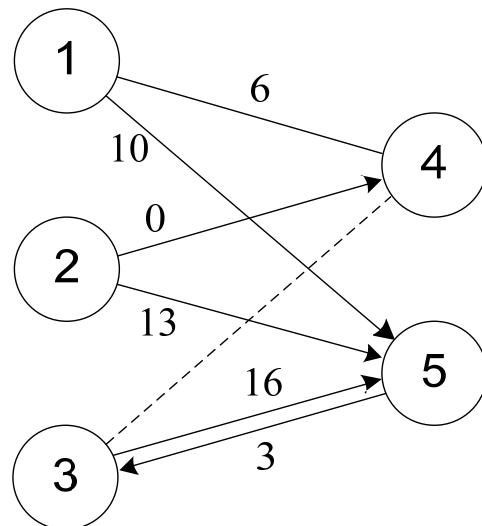
Interlevel matrix

W =	0	0	0	6	10
	0	0	0	Nan	13
	0	0	0	9	16
	0	0	0	0	0
	0	0	0	0	0

Weighted adjacency matrix

# Graph Representations

- Bipartite:
  - One- or two-way connections between nodes in two groups



$W =$	0	0	0	6	10
	0	0	0	NaN	13
	0	0	0	0	16
	6	0	0	0	0
	0	0	3	0	0

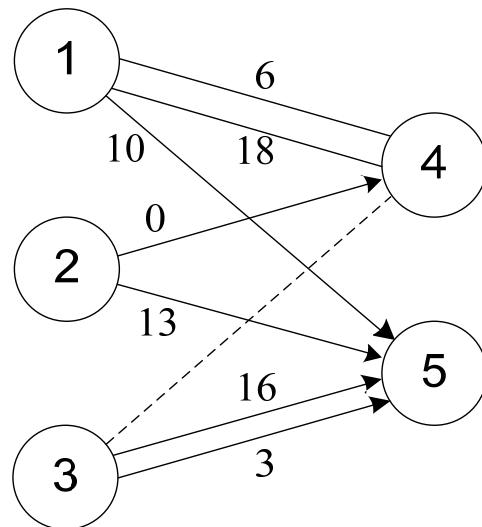
$IJC =$	4	1	6		
	5	3	3		
	1	4	6		
	2	4	0		
	1	5	10		
	2	5	13		
	3	5	16		

} Arc list matrix



# Graph Representations

- Multigraph:
  - Multiple connections, multiple modes of transport



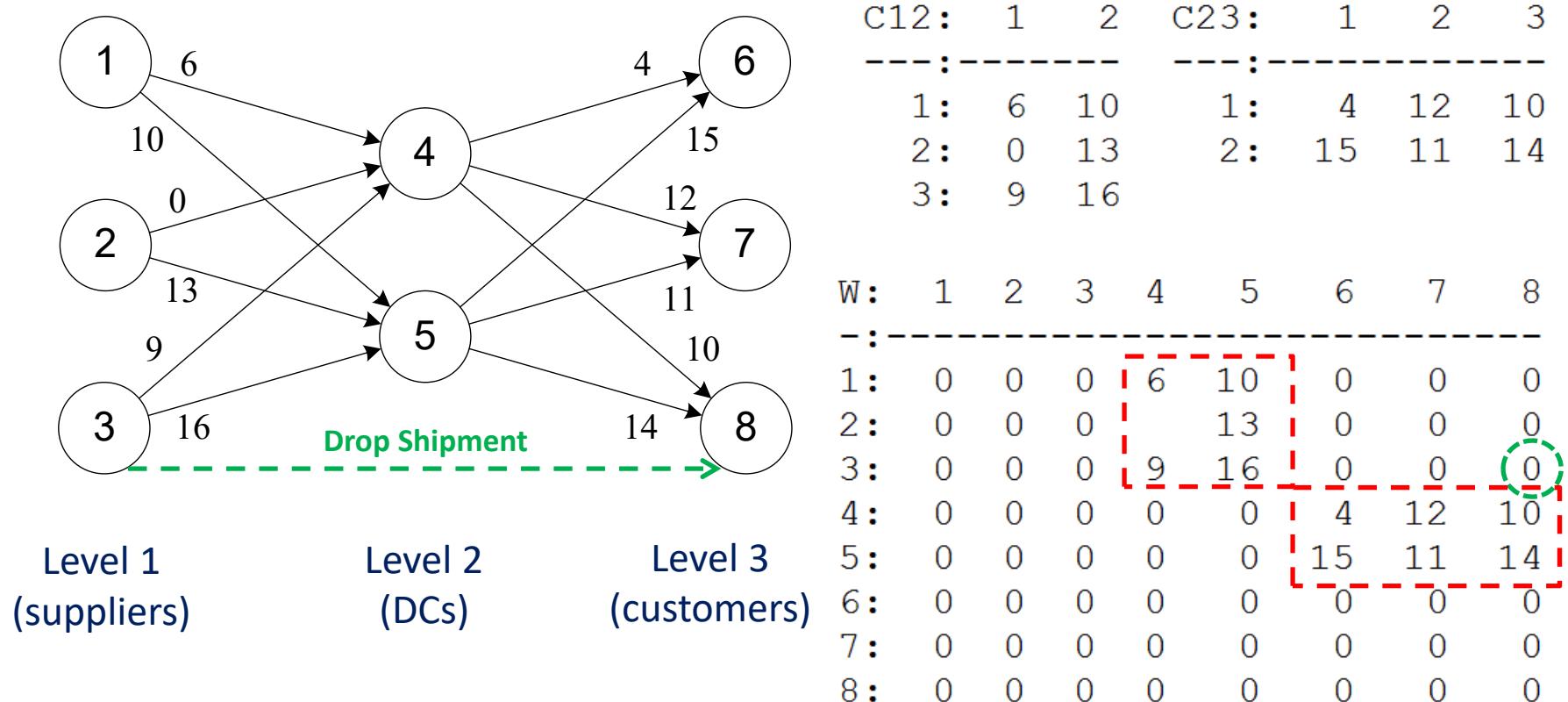
IJC =	1	-4	6
	1	-4	18
	1	5	10
	2	4	0
	2	5	13
	3	5	16
	3	5	3

no_W =	0	0	0	24	10
	0	0	0	Nan	13
	0	0	0	0	19
	24	0	0	0	0
	0	0	0	0	0

Can't represent using adjacency matrix

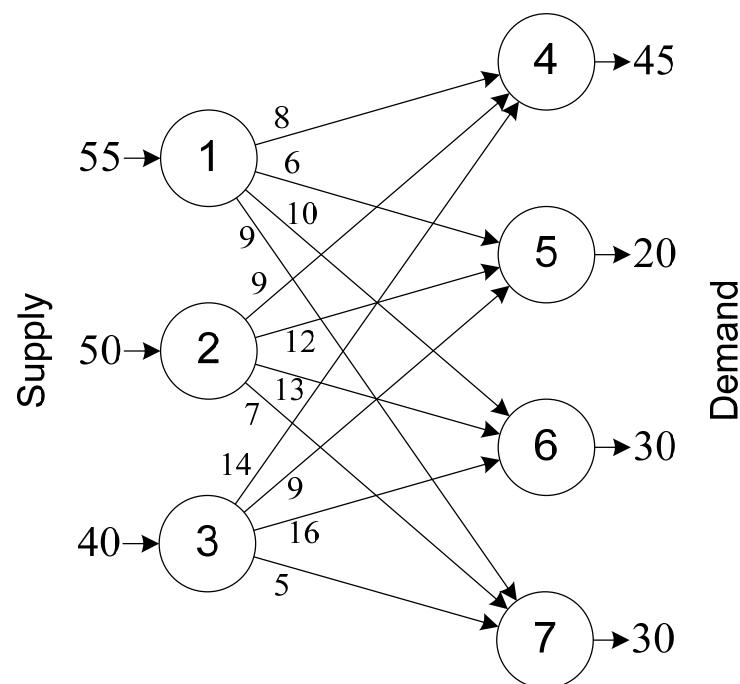
# Graph Representations

- Complete multipartite directed:
  - Typical supply chain (no drop shipments)



# Transportation Problem

- Satisfy node demand from supply nodes
  - Can be used for allocation in ALA when NFs have capacity constraints
  - Min cost/distance allocation = infinite supply at each node



Trans	4	5	6	7	Supply
1	8	6	10	9	55
2	9	12	13	7	50
3	14	9	16	5	40
Demand	45	20	30	30	

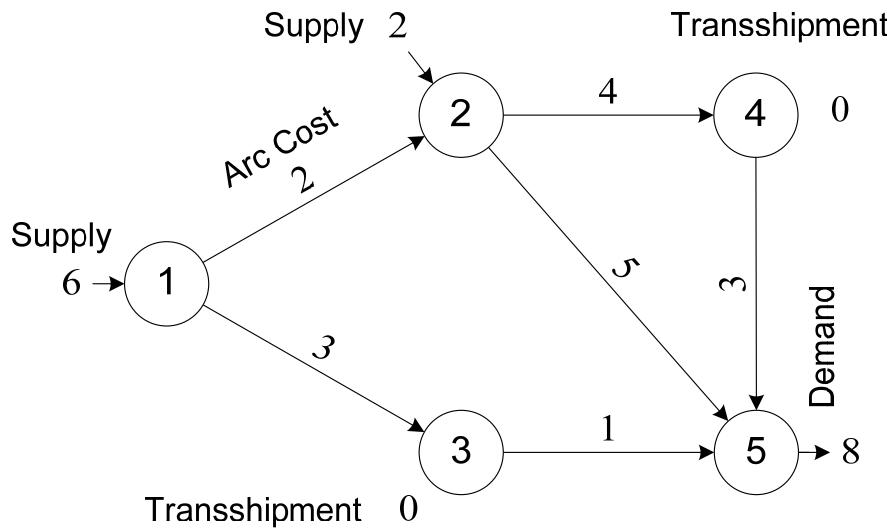
# Greedy Solution Procedure

- Procedure for transportation problem: *Continue to select lowest cost supply until all demand is satisfied*
  - Fast, but not always optimal for transportation problem
  - Dijkstra's shortest path and simplex method for LP are optimal greedy procedures

Trans	4	5	6	7	Supply
1	8	6	10	9	55
2	9	12	13	7	50
3	14	9	16	5	40
Demand	45	20	30	30	

# Min Cost Network Flow (MCNF) Problem

- Most general network problem, can solve using any type of graph representation



$s_i$  = net supply of node  $i$

$$= \begin{cases} > 0, & \text{supply node} \\ < 0, & \text{demand node} \\ = 0, & \text{transshipment node} \end{cases}$$

Arc cost:  $\mathbf{c} = [2 \ 3 \ 4 \ 5 \ 1 \ 3]'$

Net node supply:  $\mathbf{s} = [6 \ 2 \ 0 \ 0 \ -8]'$

$$\text{Incidence Matrix : } \mathbf{A} = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ -1 & 0 & 1 & 1 & 0 & 0 \\ 0 & -1 & 0 & 0 & 1 & 0 \\ 0 & 0 & -1 & 0 & 0 & 1 \\ 0 & 0 & 0 & -1 & -1 & -1 \end{bmatrix}$$

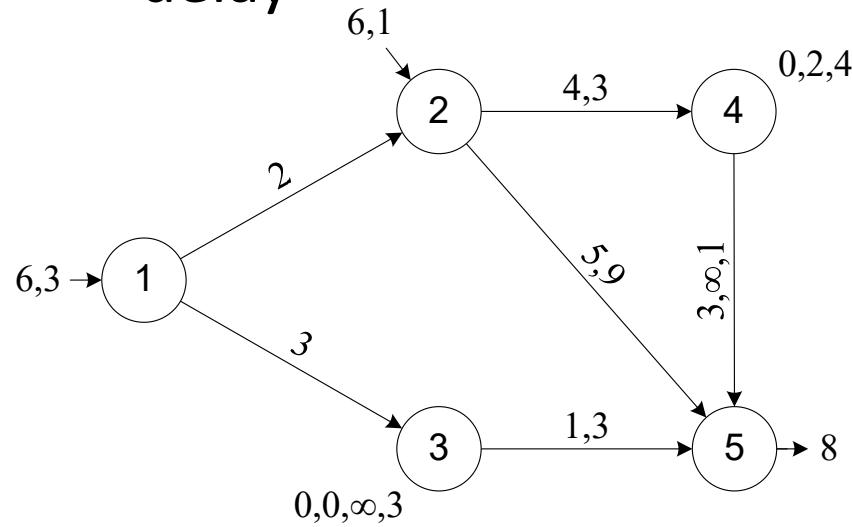
Row for node 5  
is redundant

MCNF:	lhs	C	C	C	C	C	C	rhs
Min:		2	3	4	5	1	3	
1:	6	1	1	0	0	0	0	6
2:	2	-1	0	1	1	0	0	2
3:	0	0	-1	0	0	1	0	0
4:	0	0	0	-1	0	0	1	0
lb:		0	0	0	0	0	0	
ub:		Inf	Inf	Inf	Inf	Inf	Inf	

$$\begin{aligned} \text{MCNF: } & \max \mathbf{c}'\mathbf{x} \\ \text{s.t. } & \mathbf{Ax} = \mathbf{s} \\ & \mathbf{x} \geq 0 \end{aligned}$$

# MCNF with Arc/Node Bounds and Node Costs

- Bounds on arcs/nodes can represent capacity constraints in a logistic network
- Node cost can represent production cost or intersection delay

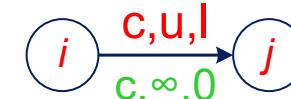


$s_i$  = net supply of node  $i$

$$= \begin{cases} > 0, & \text{supply node} \\ < 0, & \text{demand node} \\ = 0, & \text{transshipment node} \end{cases}$$

$s, nc, nu, nl$

$i$   
 $s, 0, \infty, 0$



IJCUL: 1 2 3 4 5

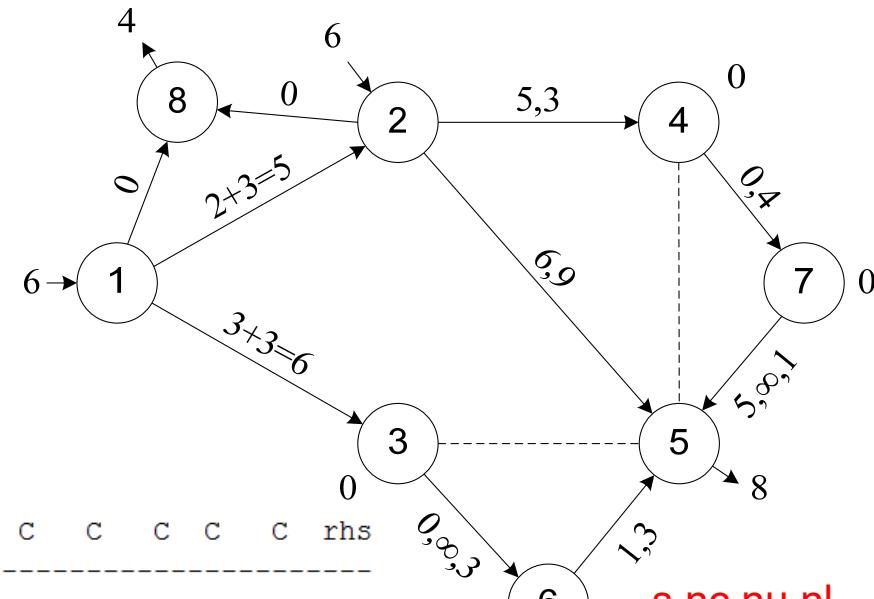
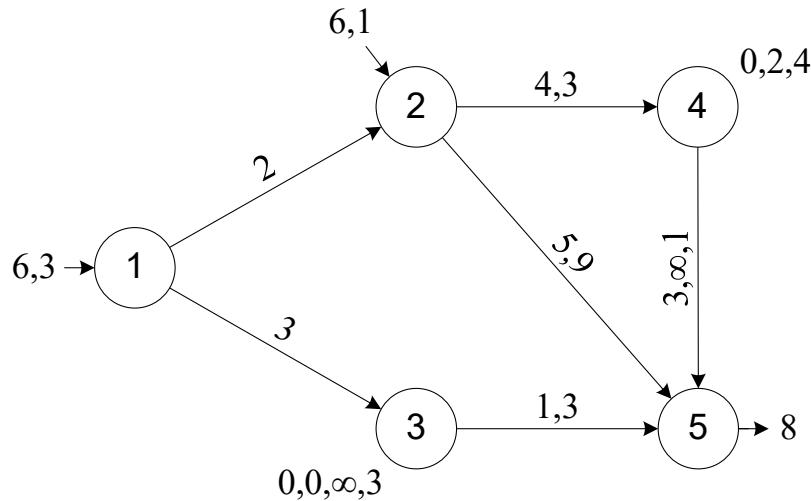
	1	2	3	4	5
1:	1	2	2	Inf	0
2:	1	3	3	Inf	0
3:	2	4	4	3	0
4:	2	5	5	9	0
5:	3	5	1	3	0
6:	4	5	3	Inf	1

SCUL: 1 2 3 4

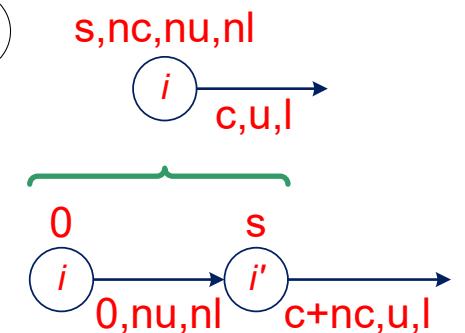
	1	2	3	4
1:	6	3	Inf	0
2:	6	1	Inf	0
3:	0	0	Inf	3
4:	0	2	4	0
5:	-8	0	Inf	0

# Expanded-Node Formulation of MCNF

- Node cost/constraints converted to arc cost/constraints
  - Dummy node (8) added so that supply = demand

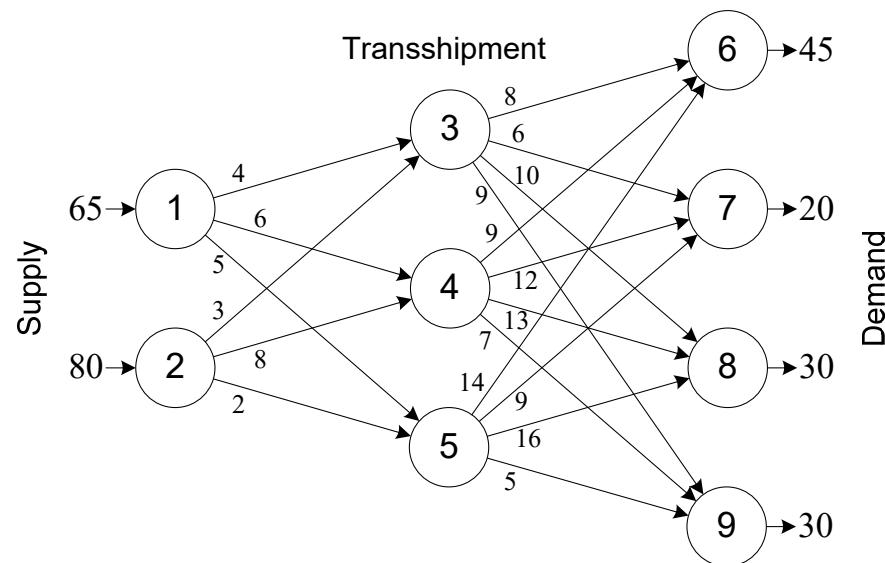


	lhs	c	c	c	c	c	c	c	c	c	rhs
-----:											
Min:		5	6	5	6	1	5	0	0	0	0
1:	6	1	1	0	0	0	0	0	1	0	6
2:	6	-1	0	1	1	0	0	0	0	1	6
3:	0	0	-1	0	0	0	0	1	0	0	0
4:	0	0	0	-1	0	0	0	0	1	0	0
5:	-8	0	0	0	-1	-1	-1	0	0	0	-8
6:	0	0	0	0	0	1	0	-1	0	0	0
7:	0	0	0	0	0	0	1	0	-1	0	0
lb:	0	0	0	0	0	1	3	0	0	0	0
ub:	Inf	Inf	3	9	3	Inf	Inf	4	Inf	Inf	0

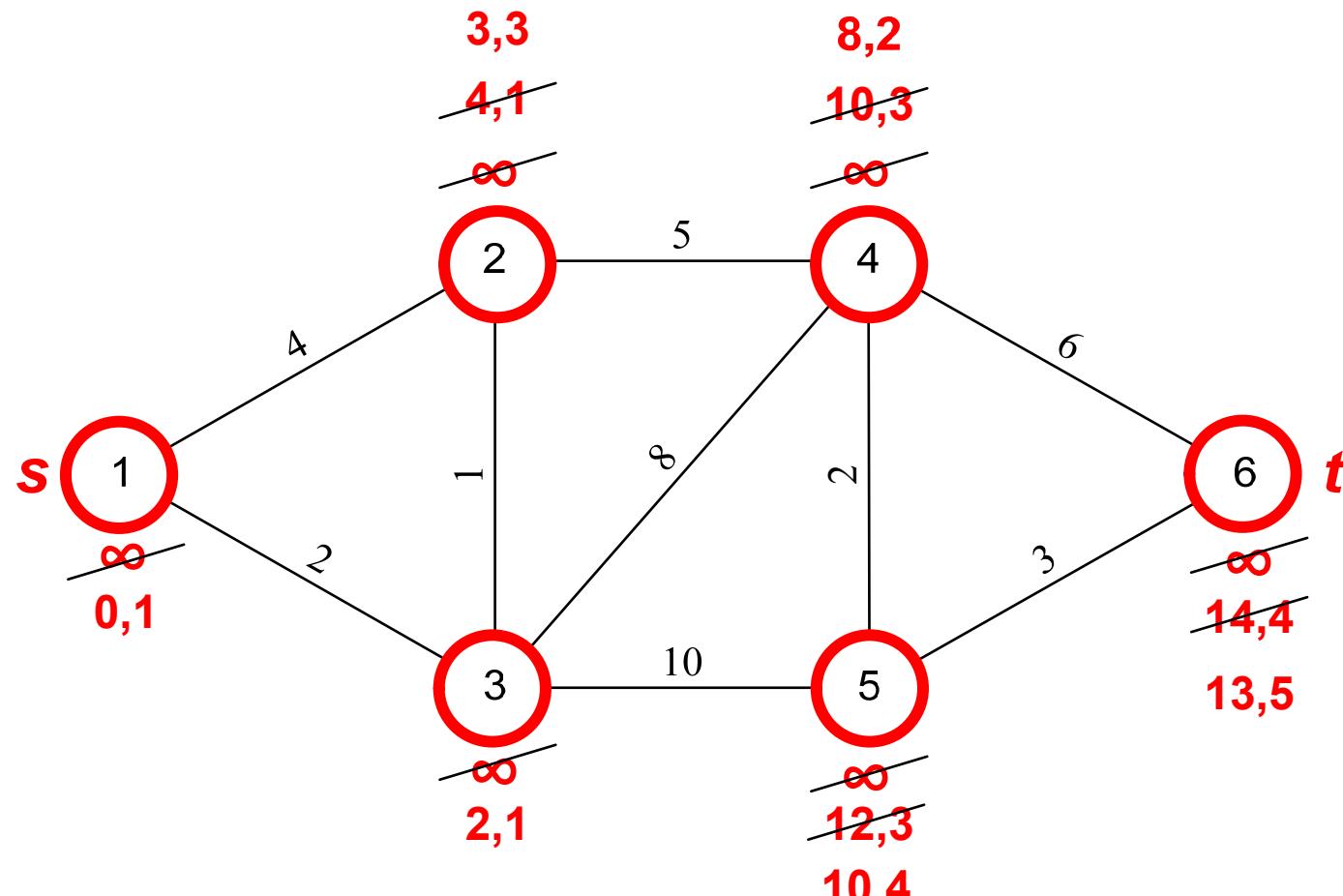


# Solving an MCNF as an LP

- Special procedures more efficient than LP were developed to solve MCNF and Transportation problems
  - e.g., Network simplex algorithm (MCNF)
  - e.g., Hungarian method (Transportation and *Transshipment*)
- Now usually easier to transform into LP since solvers are so good, with MCNF just aiding in formulation of problem:
  - Trans  $\Rightarrow$  MCNF  $\Rightarrow$  LP
  - Special, very efficient procedures only used for shortest path problem (Dijkstra)



# Dijkstra Shortest Path Procedure



Path:  $1 \leftarrow 3 \leftarrow 2 \leftarrow 4 \leftarrow 5 \leftarrow 6 : 13$

# Dijkstra Shortest Path Procedure

```

procedure dijkstra(W,n,s)
    S  $\leftarrow \{\}$ ,  $\bar{S} \leftarrow \{1, \dots, n\}$ 
    for  $i \in \bar{S}$ ,  $d(i) \leftarrow \infty$ , endfor
     $d(s) \leftarrow 0$ ,  $pred(s) \leftarrow 0$ 
    while  $|S| < n$ 
         $i \leftarrow \arg \min_j \{d(j) : j \in \bar{S}\}$ 
         $S \leftarrow S \cup i$ ,  $\bar{S} \leftarrow \bar{S} \setminus i$ 
        for  $j \in \arg \{W_{i(j)} : W_{ij} \neq 0\}$ 
            if  $d(j) > d(i) + W_{ij}$ 
                 $d(j) \leftarrow d(i) + W_{ij}$ 
                 $pred(j) \leftarrow i$ 
            endif
        endfor
    endwhile
    return d, pred

```

```

%% DIJKSTRA Matlab code,
% given W, n, and s
S = [];
nS = 1:n;
d = inf(1,n);
d(s) = 0; pred(s) = 0;
while length(S) < n
    [di, idx] = min(d(nS));
    i = nS(idx);
    S = [S i];
    nS(idx) = [];
    pred(d > di + W(i,:)) = i;
    d = min(d, di + W(i,:));
end
d, pred

```

} Index to index vector nS

} Order important

$O(2^n)$	Simplex (LP)
$O(n^4)$	Ellipsoid (LP)
$O(n^3)$	Hungarian (transportation)
$O(n^2)$	Dijkstra (linear min)
$O(m \log n)$	Dijkstra (Fibonacci heap)

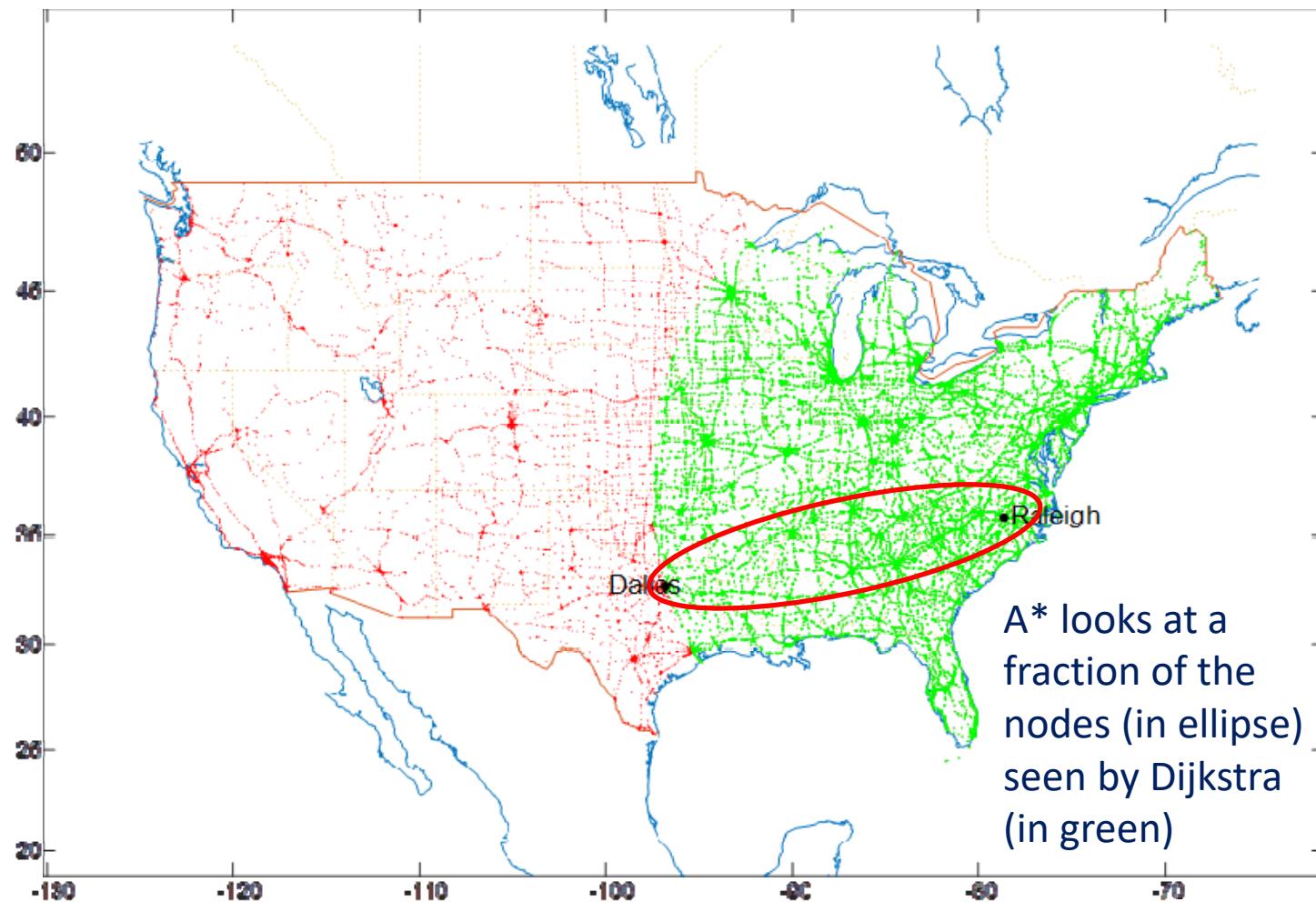
$m$  no. arcs

# Other Shortest Path Procedures

- Dijkstra requires that all arcs have nonnegative lengths
  - Is a “label setting” algorithm since step to final solution made as each node labeled
  - Can find longest path (used in CPM) by making by negating arc lengths
- Networks with some negative arcs require slower “label correcting” procedures that repeatedly check for optimality at all nodes or detect a negative cycle
  - Negative arcs used in project scheduling to represent maximum lags between activities
- A\* algorithm adds to Dijkstra an heuristic estimate of each node’s remaining distance to destination
  - Used in AI search for all types of applications (tic-tac-toe, chess)
  - In path planning applications, great circle distance from each node to destination could be used as estimate of remaining distance

# A\* Path Planning Example 1

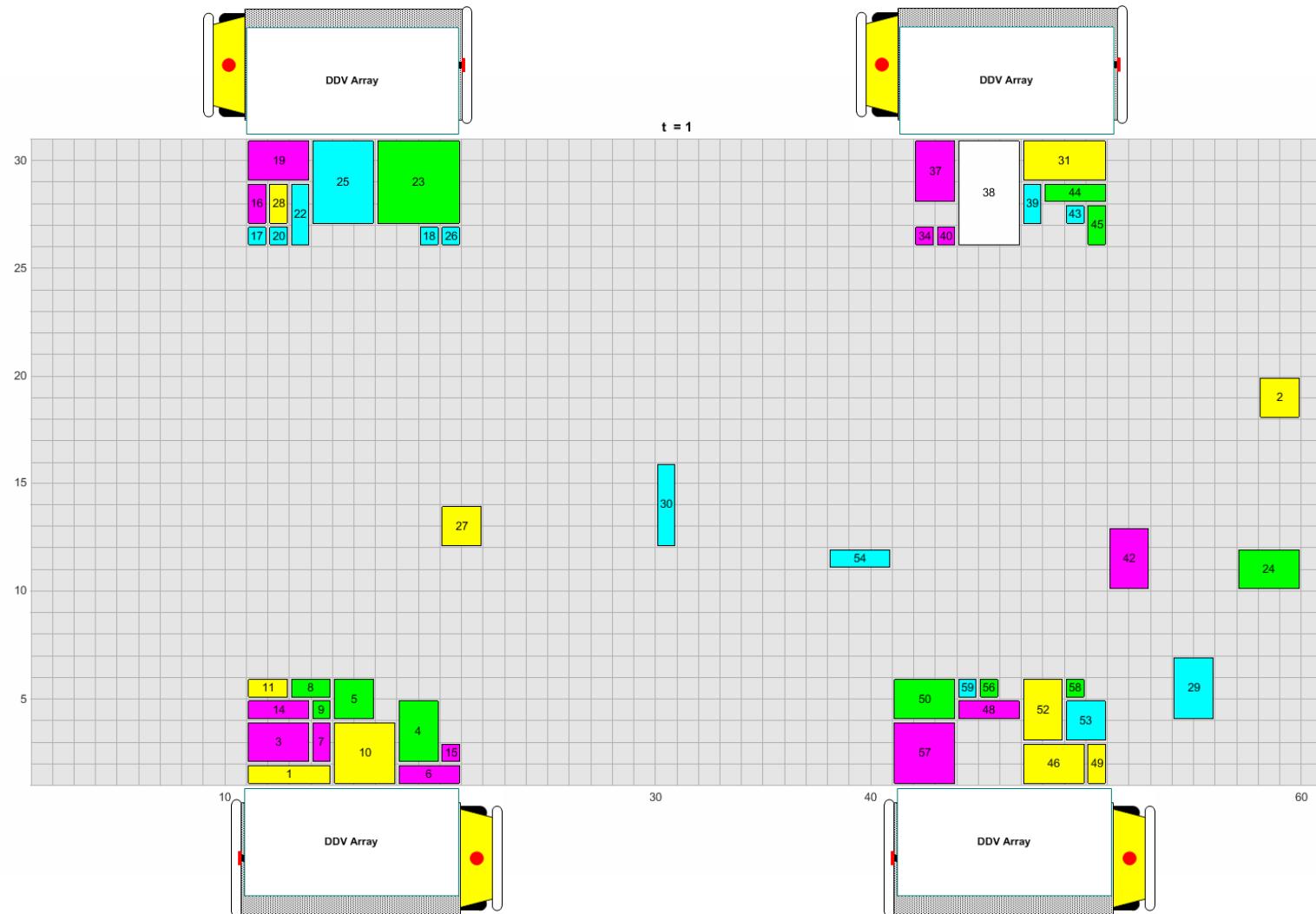
$$d_{A^*}(\text{Raleigh}, \text{Dallas}) = d_{dijk}(\text{Raleigh}, i) + d_{GC}(i, \text{Dallas}), \quad \text{for each node } i$$



# A\* Path Planning Example 2

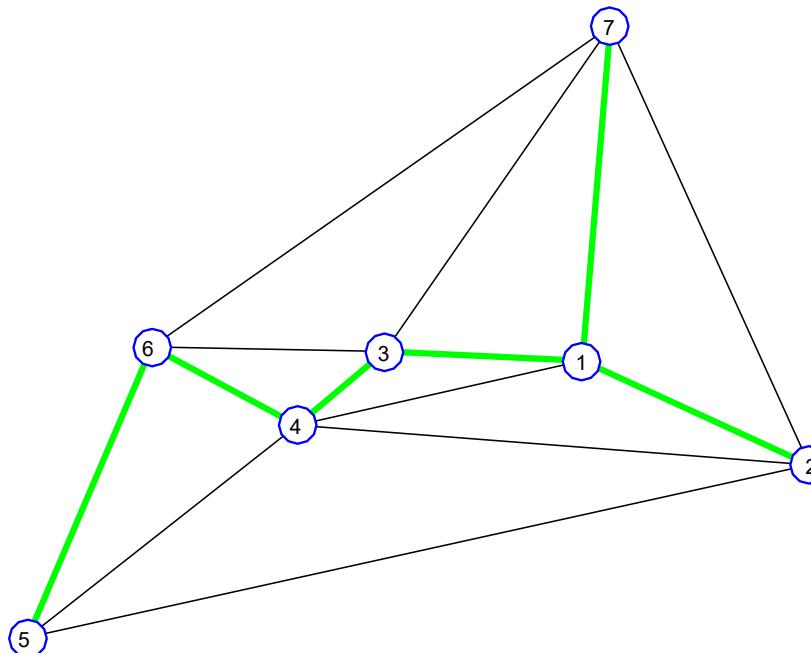
- 3-D  $(x, y, t)$  A\* used for planning path of each container in a DC
- Each container assigned unique priority that determines planning sequence
  - Paths of higher-priority containers become obstacles for subsequent containers

# A\* Path Planning Example 2



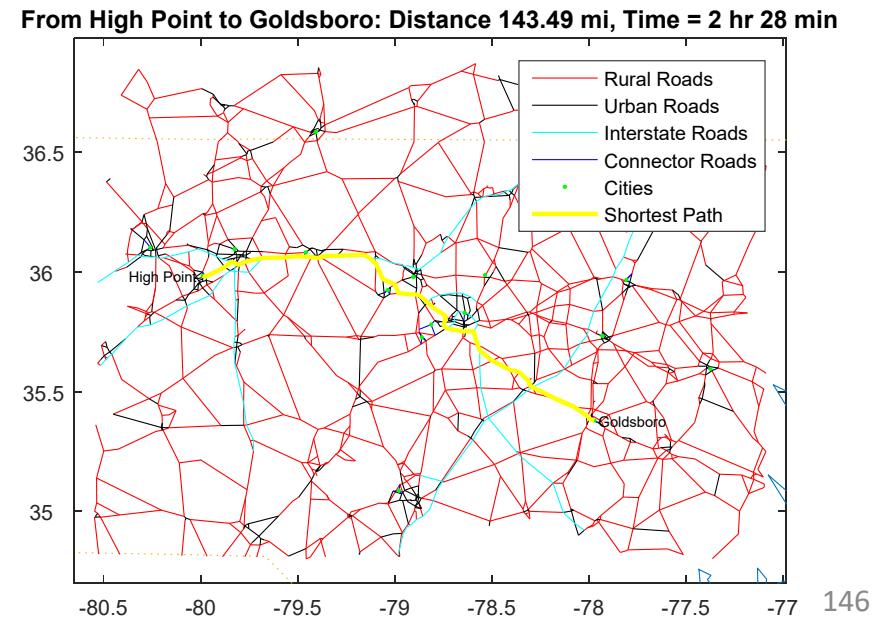
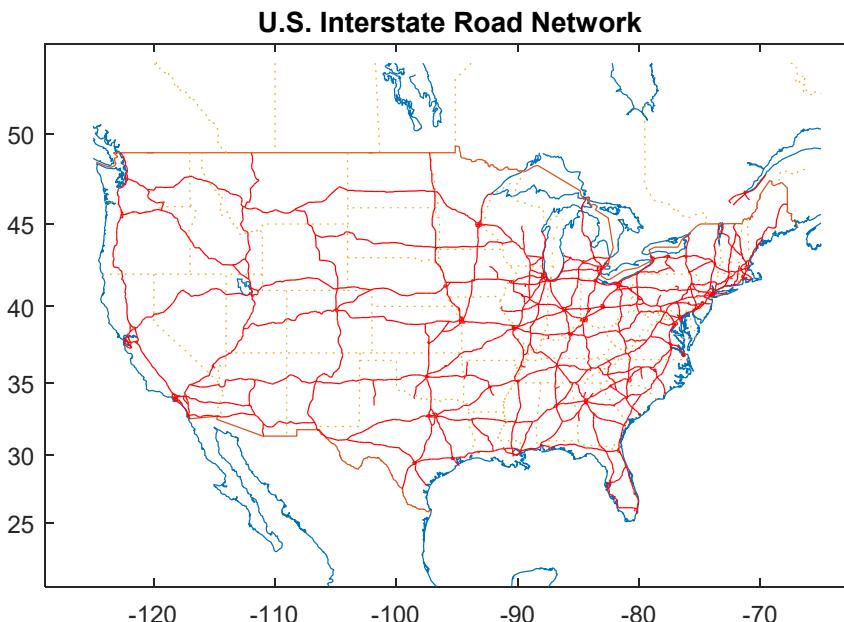
# Minimum Spanning Tree

- Find the minimum cost set of arcs that connect all nodes
  - Undirected arcs: Kruskal's algorithm (easy to code)
  - Directed arcs: Edmond's branching algorithm (hard to code)



# U.S. Highway Network

- Oak Ridge National Highway Network
  - Approximately 500,000 miles of roadway in US, Canada, and Mexico
  - Created for truck routing, does not include residential
  - Nodes attributes: XY, FIPS code
  - Arc attributes: IJD, Type (Interstate, US route), Urban



# FIPS Codes

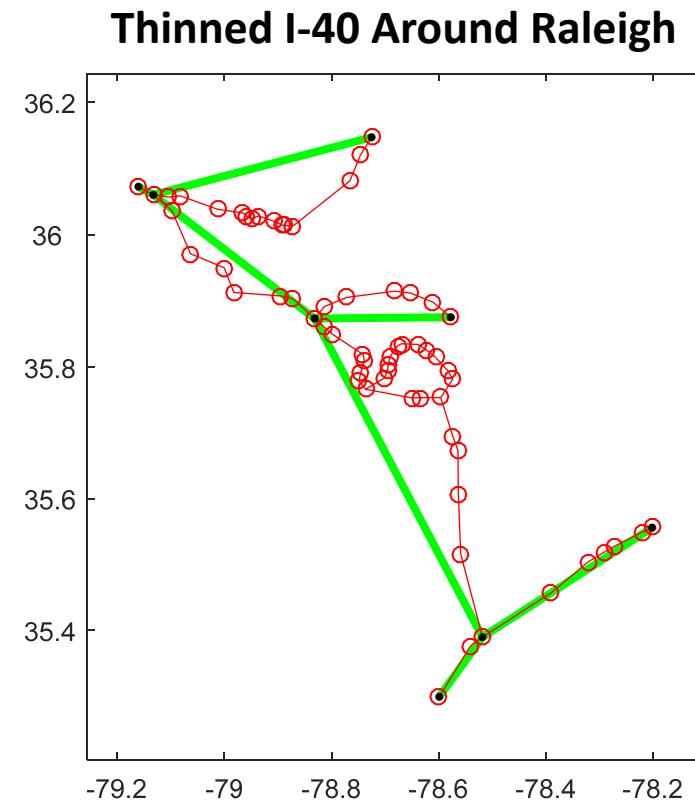
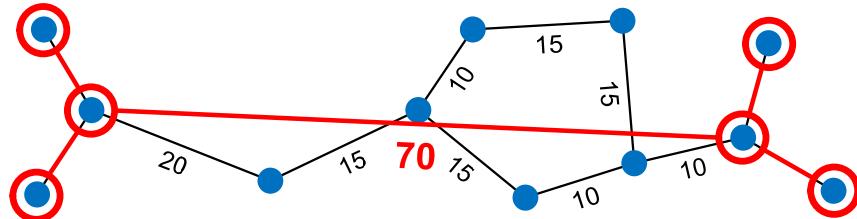
- Federal Information Processing Standard (FIPS) codes used to uniquely identify states (2-digit) and counties (3-digit)
  - 5-digit Wake county code = 2-digit state + 3-digit county  
= 37183 = 37 NC FIPS + 183 Wake FIPS

1 AL Alabama	22 LA Louisiana	40 OK Oklahoma
2 AK Alaska	23 ME Maine	41 OR Oregon
4 AZ Arizona	24 MD Maryland	42 PA Pennsylvania
5 AR Arkansas	25 MA Massachusetts	44 RI Rhode Island
6 CA California	26 MI Michigan	45 SC South Carolina
8 CO Colorado	27 MN Minnesota	46 SD South Dakota
9 CT Connecticut	28 MS Mississippi	47 TN Tennessee
10 DE Delaware	29 MO Missouri	48 TX Texas
11 DC Dist Columbia	30 MT Montana	49 UT Utah
12 FL Florida	31 NE Nebraska	50 VT Vermont
13 GA Georgia	32 NV Nevada	51 VA Virginia
15 HI Hawaii	33 NH New Hampshire	53 WA Washington
16 ID Idaho	34 NJ New Jersey	54 WV West Virginia
17 IL Illinois	35 NM New Mexico	55 WI Wisconsin
18 IN Indiana	36 NY New York	56 WY Wyoming
19 IA Iowa	37 NC North Carolina	72 PR Puerto Rico
20 KS Kansas	38 ND North Dakota	88 Canada
21 KY Kentucky	39 OH Ohio	91 Mexico

# Road Network Modifications

## 1. Thin

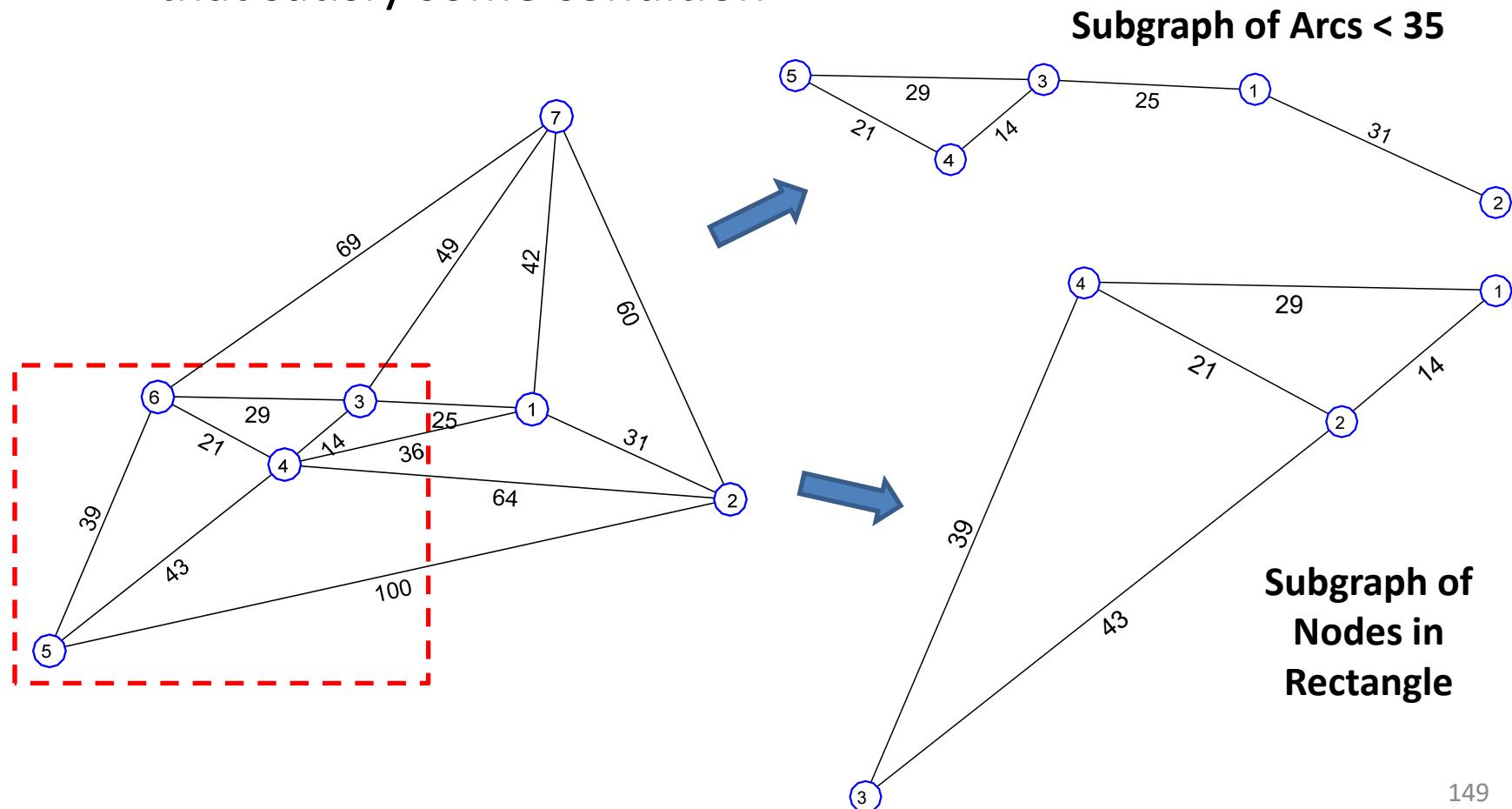
- Remove all degree-2 nodes from network
- Add cost of both arcs incident to each degree-2 node
- If results in multiple arcs between pair of nodes, keep minimum cost



# Road Network Modifications

## 2. Subgraph

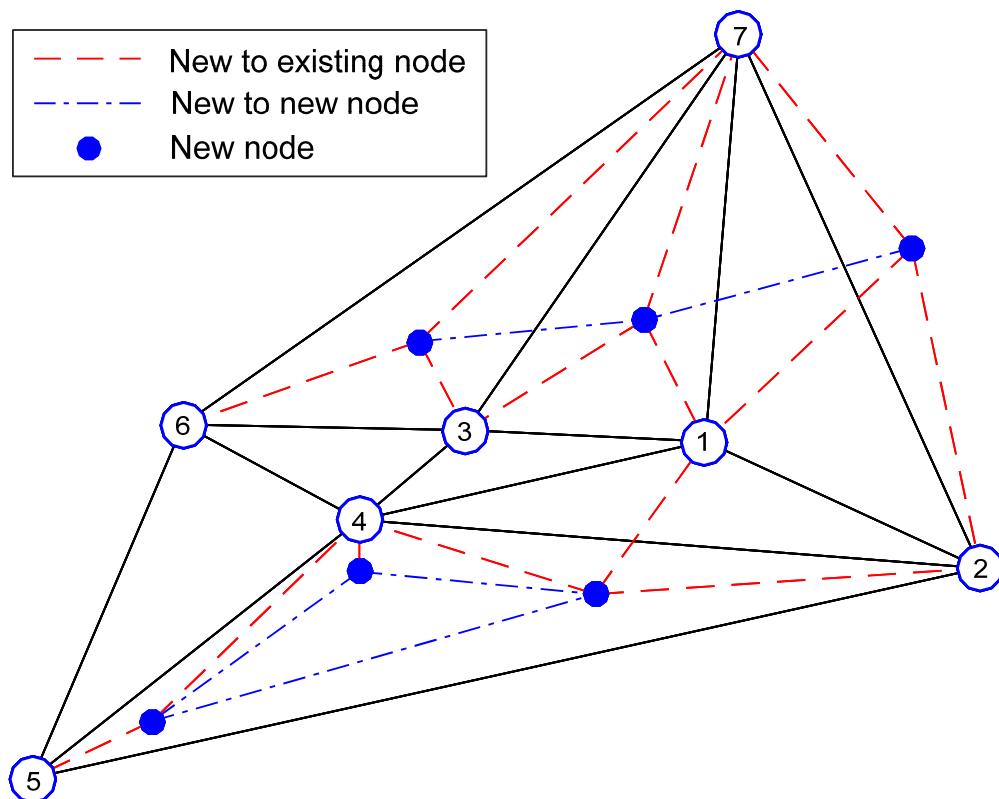
- Extract portion of graph with only those nodes and/or arcs that satisfy some condition



# Road Network Modifications

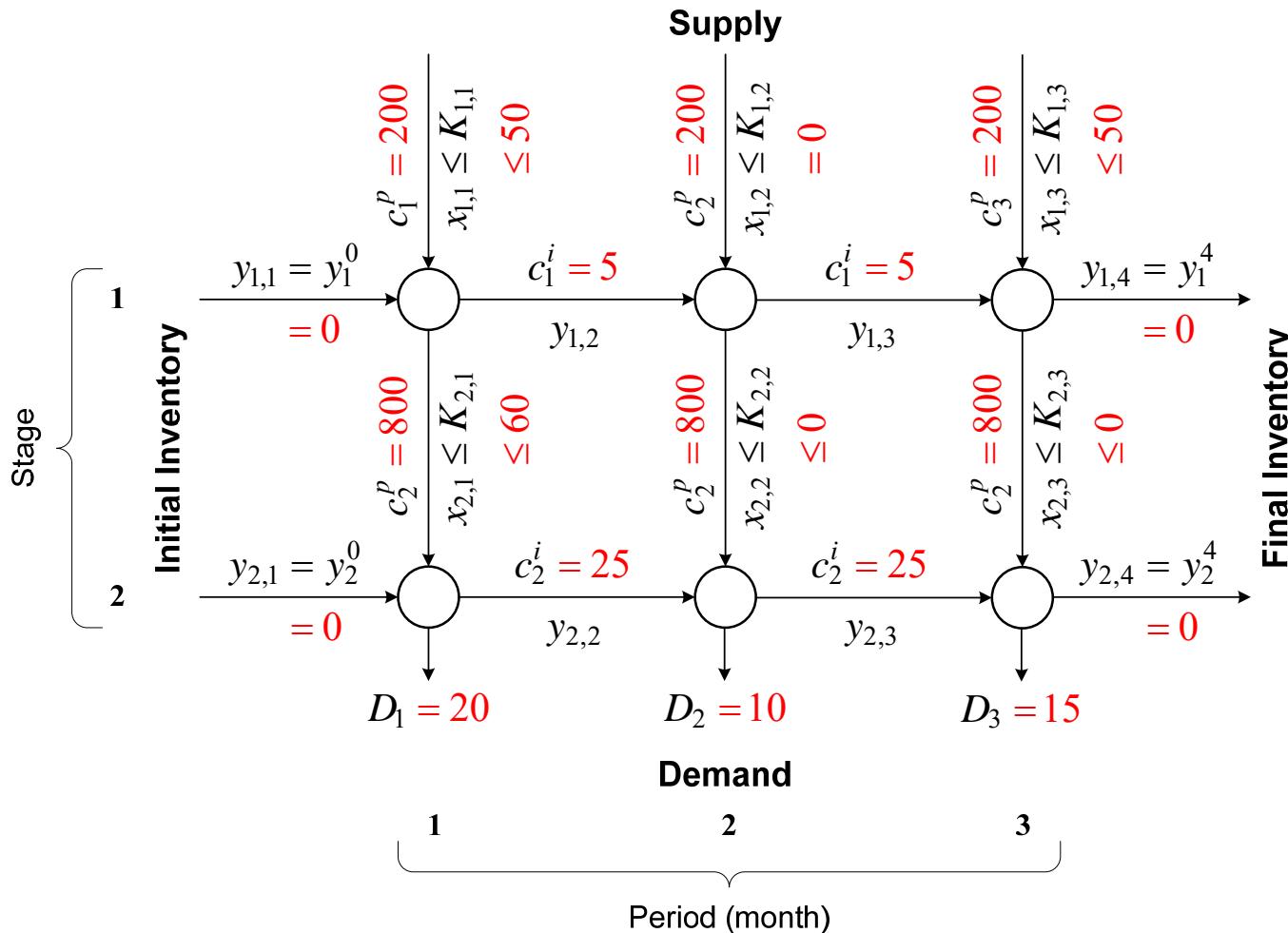
## 3. Add connector

- Given new nodes, add arcs that connect the new nodes to the existing nodes in a graph and to each other



- Distance of connector arcs = GC distance x circuity factor (1.5)
- New node connected to 3 closest existing nodes, except if
  - Ratio of closest to 2<sup>nd</sup> and 3<sup>rd</sup> closest < threshold (0.1)
  - Distance shorter using other connector and graph

# Production and Inventory: One Product



$$h = 0.3 \frac{\$}{\$-\text{yr}} = \frac{0.3}{\text{yr}}$$

$$\frac{h}{T} = \frac{0.3}{12} = \frac{0.025}{\text{month}}$$

$$c_m^i = \frac{h}{T} \sum_{j=1}^m c_j^p$$

$$c_1^i = \frac{0.3}{12} 200 = 5$$

$$c_2^i = \frac{0.3}{12} (200 + 800) \\ = 25$$

# Production and Inventory: One Product

$$\min \sum_{m=1}^M \sum_{t=1}^T c_m^p x_{mt} + \sum_{m=1}^M \sum_{t=1}^{T+1} c_m^i y_{mt}$$

subject to

Flow balance  $\left\{ \begin{array}{l} x_{mt} - x_{(m+1)t} + y_{mt} - y_{m(t+1)} = 0, \quad m = 1, \dots, M-1; t = 1, \dots, T \\ x_{Mt} + y_{Mt} - y_{M(t+1)} = D_t, \quad t = 1, \dots, T \end{array} \right.$

Capacity  $\left\{ \begin{array}{l} x_{mt} \leq K_{mt}, \quad m = 1, \dots, M; t = 1, \dots, T \end{array} \right.$

Initial/Final inventory  $\left\{ \begin{array}{l} y_{m1} = y_m^0, \quad m = 1, \dots, M \\ y_{m(T+1)} = y_m^{T+1}, \quad m = 1, \dots, M \end{array} \right. \quad \left. \begin{array}{l} x, y \geq 0 \text{ and continuous} \end{array} \right\}$

Use var. LB & UB instead of constraints

where

$M$  = number of production stages

$T$  = number of periods of production

$c_m^p$  = production cost in stage  $m$  (\$/ton)

$D_t$  = demand in period  $t$  (ton)

$x_{mt}$  = production at stage  $m$  in period  $t$  (ton)

$K_{mt}$  = capacity of stage  $m$  in period  $t$  (ton)

$c_m^i$  = inventory cost for stage  $m$  (\$/ton)

$y_m^0$  = initial inventory of stage  $m$  (ton)

$y_{mt}$  = stage- $m$  inventory period  $t-1$  to  $t$  (ton)

$y_m^{T+1}$  = final inventory of stage  $m$  (ton)

# Production and Inventory: Multiple Products

	$c^p$	$c^i$	$c^s$	$\mathbf{0}$	$c^p$	$c^i$	$c^s$	$\mathbf{0}$
<b>Product 1</b>	Flow balance $x \quad y$				$\frac{-K}{k} \leq 0$			$0$
	Capacity $x$							
		Setup $z$		$\frac{1}{k} \leq 0$				
<b>Product 2</b>	$0$				Flow balance $x \quad y$			
					Capacity $x$		$\frac{-K}{k} \leq 0$	
						Setup $z$	$\frac{1}{k} \leq 0$	
<b>Linking</b>		$k_1$	$+$				$k_2$	$= 1$

# Production and Inventory: Multiple Products

$k_{mtg} \in \{0,1\}$ , production indicator

$k_{mtg}$	1	2	3	4	5	6	7
1	0	1	1	0	1	1	1
2	0	0	0	1	0	0	0

$z_{mtg} \in \{0,1\}$ , setup indicator

$z_{mtg}$	1	2	3	4	5	6	7
1	0	1	0	0	1	0	0
2	0	0	0	1	0	0	0

	$-z_t$	$+$	$k_t$	$-$	$k_{t-1}$	$\leq$	0
	0		0		0		0
	0		0		1		-1
Don't want (not feasible)		0	1	0	1		
	0		1		1		0
	1		0		0		-1
Want (feasible)		1	0	1	-2		
	1		1		0		0
	1		1		1		-1

Feasible, but not min cost

# Production and Inventory: Multiple Products

$$\min \sum_{m=1}^M \sum_{t=1}^T \sum_{g=1}^G c_{mg}^p x_{mtg} + \sum_{m=1}^M \sum_{t=1}^{T+1} \sum_{g=1}^G c_{mg}^i y_{mtg} + \sum_{m=1}^M \sum_{t=1}^T \sum_{g=1}^G c_{mg}^s z_{mtg} (+ \underbrace{\mathbf{0}_{MTG}}_{\text{dummy}} k_{mtg})$$

subject to

Flow balance  $\left\{ \begin{array}{ll} x_{mtg} - x_{(m+1)tg} + y_{mtg} - y_{m(t+1)g} = 0, & m=1, \dots, M-1; t=1, \dots, T; g=1, \dots, G \\ x_{Mtg} + y_{Mtg} - y_{M(t+1)g} = D_{tg}, & t=1, \dots, T; g=1, \dots, G \end{array} \right.$

Capacity  $\left\{ \begin{array}{ll} x_{mtg} \leq K_{mg} k_{mtg}, & m=1, \dots, M; t=1, \dots, T; g=1, \dots, G \end{array} \right.$

Setup  $\left\{ \begin{array}{ll} -z_{m1g} + k_{m1g} \leq k_{mg}^0, & m=1, \dots, M; g=1, \dots, G \\ -z_{mtg} + k_{mtg} - k_{m(t-1)g} \leq 0, & m=1, \dots, M; t=2, \dots, T; g=1, \dots, G \end{array} \right.$

Linking  $\left\{ \begin{array}{ll} \sum_{g=1}^G k_{mtg} = 1, & m=1, \dots, M; t=1, \dots, T \end{array} \right.$

$$y_{m1g} = y_{mg}^0, \quad m=1, \dots, M; g=1, \dots, G$$

$$y_{m(T+1)g} = y_{mg}^{T+1}, \quad m=1, \dots, M; g=1, \dots, G$$

$x, y \geq 0$  and continuous;  $k, z$  binary MILP

# Production and Inventory: Multiple Products

where  $\mathbf{0}_{MTG}$  is a matrix of zeroes and

$M$  = number of production stages

$T$  = number of periods of production

$G$  = number of products produced

$c_{mg}^p$  = production cost of product  $g$  at stage  $m$  (\$/ton)

$x_{mtg}$  = production at stage  $m$  in period  $t$  of product  $g$  (ton)

$c_{mg}^i$  = inventory cost of product  $g$  for stage  $m$  (\$/ton)

$y_{mtg}$  = inventory at stage  $m$  between periods  $t - 1$  and  $t$  of product  $g$  (ton)

$c_{mg}^s$  = stage- $m$  product- $g$  setup cost (\$)

$z_{mtg}$  = setup indicator at stage  $m$  in period  $t$  for product  $g$

$k_{mtg}$  = production indicator at stage  $m$  in period  $t$  for product  $g$

$D_{tg}$  = demand for product  $g$  in period  $t$  (ton)

$K_{mg}$  = capacity for product  $g$  in stage  $m$  (ton)

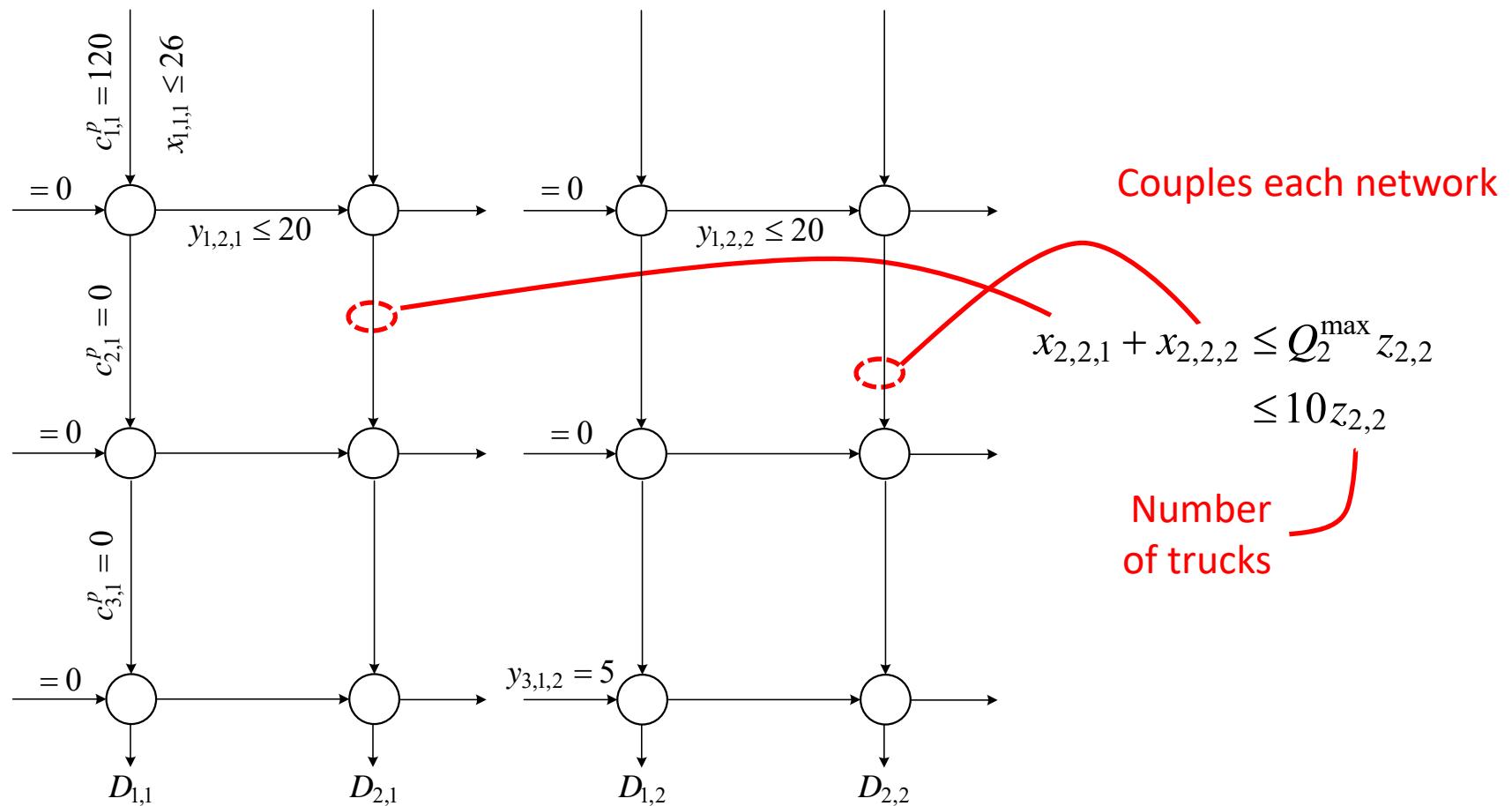
$k_{mg}^0$  = initial setup at stage  $m$  for product  $g$

$y_{mg}^0$  = initial product  $g$  inventory at stage  $m$  (ton)

$y_{mg}^{T+1}$  = final product  $g$  inventory at stage  $m$  (ton)

# Example: HW 6 Problem 4

- Separate networks for each raw material are coupled via sharing the same trucks (added as constraint to model)



# Example: HW 6 Problem 4

- Math programming model:

$$\min \sum_{m=1}^M \sum_{t=1}^T \sum_{g=1}^G c_{mg}^p x_{mtg} + \sum_{m=1}^M \sum_{t=1}^{T+1} \sum_{g=1}^G c_{mg}^i y_{mtg} + \sum_{m=1}^M \sum_{t=1}^T c_m^t z_{mt}$$

subject to

$$x_{mtg} - x_{(m+1)tg} + y_{mtg} - y_{m(t+1)g} = 0, \quad m = 1, \dots, M-1; \quad t = 1, \dots, T; \quad g = 1, \dots, G$$

$$x_{Mtg} + y_{Mtg} - y_{M(t+1)g} = D_{tg}, \quad t = 1, \dots, T; \quad g = 1, \dots, G$$

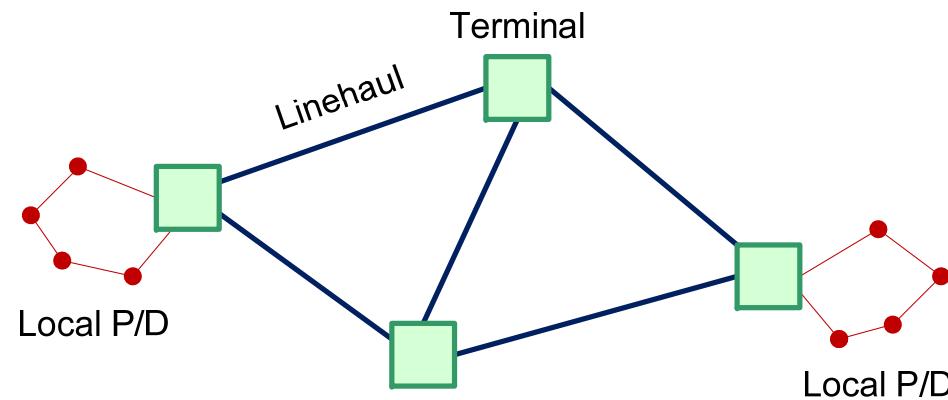
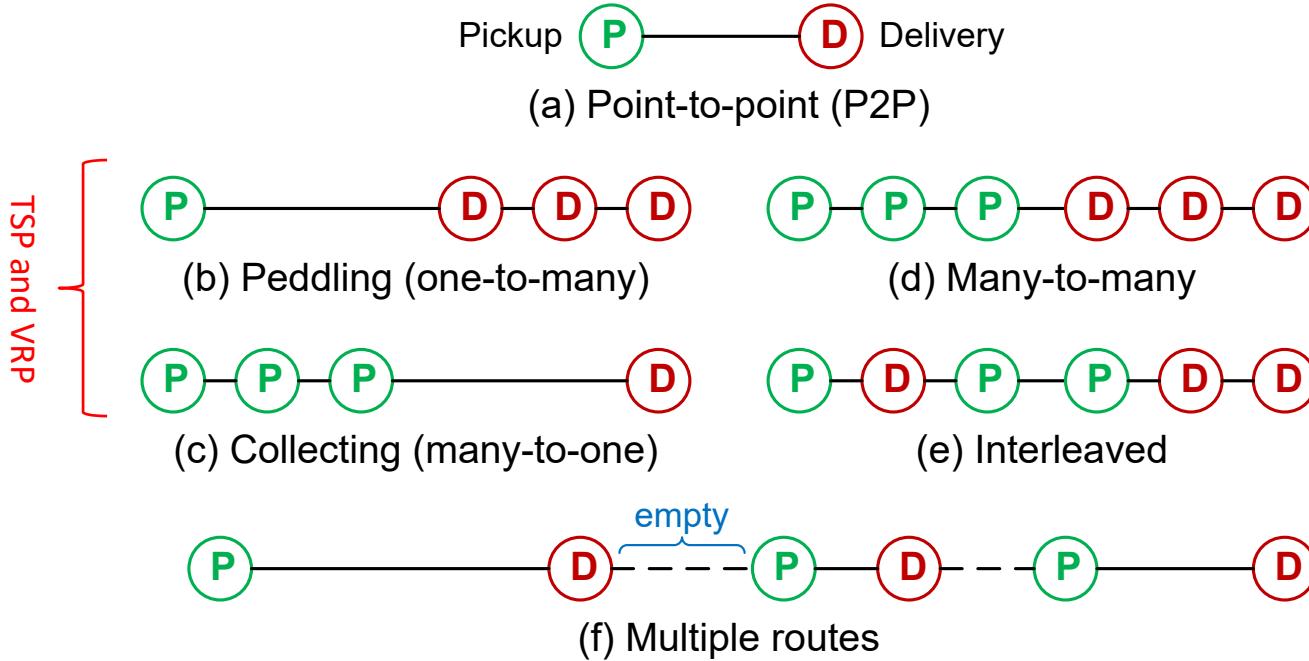
$$\sum_{g=1}^G x_{mtg} \leq Q_m^{\max} z_{mt}, \quad m = 1, \dots, M; \quad t = 1, \dots, T$$

Use LB,UB for capacity constraints

# Topics

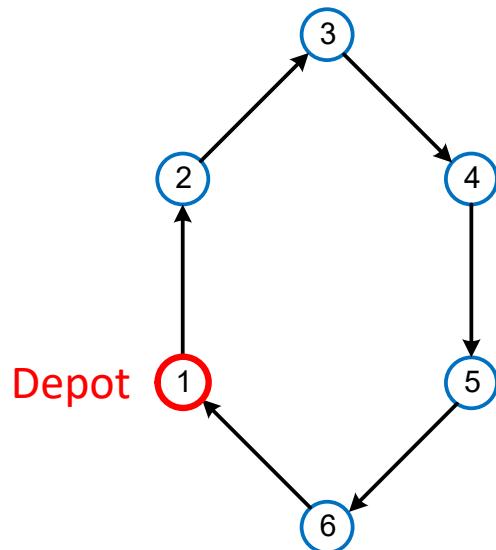
1. Introduction
2. Facility location
3. Freight transport
  - Midterm exam
4. Network models
- 5. Routing**
6. Warehousing
  - Final project
  - Final exam

# Routing Alternatives



# TSP

- Problem: find connected sequence through all nodes of a graph that minimizes total arc cost
  - Subroutine in most vehicle routing problems
  - Node sequence can represent a route only if all pickups and/or deliveries occur at a single node (depot)



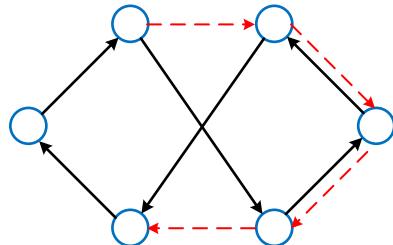
Node sequence = permutation + start node

1	2	3	4	5	6	1
---	---	---	---	---	---	---

$$n = 6 \Rightarrow (n-1)! = 120 \text{ possible solutions}$$

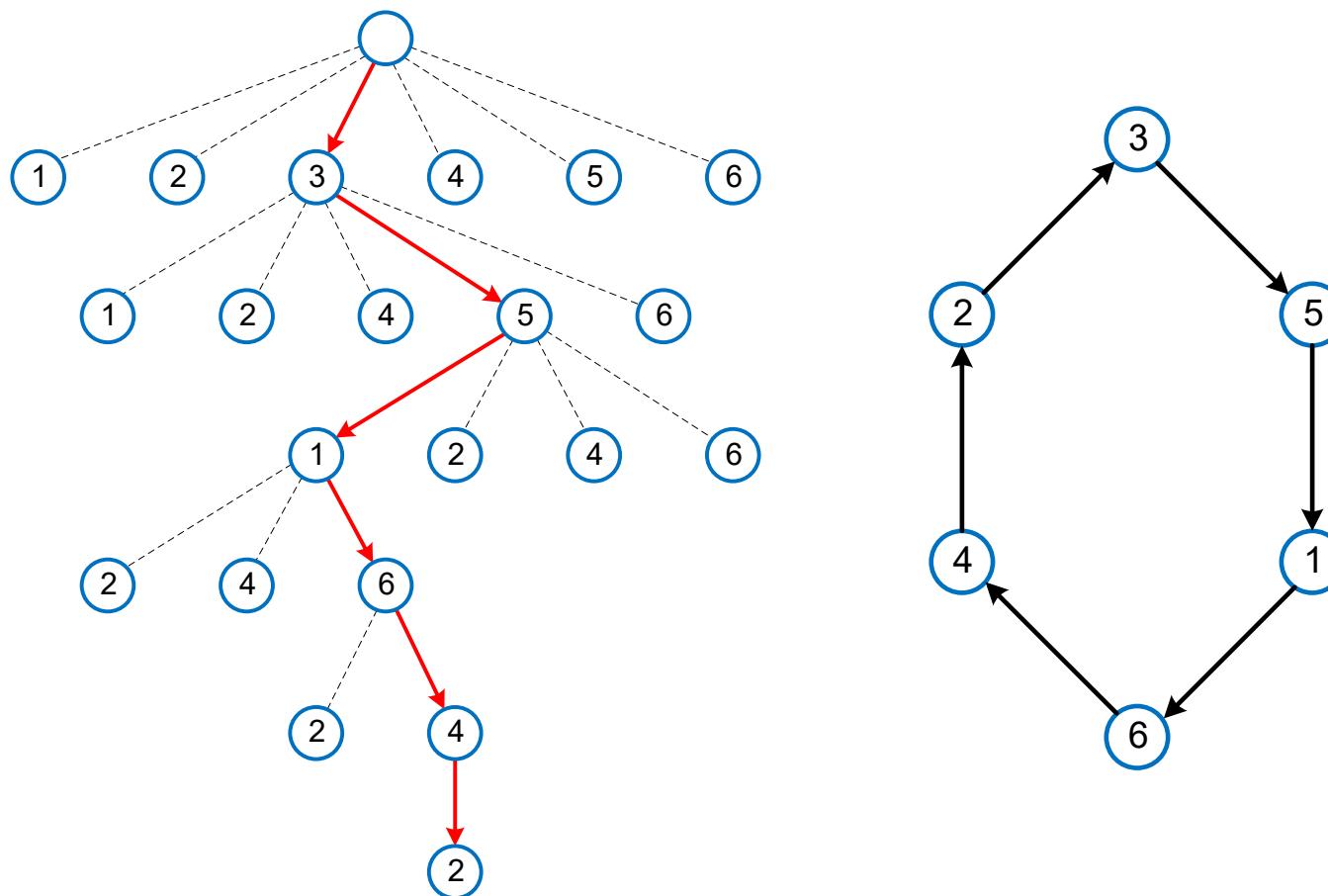
# TSP

- TSP can be solved by a mix of *construction* and *improvement* procedures
  - BIP formulation has an exponential number of constraints to eliminate subtours ( $\Rightarrow$  column generation techniques)
- Asymmetric: only best known solutions for large  $n$   
 $(n-1)! \quad n=13 \Rightarrow \approx \frac{1}{2}$  billion solutions
- Symmetric: solved to optimal using BIP  
 $c_{ij} = c_{ji} \Rightarrow \frac{(n-1)!}{2}$  solutions
- Euclidian: arcs costs = distance between nodes



# TSP Construction

- Construction easy since any permutation is feasible and can then be improved

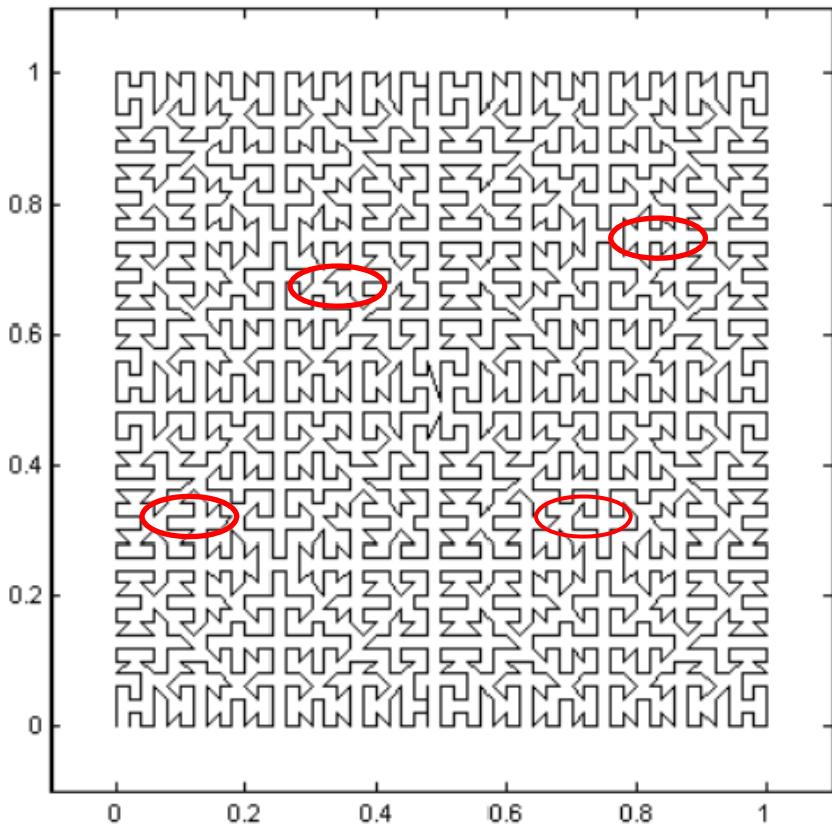


# Spacefilling Curve

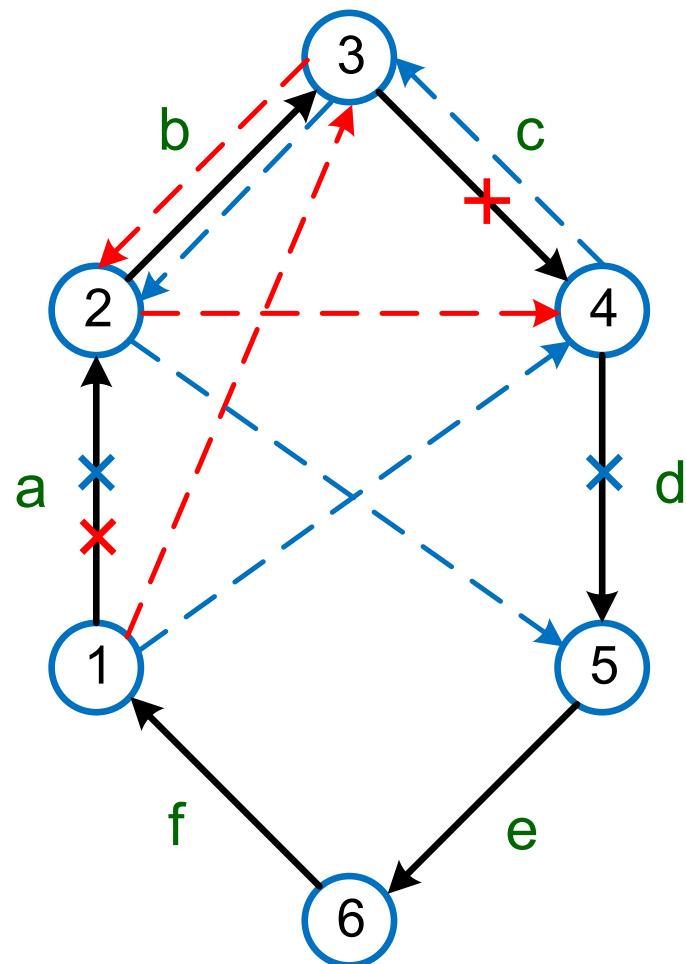
1.0	0.250	0.254	0.265	0.298	0.309	0.438	0.441	0.452	0.485	0.496	0.500
0.9	0.246	0.257	0.271	0.292	0.305	0.434	0.445	0.458	0.479	0.493	0.504
0.8	0.235	0.229	0.279	0.283	0.333	0.423	0.417	0.467	0.471	0.521	0.515
0.7	0.202	0.208	0.158	0.154	0.354	0.390	0.396	0.596	0.592	0.542	0.548
0.6	0.191	0.180	0.167	0.146	0.132	0.379	0.618	0.604	0.583	0.570	0.559
0.5	0.188	0.184	0.173	0.140	0.129	0.375	0.621	0.610	0.577	0.566	0.563
0.4	0.059	0.070	0.083	0.104	0.118	0.871	0.632	0.646	0.667	0.680	0.691
0.3	0.048	0.042	0.092	0.096	0.896	0.860	0.854	0.654	0.658	0.708	0.702
0.2	0.015	0.021	0.971	0.967	0.917	0.827	0.833	0.783	0.779	0.729	0.735
0.1	0.004	0.993	0.979	0.958	0.945	0.816	0.805	0.792	0.771	0.757	0.746
0.0	0.000	0.996	0.985	0.952	0.941	0.813	0.809	0.798	0.765	0.754	0.750
$\theta$	0.0	0.1	0.2	0.3	0.4	0.5	0.6	0.7	0.8	0.9	1.0

Sequence determined by  
sorting position along 1-D  
line covering 2-D space

2:	0.021
3:	0.154
1:	0.471
4:	0.783



# Two-Opt Improvement



	a	b	c	d	e	f	
	1	2	3	4	5	6	1
a-c	1	3	2	4	5	6	1
a-d	1	4	3	2	5	6	1
a-e	1	5	4	3	2	6	1
b-d							
b-e							
b-f							
c-e							
c-f							
d-f							

Sequences considered at end to verify local optimum:  $n$  nodes  $\Rightarrow \sum_{i=1}^{n-2} \sum_{j=i+2}^{n-1} (1) = \frac{n^2 - 5n}{2} + 3 = 9$  for  $n = 6$

# TSP Comparison

	TSP Procedure	Total Cost
1	Spacefilling curve	482.7110
2	1 + 2-opt	456
3	Convex hull insert + 2-opt	452
4	Nearest neighbor + 2-opt	439.6
5	Random construction + 2-opt	450, 456
6	Eil51 in TSPLIP	426* optimal

# Multi-Stop Routing

- Each shipment might have a different origin and/or destination  $\Rightarrow$  node/location sequence not adequate



$$L = (y_1, \dots, y_n) = (1, 2, 3) \quad n \text{ shipments}$$

$$R = (z_1, \dots, z_{2n}) = (3, 1, 2, 2, 1, 3) \quad 2n\text{-element route sequence}$$

$$X = (x_1, \dots, x_{2n}) = (5, 1, 3, 4, 2, 6) \quad 2n\text{-element location (node) sequence}$$

$c_{ij}$  = cost between locations  $i$  and  $j$

$$c(R) = \sum_{i=1}^{2n-1} c_{x_i, x_{i+1}} = 60 + 30 + 250 + 30 + 60 = 430, \quad \text{total cost of route } R$$

# Min Cost Insert

	1	1				
1	•		•	2	2	x
2	2	•	•	2		$c_2$
3		•	2	2	•	$c_3^*$
4		•	2	•	2	$c_4$
5	2	•	2	•		$c_5$

	1	2	2	1	
1	3	•	3	•	•
2	3	•	•	3	•
3	3	•	•	•	3
4	3	•	•	•	• 3
5		•	3	3	•
6		•	3	• 3	•
7		•	3	•	• 3
:	:		:		:

# Route Sequencing Procedures

- **Online** procedure: add a shipment to an existing route as it becomes available
  - Insert and Improve: for each shipment,  $\text{mincostinsert} + \text{twoopt}$
- **Offline** procedure: consider all shipments to decide order in which each added to route
  - Savings and Improve: determine insert ordering based on “savings,” then improve final route

# Insert and Improve Online Procedure

```

procedure insertImprove( $y_i \in L$ )
     $R = (y_1, y_1)$ 
    for  $i = 2, \dots, |L|$ 
         $R = \text{minCostInsert}(y_i, R)$ 
         $R = \text{twoOpt}(R)$ 
    endfor
    return  $R$ 

```

```

subprocedure minCostInsert( $y, z_i \in R$ )
     $c_R = c(R)$ 
    for  $i = 1, \dots, |R| + 1$ , for  $j = 1, \dots, |R| + 1$ 
         $R' = (z_1, \dots, z_{i-1}, y, z_i, \dots, z_{j-1}, y, z_j, \dots, z_{|R|})$ 
        if  $c(R') < c_R$ ,  $c_R = c(R')$ ,  $R = R'$ , endif
    endfor, endfor
    return  $R$ 

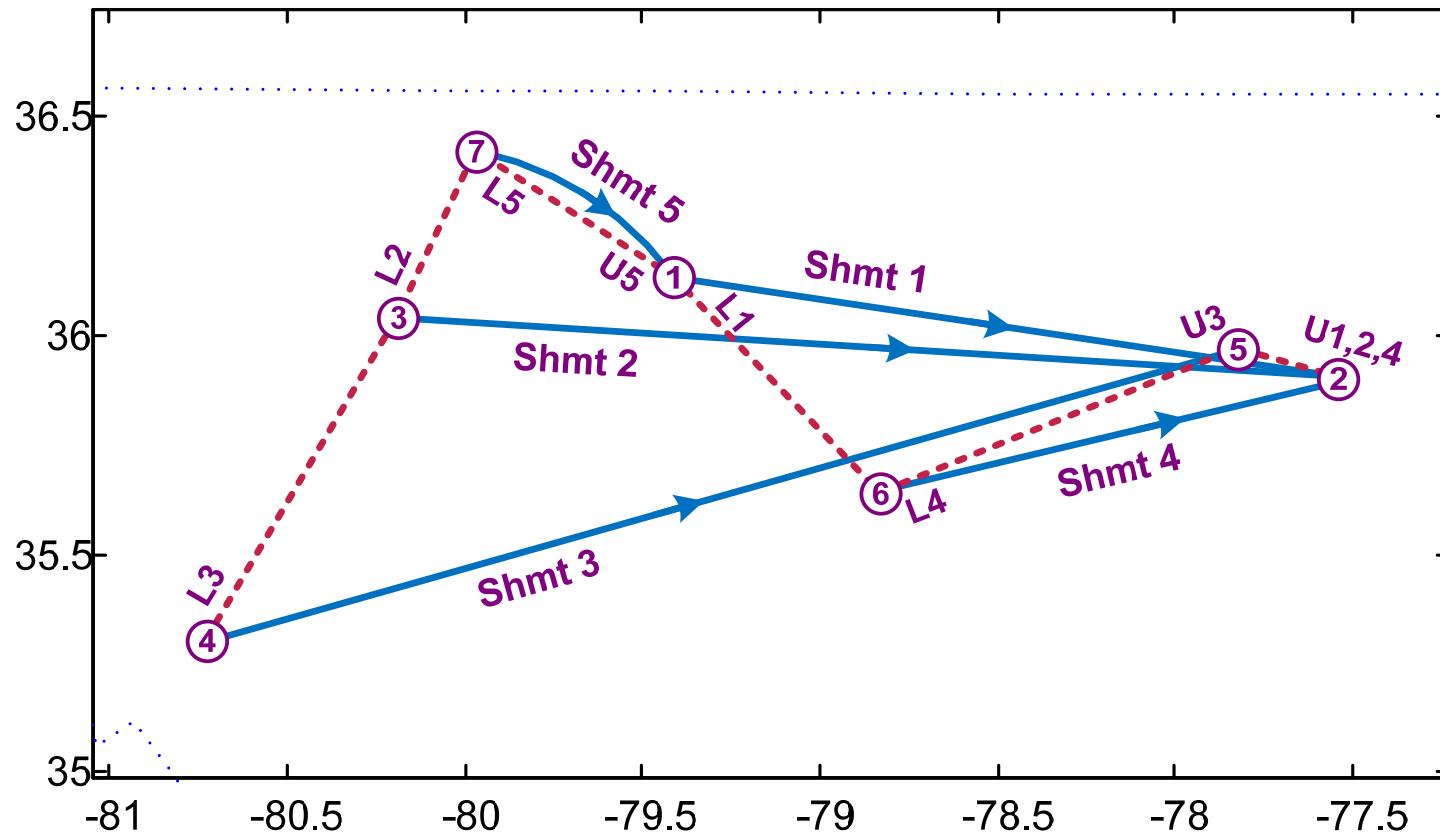
```

```

subprocedure twoOpt( $z_i \in R$ )
     $c_R = c(R)$ 
    repeat
         $done = \text{true}$ ,  $i = 1$ ,  $j = 2$ 
        while  $done$  and  $i < |R|$ 
            while  $done$  and  $j < |R| + 1$ 
                 $R' = (z_1, \dots, z_{i-1}, \text{reverseSequence}(z_i, \dots, z_j), z_{j+1}, \dots, z_{|R|})$ 
                if  $c(R') < c_R$ 
                     $c_R = c(R')$ ,  $R = R'$ ,  $done = \text{false}$ 
                endif
                 $j = j + 1$ 
            endwhile
             $i = i + 1$ ,  $j = i + 1$ 
        endwhile
    until  $done = \text{true}$ 
    return  $R$ 

```

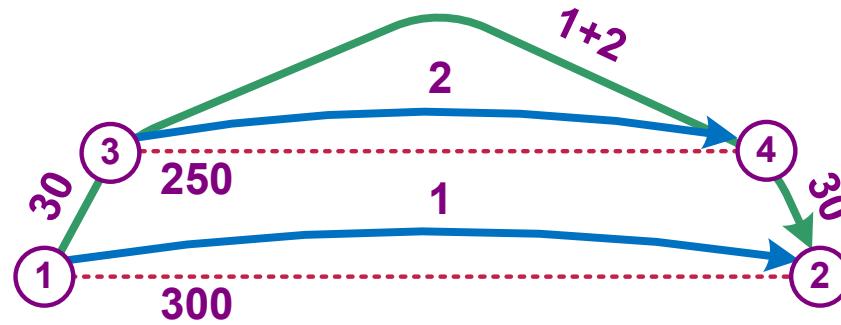
# 5-Shipment Example



$$R = (3, 2, 5, 5, 1, 4, 3, 1, 2, 4)$$

$$X = (4, 3, 7, 1, 1, 6, 5, 2, 2, 2)$$

# Pairwise Savings



$s_{ij}$  = pairwise savings between shipments  $i$  and  $j$

$$= c_i + c_j - c_{ij}$$

$$s_{1,2} = 300 + 250 - 310$$

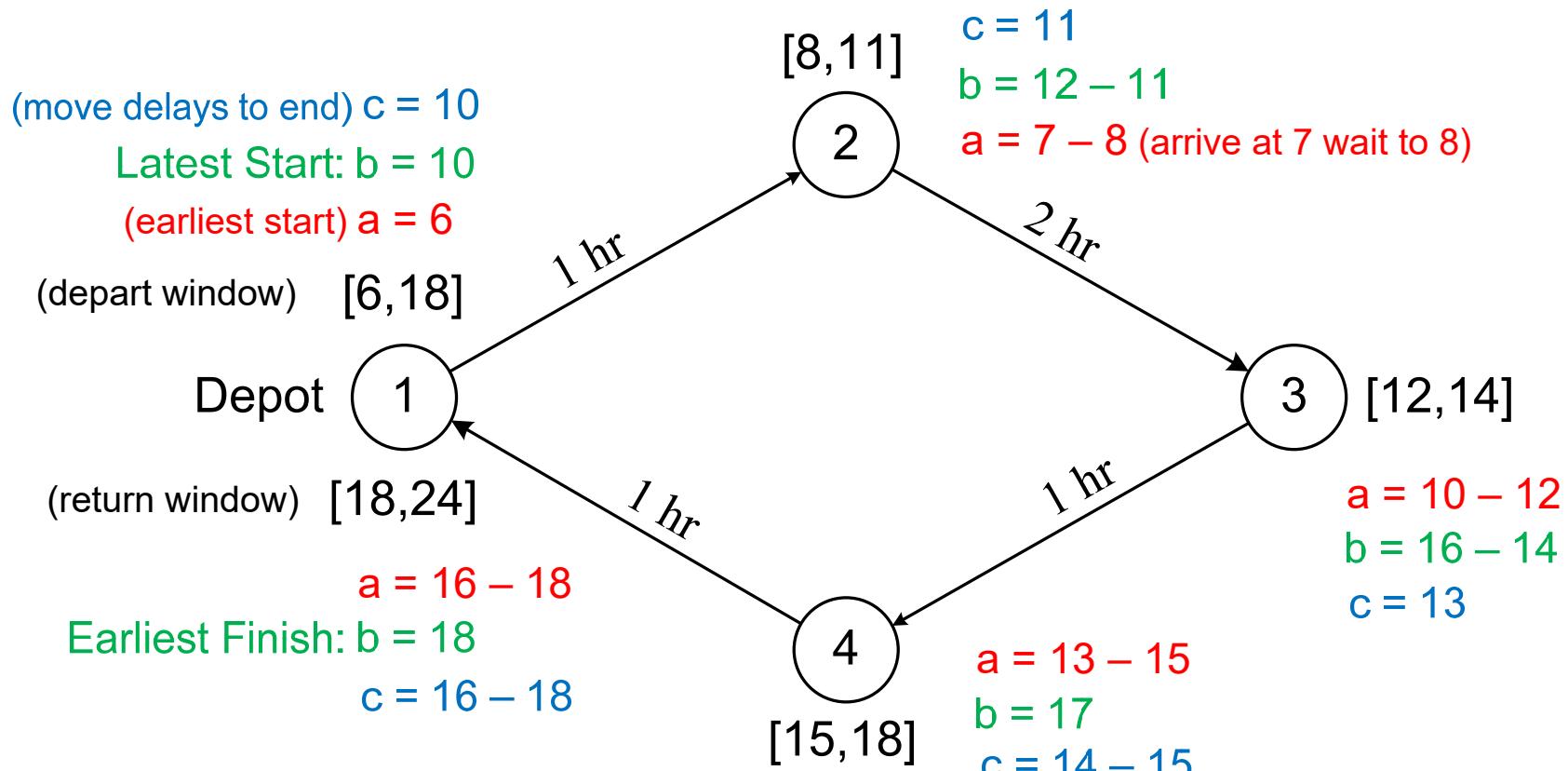
$$= 240$$

# Vehicle Routing Problem

- VRP = TSP + vehicle constraints
- Constraints:
  - Capacity (weight, cube, etc.)
  - Maximum TC (HOS: 11 hr max)
  - Time windows (with/without delay at customer)
    - VRP uses absolute windows that can be checked by simple scanning
    - Project scheduling uses relative windows solved by shortest path with negative arcs
  - Maximum number of routes/vehicles (hard)
- Criteria:
  1. Number of routes/vehicles
  2. TC

# Time Window Example

[0,24] hr; Loading/unloading time = 0; Capacity =  $\infty$ ; LB = 5 hr



Earliest Finish – Latest Start =  $18 - 10 = 8$  hr = 5 travel + 3 delay

# Savings Route Construction Procedure

- Pairs of shipments are ordered in terms of their decreasing pairwise savings to create  $\mathbf{i}$  and  $\mathbf{j}$
- Creates set of multi-shipment routes

$$R = \{R_1, \dots, R_m\}$$

- Shipments with no pairwise savings are not included (use sh2rte to add)
- Given savings pair  $i, j$ , the original Clark-Wright savings procedure (vrpsavings):
  - adds  $i$  to route only if  $j$  at beginning or end of route
  - routes combined only if  $i$  and  $j$  are endpoints of each route

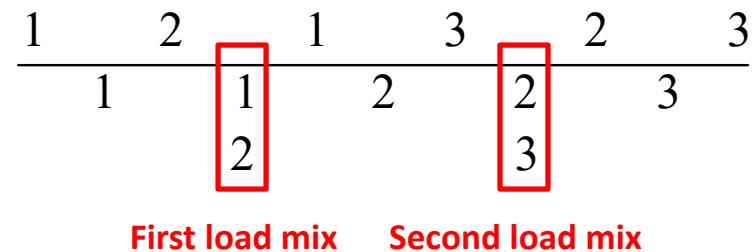
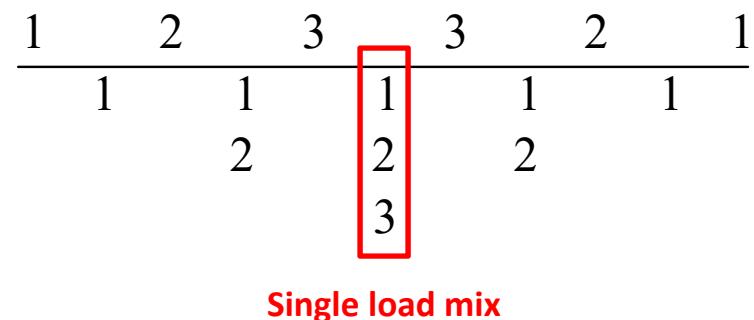
```

procedure savings( $\mathbf{i}, \mathbf{j}$ )
   $R \leftarrow \{\}$ 
  for  $k = \{1, \dots, |\mathbf{i}|\}$ 
    if  $i_k \notin R$  and  $j_k \notin R$    1. Form new route
       $R \leftarrow R \cup minCostInsert(i_k, j_k)$ 
    elseif ( $i_k \notin R$  and  $j_k \in R$ ) or ( $i_k \in R$  and  $j_k \notin R$ )
      if  $j_k \notin R$    2. Add shipment to route
         $temp \leftarrow i_k, i_k \leftarrow j_k, j_k \leftarrow temp$ 
      endif
       $h \leftarrow \arg\{R_l : j_k \in R_l\}$ 
       $R' \leftarrow minCostInsert(i_k, R_h)$ 
      if  $c(R') < c(i_k) + c(R_h)$ 
         $R_h \leftarrow R'$ 
      endif
    else   3. Combine two routes
       $g \leftarrow \arg\{R_l : i_k \in R_l\}, h \leftarrow \arg\{R_l : j_k \in R_l\}$ 
      if  $g \neq h$ 
         $R' \leftarrow minCostInsert(R_g, R_h)$ 
        if  $c(R') < c(R_g) + c(R_h)$ 
           $R_g \leftarrow \{\}, R_h \leftarrow R'$ 
        endif
      endif
    endif
  endfor
  return  $R$ 

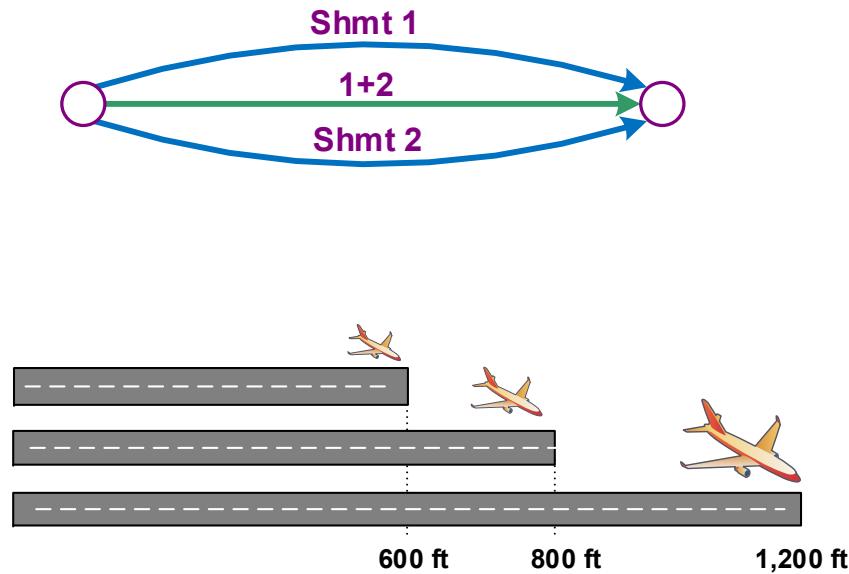
```

# Min TLC Multi-Stop Routing

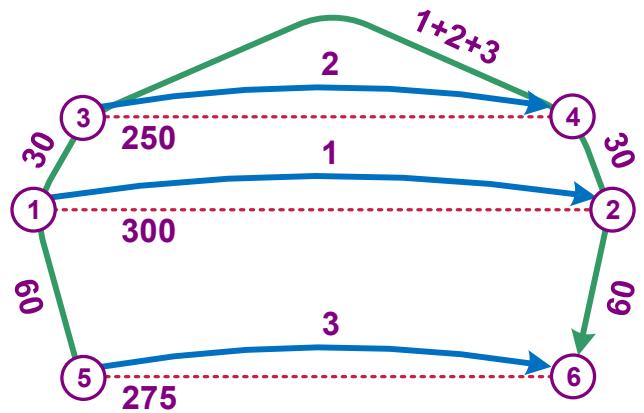
- Periodic consolidated shipments that have the same frequency/interval
  - Similar to a “milk run”
- Min TLC of aggregate shipment may not be feasible
  - Different combinations of shipments (*load mix*) may be on board during each segment of route
  - Each load mix may have different max payload
  - If needed, reduce all shipment sizes in proportion to load mix with the minimum max payload



# Cost Allocation for Routing



# Cost Allocation for Routing



# Shapley Value

$$c_L^{\text{sav}} = \sum_{i=1}^n c_i^0 - c_L$$

$$c_{ij}^{\text{sav}} = c_i^0 + c_j^0 - c_{(i,j)}$$

$$\alpha_i = \sum_{0 \leq m \leq n-1} \frac{m!(n-m-1)!}{n!} \sum_{\substack{M \subset N \setminus i \\ |M|=m}} (\sigma_{M \cup \{i\}} - \sigma_M)$$

$$c_i^{\text{sav}} = \frac{c_L^{\text{sav}}}{n} + \frac{1}{n-1} \sum_{j=1}^n \frac{c_{ij}^{\text{sav}} + c_{ji}^{\text{sav}}}{2} - \frac{1}{n(n-1)} \sum_{j=1}^n \sum_{k=1}^n c_{jk}^{\text{sav}}$$

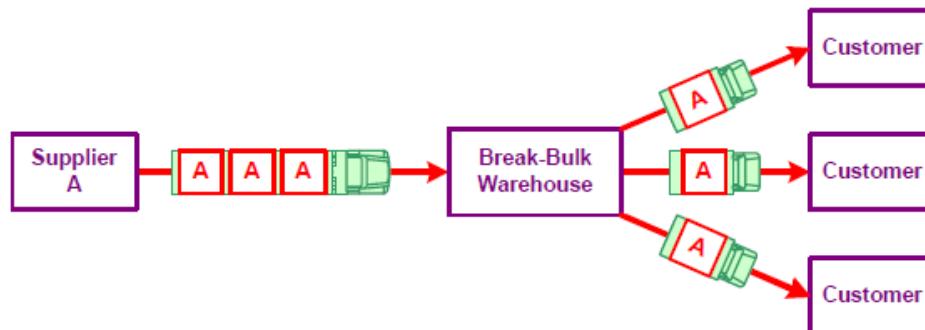
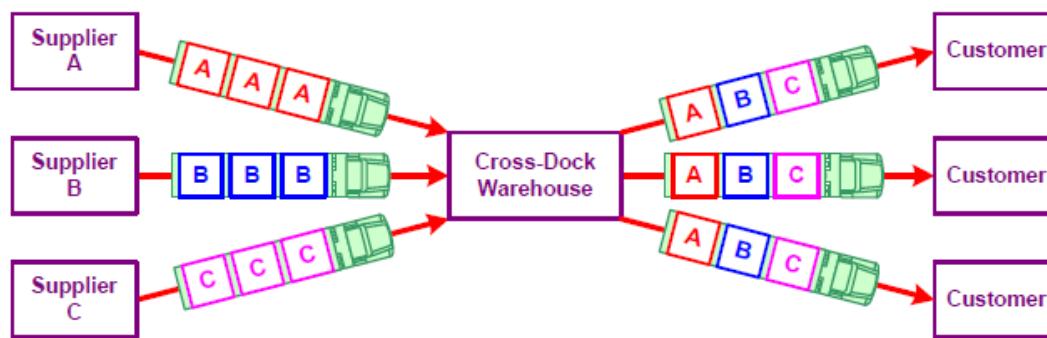
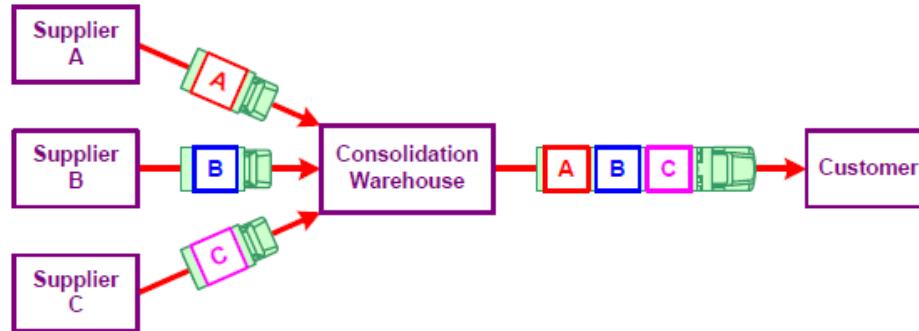
# Topics

1. Introduction
2. Facility location
3. Freight transport
  - Midterm exam
4. Network models
5. Routing
6. **Warehousing**
  - Final project
  - Final exam

# Warehousing

- *Warehousing* are the activities involved in the design and operation of warehouses
- A *warehouse* is the point in the supply chain where raw materials, work-in-process (WIP), or finished goods are stored for varying lengths of time.
- Warehouses can be used to add value to a supply chain in two basic ways:
  1. Storage. Allows product to be available where and when its needed.
  2. Transport Economies. Allows product to be collected, sorted, and distributed efficiently.
- A *public warehouse* is a business that rents storage space to other firms on a month-to-month basis. They are often used by firms to supplement their own *private warehouses*.

# Types of Warehouses



# Warehouse Design Process

- The objectives for warehouse design can include:
  - maximizing cube utilization
  - minimizing total storage costs (including building, equipment, and labor costs)
  - achieving the required storage throughput
  - enabling efficient order picking
- In planning a storage layout: either a storage layout is required to fit into an existing facility, or the facility will be designed to accommodate the storage layout.

# Warehouse Design Elements

- The design of a new warehouse includes the following elements:
  1. Determining the layout of the storage locations (i.e., the warehouse layout).
  2. Determining the number and location of the input/output (I/O) ports (e.g., the shipping/receiving docks).
  3. Assigning items (stock-keeping units or *SKUs*) to storage locations (*slots*).
- A typical objective in warehouse design is to minimize the overall storage cost while providing the required levels of service.

# Design Trade-Off

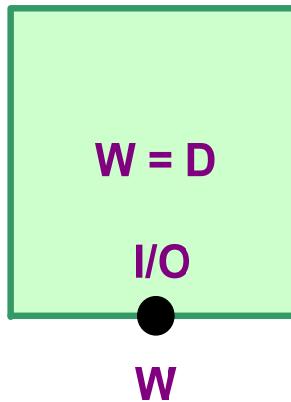
- Warehouse design involves the trade-off between building and handling costs:

$\min \text{ Building Costs}$     vs.     $\min \text{ Handling Costs}$



$\max \text{ Cube Utilization}$     vs.     $\max \text{ Material Accessibility}$

# Shape Trade-Off



D

VS.



D

Square shape minimizes perimeter length for a given area, thus minimizing building costs

Aspect ratio of 2 ( $W = 2D$ ) min. expected distance from I/O port to slots, thus minimizing handling costs

# Storage Trade-Off

B	C	E
A	B	D
A	B	C

vs.

	B		Honeycomb loss	
A	B	C		
A	B	C	D	E

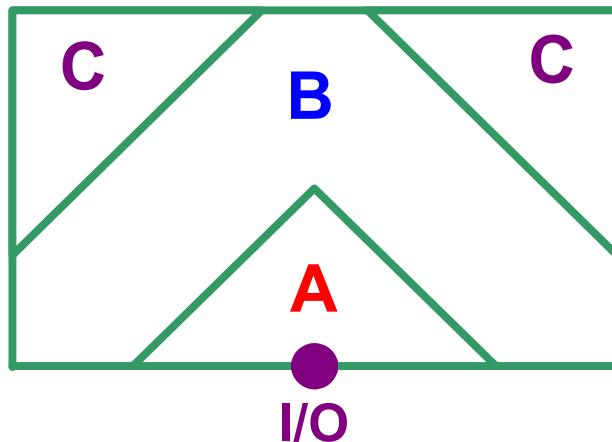
Maximizes cube utilization,  
but minimizes material  
accessibility

Making at least one unit of  
each item accessible  
decreases cube utilization

# Storage Policies

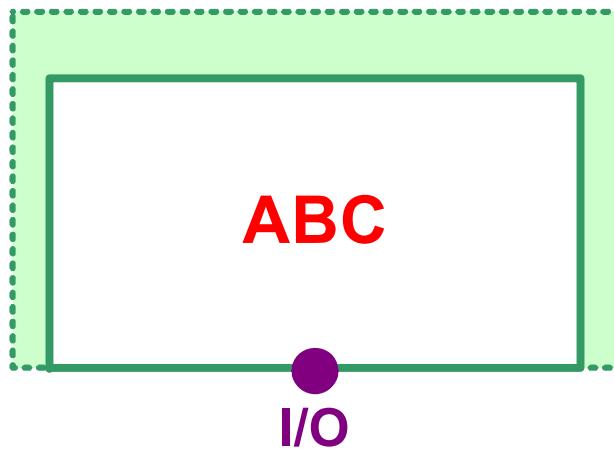
- A storage policy determines how the slots in a storage region are assigned to the different SKUs to be stored in the region.
- The differences between storage policies illustrate the trade-off between minimizing building cost and minimizing handling cost.
- Type of policies:
  - Dedicated
  - Randomized
  - Class-based

# Dedicated Storage



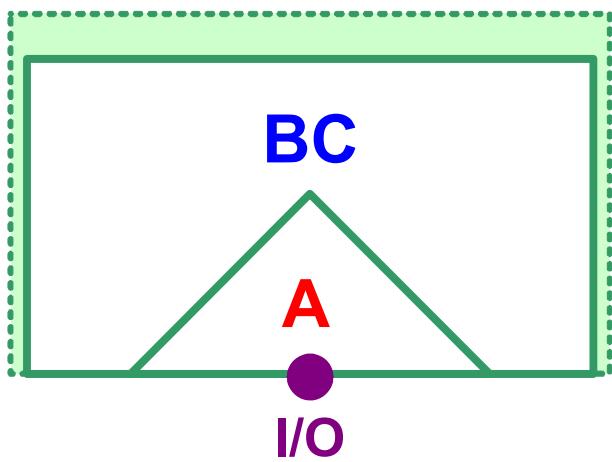
- Each SKU has a predetermined number of slots assigned to it.
- Total capacity of the slots assigned to each SKU must equal the storage space corresponding to the maximum inventory level of each *individual* SKU.
- Minimizes handling cost.
- Maximizes building cost.

# Randomized Storage



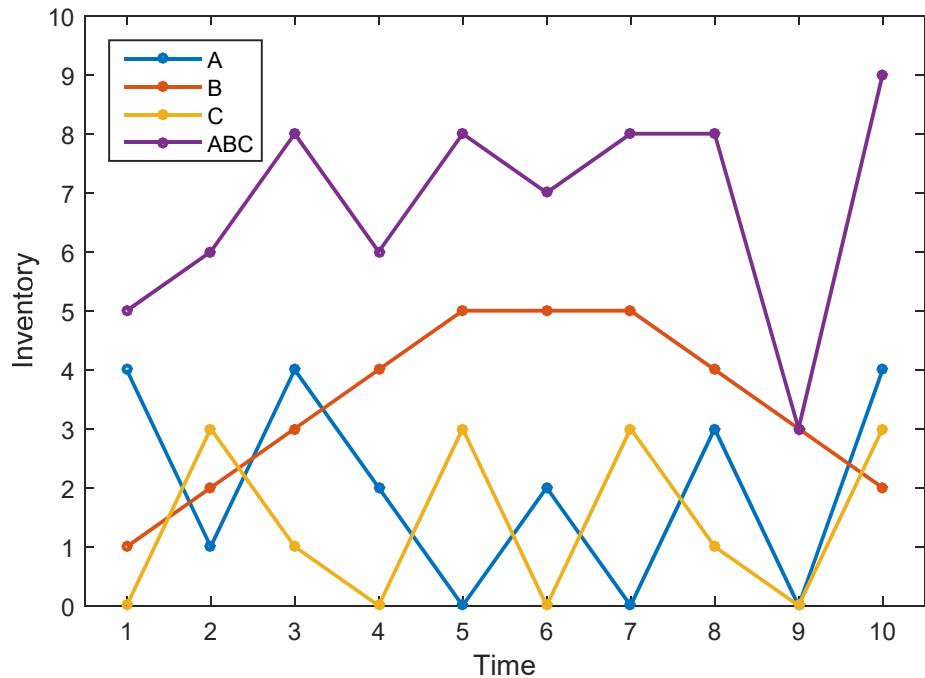
- Each SKU can be stored in any available slot.
- Total capacity of all the slots must equal the storage space corresponding to the maximum *aggregate* inventory level of all of the SKUs.
- Maximizes handling cost.
- Minimizes building cost.

# Class-based Storage



- Combination of dedicated and randomized storage, where each SKU is assigned to one of several different storage classes.
- Randomized storage is used for each SKU within a class, and dedicated storage is used between classes.
- Building and handling costs between dedicated and randomized.

# Individual vs Aggregate SKUs



Time	Dedicated			Random		Class-Based	
	A	B	C	ABC	AB	AC	BC
1	4	1	0	5	5	4	1
2	1	2	3	6	3	4	5
3	4	3	1	8	7	5	4
4	2	4	0	6	6	2	4
5	0	5	3	8	5	3	8
6	2	5	0	7	7	2	5
7	0	5	3	8	5	3	8
8	3	4	1	8	7	4	5
9	0	3	0	3	3	0	3
10	4	2	3	9	6	7	5

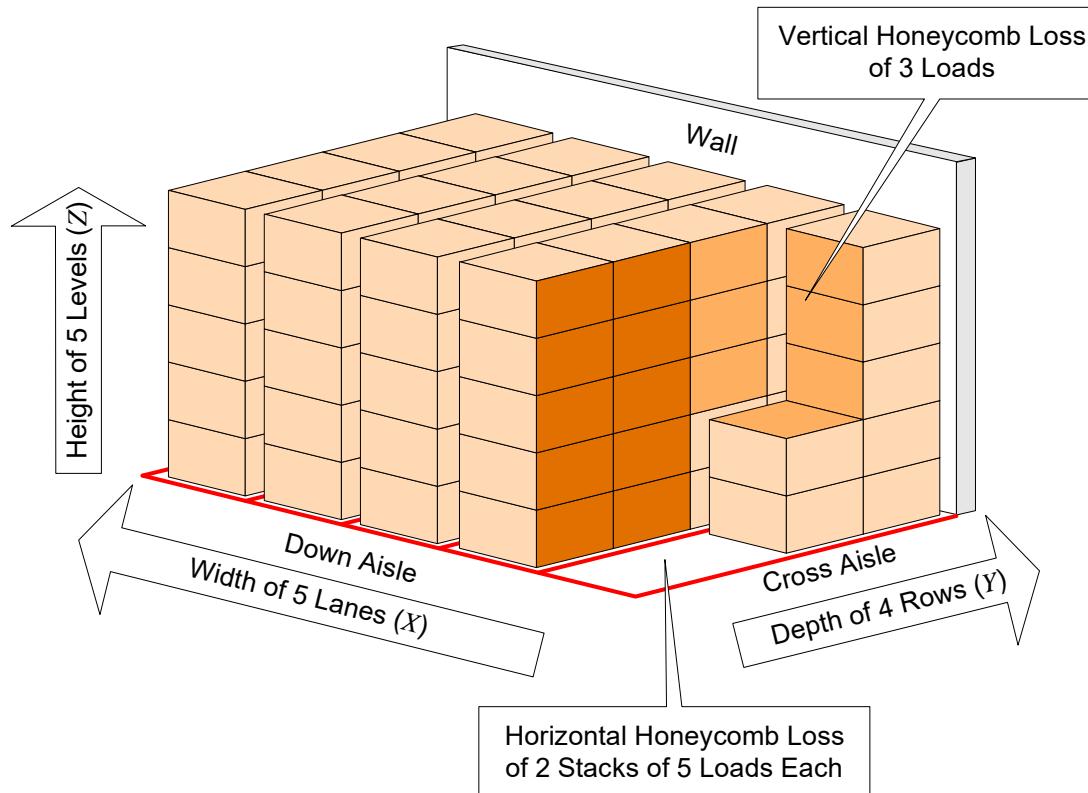
$M_i$	4	5	3	9	7	7	8
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# Cube Utilization

- *Cube utilization* is percentage of the total space (or “cube”) required for storage actually occupied by items being stored.
- There is usually a trade-off between cube utilization and material accessibility.
- Bulk storage using block stacking can result in the minimum cost of storage, but material accessibility is low since only the top of the front stack is accessible.
- Storage racks are used when support and/or material accessibility is required.

# Honeycomb Loss

- *Honeycomb loss*, the price paid for accessibility, is the unusable empty storage space in a lane or stack due to the storage of only a single SKU in each lane or stack



# Estimating Cube Utilization

- The (3-D) cube utilization for dedicated and randomized storage can estimated as follows:

$$\text{Cube utilization} = \frac{\text{item space}}{\text{total space}} = \frac{\text{item space}}{\text{item space} + \left( \begin{array}{c} \text{honeycomb} \\ \text{loss} \end{array} \right) + \left( \begin{array}{c} \text{down aisle} \\ \text{space} \end{array} \right)}$$

$$CU(3\text{-D}) = \begin{cases} \frac{x \cdot y \cdot z \cdot \sum_{i=1}^N M_i}{TS(D)}, & \text{dedicated} \\ \frac{x \cdot y \cdot z \cdot M}{TS(D)}, & \text{randomized} \end{cases}$$

where

$x$  = lane/unit-load width

$y$  = unit-load depth

$z$  = unit-load height

$M_i$  = maximum number of units of SKU  $i$

$M$  = maximum number of units of all SKUs

$N$  = number of different SKUs

$D$  = number of rows

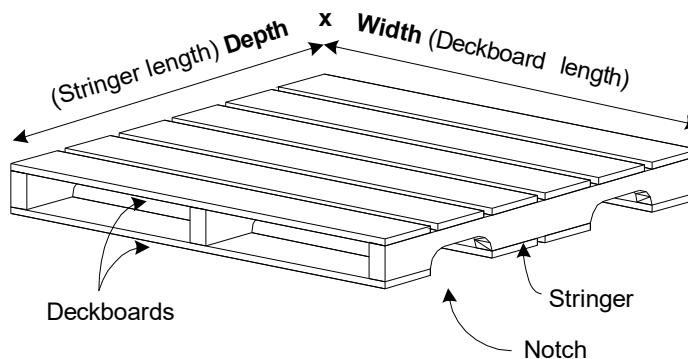
$TS(D)$  = total 3-D space (given  $D$  rows of storage).

$TA(D)$  = total 2-D area (given  $D$  rows of storage).

$$CU(2\text{-D}) = \begin{cases} \frac{x \cdot y \cdot \sum_{i=1}^N \left[ \frac{M_i}{H} \right]}{TA(D)}, & \text{dedicated} \\ \frac{x \cdot y \cdot \left[ \frac{M}{H} \right]}{TA(D)}, & \text{randomized} \end{cases}$$

# Unit Load

- *Unit load*: single unit of an item, or multiple units restricted to maintain their integrity
- Linear dimensions of a unit load:



*Depth (stringer length) × Width (deckboard length)*

$$y \times x$$

- Pallet height (5 in.) + load height gives  $z$ :  $y \times x \times z$

# Cube Utilization for Dedicated Storage

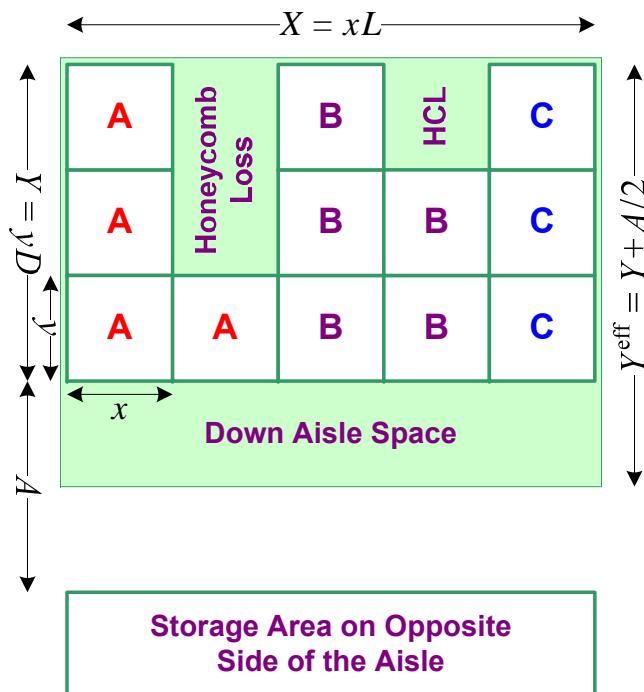
Storage Area at Different Lane Depths												Item Space	Lanes	Total Space	Cube Util.
$D = 1$	$A/2 = 1$	A	A	A	A	B	B	B	B	C	C	12	12	24	50%
$D = 2$	$A/2 = 1$	A	A	B	B		C					12	7	21	57%
		A	A	B	B	B	C	C							
$D = 3$	$A/2 = 1$	A		B		C						12	5	20	60%
		A		B	B	C									

# Total Space/Area

- The total space required, as a function of lane depth  $D$ :

Total space (3-D):  $TS(D) = X \cdot \underbrace{\left( Y + \frac{A}{2} \right)}_{\text{Eff. lane depth}} \cdot Z = xL(D) \cdot \left( yD + \frac{A}{2} \right) \cdot zH$

Total area (2-D):  $TA(D) = \frac{TS(D)}{Z} = X \cdot Y^{\text{eff}} = xL(D) \cdot \left( yD + \frac{A}{2} \right)$



where

$X$  = width of storage region (row length)

$Y$  = depth of storage region (lane depth)

$Z$  = height of storage region (stack height)

$A$  = down aisle width

$L(D)$  = number of lanes (given  $D$  rows of storage)

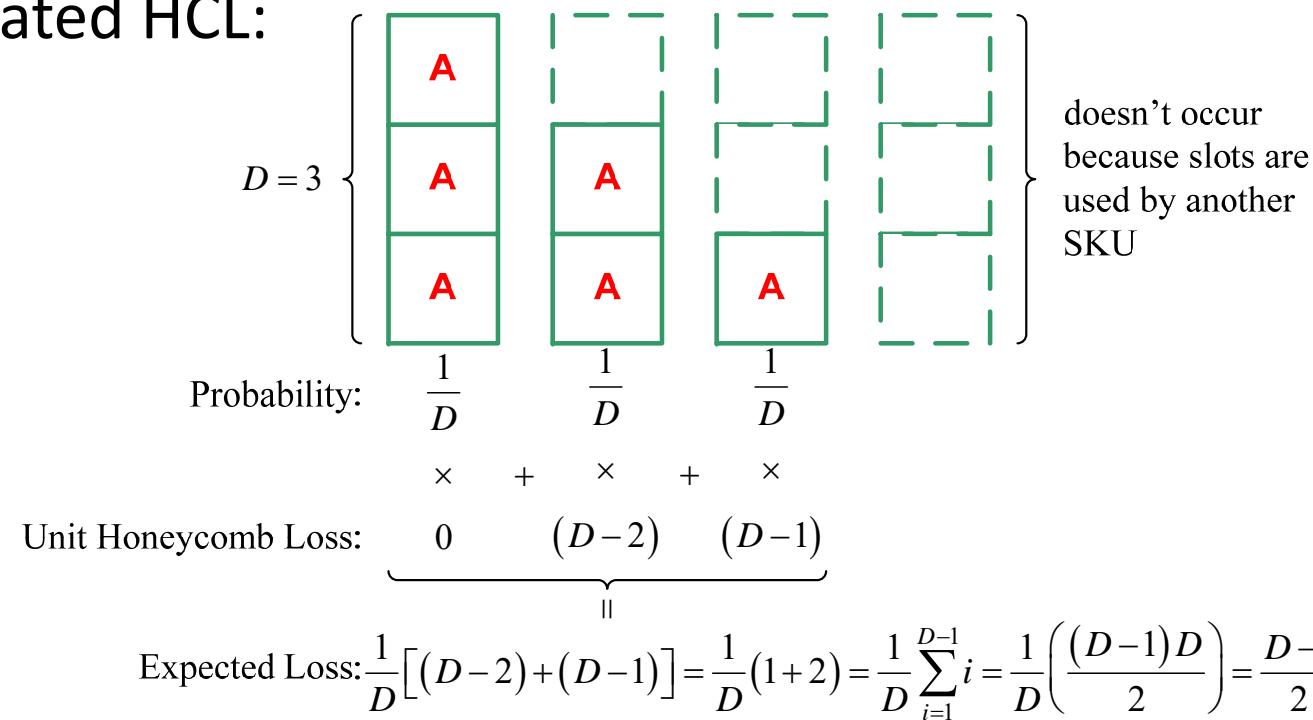
$H$  = number of levels.

# Number of Lanes

- Given  $D$ , estimated total number of lanes in region:

$$\text{Number of lanes: } L(D) = \begin{cases} \sum_{i=1}^N \left\lceil \frac{M_i}{DH} \right\rceil, & \text{dedicated} \\ \left\lceil \frac{M + NH \left( \frac{D-1}{2} \right) + N \left( \frac{H-1}{2} \right)}{DH} \right\rceil, & \text{randomized } (N > 1) \end{cases}$$

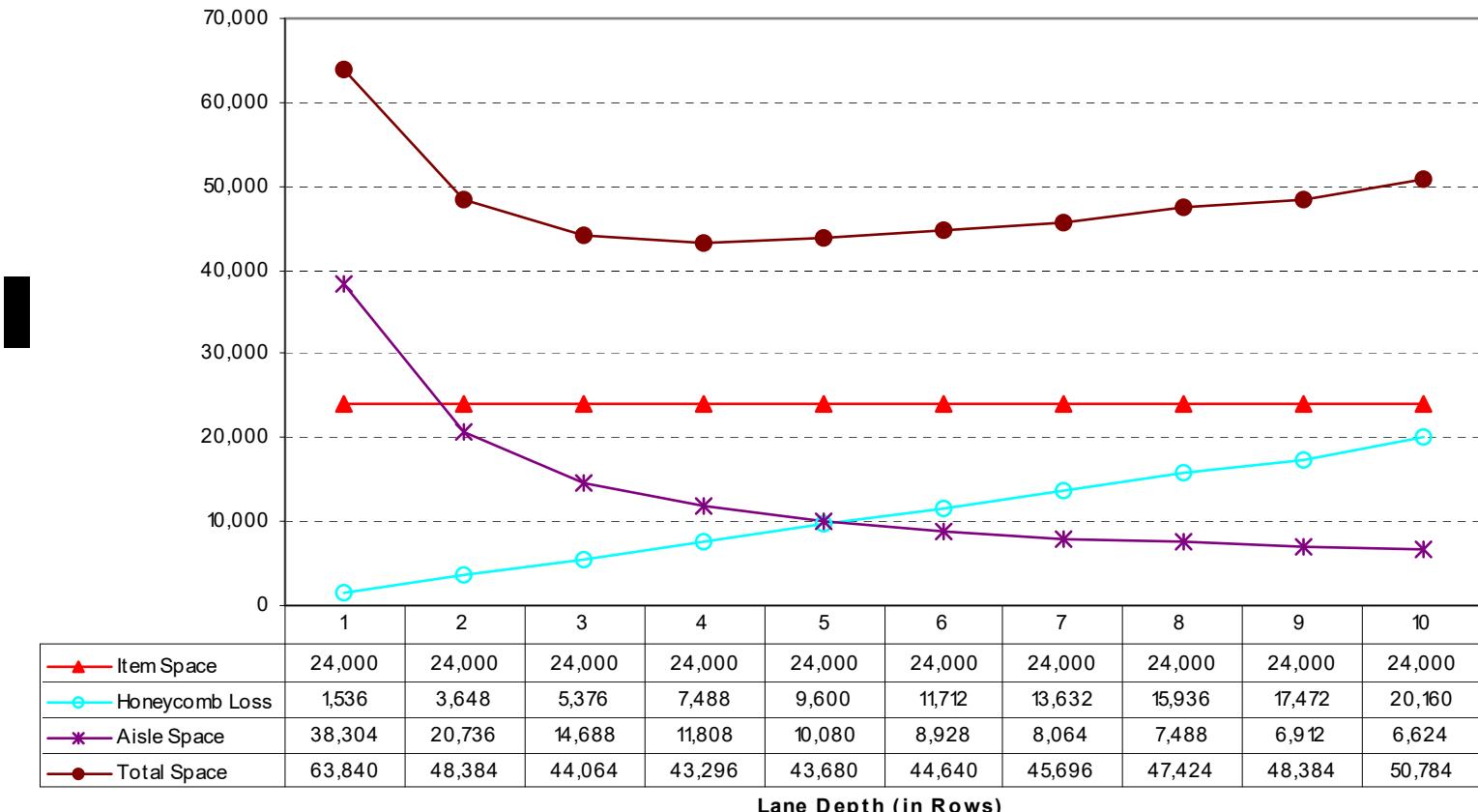
- Estimated HCL:



# Optimal Lane Depth

- Solving for  $D$  in  $dTS(D)/dD = 0$  results in:

Optimal lane depth for randomized storage (in rows):  $D^* = \left\lfloor \sqrt{\frac{A(2M - N)}{2NyH}} + \frac{1}{2} \right\rfloor$



# Max Aggregate Inventory Level

- Usually can determine max inventory level for each SKU:
  - $M_i$  = maximum number of units of SKU  $i$
- Since usually don't know  $M$  directly, but can estimate it **if**
  - SKUs' inventory levels are uncorrelated
  - Units of each item are either stored or retrieved at a constant rate

$$M = \left\lfloor \sum_{i=1}^N \frac{M_i}{2} + \frac{1}{2} \right\rfloor$$

- Can add include safety stock for each item,  $SS_i$ 
  - For example, if the order size of three SKUs is 50 units and 5 units of each item are held as safety stock

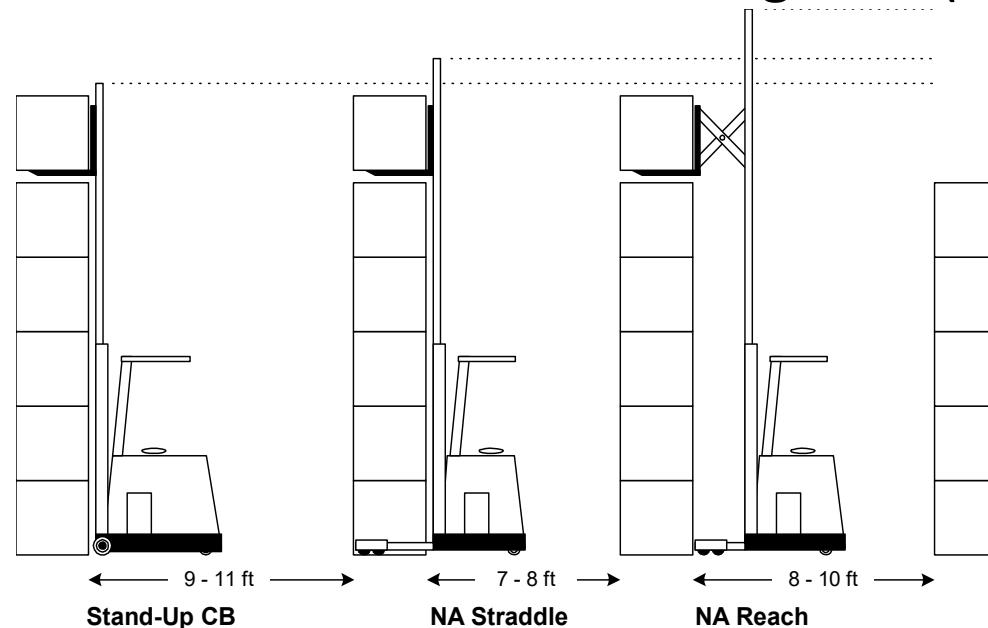
$$M = \left\lfloor \sum_{i=1}^N \left( \frac{M_i - SS_i}{2} + SS_i \right) + \frac{1}{2} \right\rfloor = \left\lfloor 3 \left( \frac{50}{2} + 5 \right) + \frac{1}{2} \right\rfloor = 90$$

# Steps to Determine Area Requirements

1. For randomized storage, assumed to know  $N, H, x, y, z, A$ , and all  $M_i$ 
  - Number of levels,  $H$ , depends on building clear height (for block stacking) or shelf spacing
  - Aisle width,  $A$ , depends on type of lift trucks used
2. Estimate maximum aggregate inventory level,  $M$
3. If  $D$  not fixed, estimate optimal land depth,  $D^*$
4. Estimate number of lanes required,  $L(D^*)$
5. Determine total 2-D area,  $TA(D^*)$

# Aisle Width Design Parameter

- Typically,  $A$  (and sometimes  $H$ ) is a parameter used to evaluate different overall design alternatives
- Width depends on type of lift trucks used, a narrower aisle truck
  - reduces area requirements (building costs)
  - costs more and slows travel and loading time (handling costs)



# Example 1: Area Requirements

Units of items A, B, and C are all received and stored as  $42 \times 36 \times 36$  in. ( $y \times x \times z$ ) pallet loads in a storage region that is along one side of a 10-foot-wide down aisle in the warehouse of a factory. The shipment size received for each item is 31, 62, and 42 pallets, respectively. Pallets can be stored up to three deep and four high in the region.

$$x = \frac{36}{12} = 3' \quad M_A = 31 \quad A = 10'$$

$$y = 3.5' \quad M_B = 62 \quad D = 3$$

$$z = 3' \quad M_C = 42 \quad H = 4$$

$$N = 3$$

# Example 1: Area Requirements

1. If a dedicated policy is used to store the items, what is the 2-D cube utilization of this storage region?

$$L(D) = L(3) = \sum_{i=1}^N \left\lceil \frac{M_i}{DH} \right\rceil = \left\lceil \frac{31}{3(4)} \right\rceil + \left\lceil \frac{62}{3(4)} \right\rceil + \left\lceil \frac{42}{3(4)} \right\rceil = 3 + 6 + 4 = 13 \text{ lanes}$$

$$TA(3) = xL(D) \cdot \left( yD + \frac{A}{2} \right) = 3(13) \cdot \left( 3.5(3) + \frac{10}{2} \right) = 605 \text{ ft}^2$$

$$CU(3) = \frac{\text{item space}}{TA(3)} = \frac{x \cdot y \cdot \sum_{i=1}^N \left\lceil \frac{M_i}{H} \right\rceil}{TA(3)} = \frac{3 \cdot 3.5 \cdot \left( \left\lceil \frac{31}{4} \right\rceil + \left\lceil \frac{62}{4} \right\rceil + \left\lceil \frac{42}{4} \right\rceil \right)}{605} = 61\%$$

# Example 1: Area Requirements

2. If the shipments of each item are uncorrelated with each other, no safety stock is carried for each item, and retrievals to the factory floor will occur at a constant rate, what is an estimate the maximum number of units of all items that would ever occur?

$$M = \left\lceil \sum_{i=1}^N \frac{M_i}{2} + \frac{1}{2} \right\rceil = \left\lceil \frac{31+62+42}{2} + \frac{1}{2} \right\rceil = 68$$

# Example 1: Area Requirements

3. If a randomized policy is used to store the items, what is total 2-D area needed for the storage region?

$$D = 3$$

$$\begin{aligned} L(3) &= \left\lceil \frac{M + NH \left( \frac{D-1}{2} \right) + N \left( \frac{H-1}{2} \right)}{DH} \right\rceil \\ &= \left\lceil \frac{68 + 3(4) \left( \frac{3-1}{2} \right) + N \left( \frac{4-1}{2} \right)}{3(4)} \right\rceil = 8 \text{ lanes} \end{aligned}$$

$$TA(3) = xL(D) \cdot \left( yD + \frac{A}{2} \right) = 3(8) \cdot \left( 3.5(3) + \frac{10}{2} \right) = 372 \text{ ft}^2$$

# Example 1: Area Requirements

4. What is the optimal lane depth for randomized storage?

$$D^* = \left\lfloor \sqrt{\frac{A(2M-N)}{2NyH}} + \frac{1}{2} \right\rfloor = \left\lfloor \sqrt{\frac{10(2(68)-3)}{2(3)3.5(4)}} + \frac{1}{2} \right\rfloor = 4$$

5. What is the change in total area associated with using the optimal lane depth as opposed to storing the items three deep?

$$D = 4 \Rightarrow L(4) = \left\lceil \frac{68 + 3(4)\left(\frac{4-1}{2}\right) + N\left(\frac{4-1}{2}\right)}{3(4)} \right\rceil = 6 \text{ lanes}$$

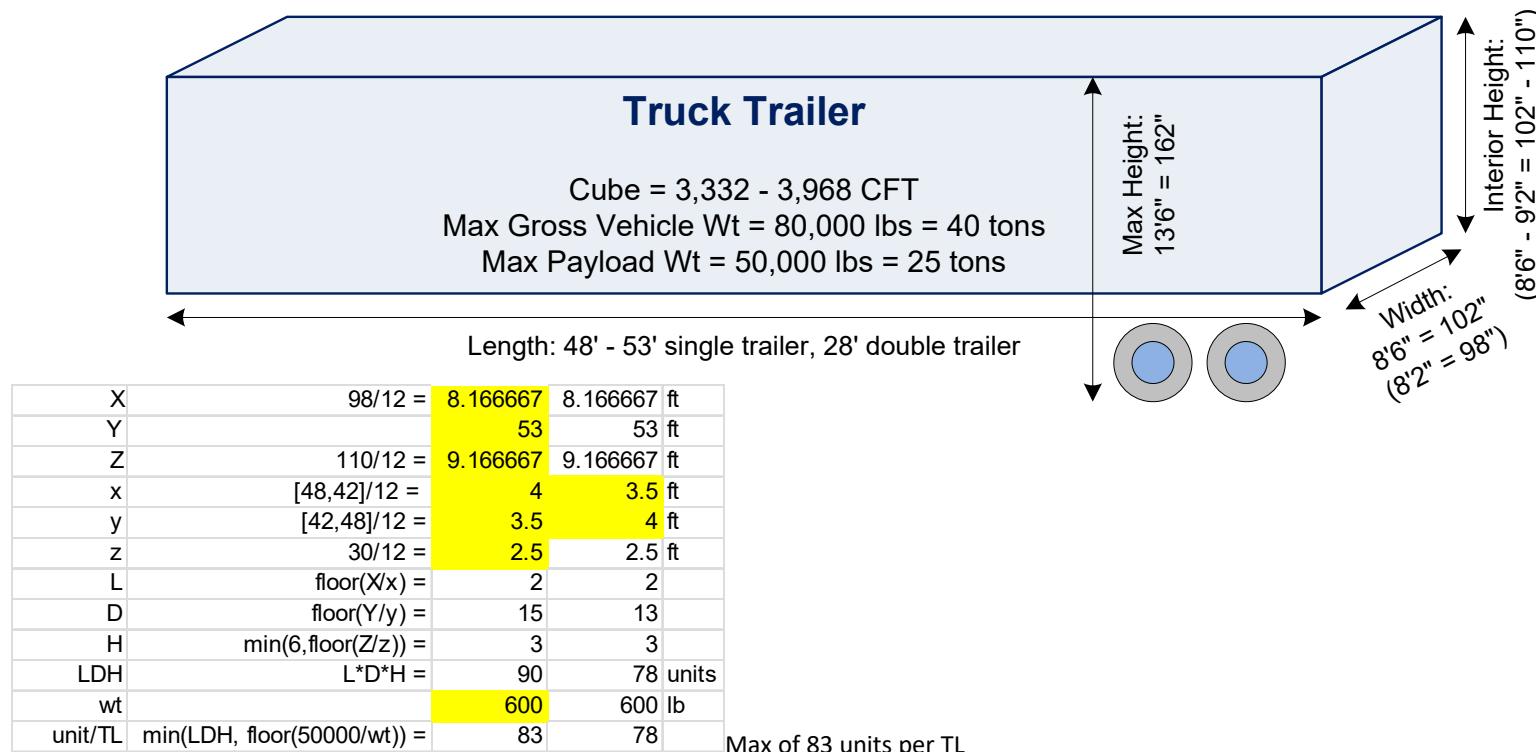
$$\Rightarrow TA(4) = 3(6) \cdot \left( 3.5(4) + \frac{10}{2} \right) = 342 \text{ ft}^2$$

$$D = 3 \Rightarrow TA(3) = 372 \text{ ft}^2$$

# Example 2: Trailer Loading

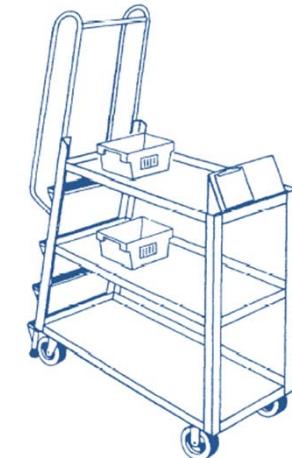
How many identical  $48 \times 42 \times 36$  in. four-way containers can be shipped in a full truckload? Each container load:

1. Weighs 600 lb
2. Can be stacked up to six high without causing damage from crushing
3. Can be rotated on the trucks with respect to their width and depth.

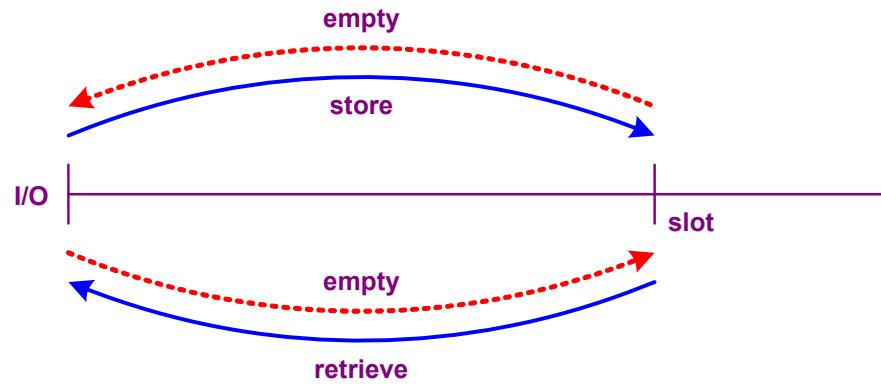


# Storage and Retrieval Cycle

- A storage and retrieval (S/R) cycle is one complete roundtrip from an I/O port to slot(s) and back to the I/O
- Type of cycle depends on load carrying ability:
  - Carrying one load at-a-time (load carried on a pallet):
    - Single command
    - Dual command
  - Carrying multiple loads (order picking of small items):
    - Multiple command



# Single-Command S/R Cycle



Expected time for each SC S/R cycle:

$$t_{SC} = \frac{d_{SC}}{v} + t_L + t_U = \frac{d_{SC}}{v} + 2t_{L/U}$$

where

$d_{SC}$  = expected distance per SC cycle

$v$  = average travel speed (e.g.: 2 mph = 176 fpm walking; 7 mph = 616 fpm riding)

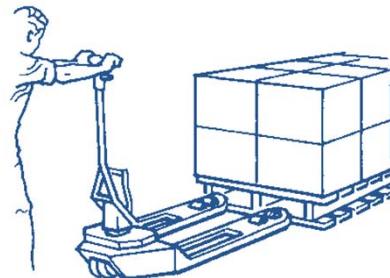
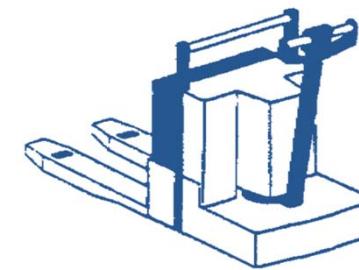
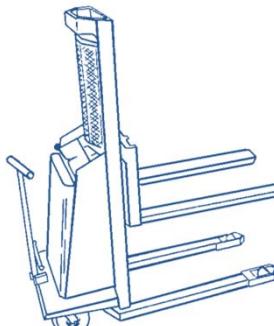
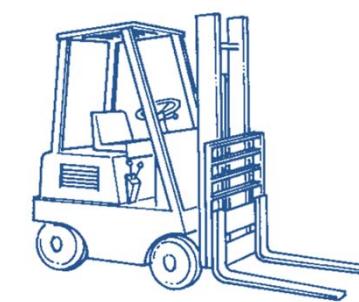
$t_L$  = loading time

$t_U$  = unloading time

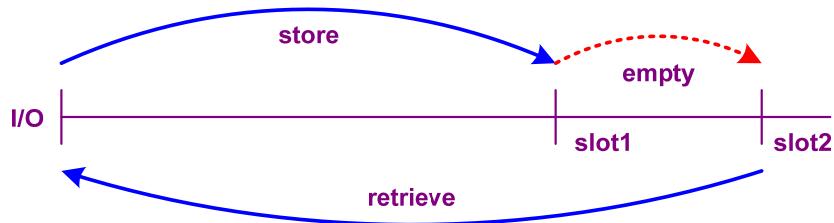
$t_{L/U}$  = loading/unloading time, if same value

- Single-command (SC) cycles:
  - Storage: carry one load to slot for storage and return empty back to I/O port, or
  - Retrieval: travel empty to slot to retrieve load and return with it back to I/O port

# Industrial Trucks: Walk vs. Ride

Walk (2 mph = 176 fpm)	Ride (7 mph = 616 fpm)
	
Pallet Jack	Pallet Truck
	
Walkie Stacker	Sit-down Counterbalanced Lift Truck

# Dual-Command S/R Cycle

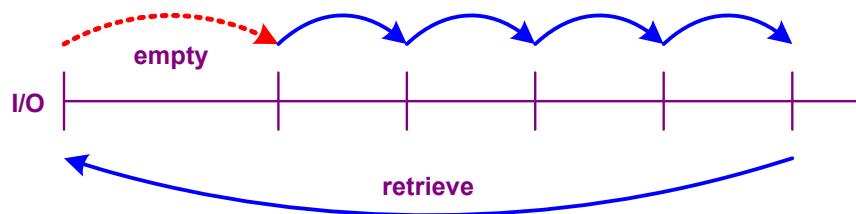


Expected time for each SC S/R cycle:

$$t_{DC} = \frac{d_{DC}}{v} + 2t_L + 2t_U = \frac{d_{DC}}{v} + 4t_{L/U}$$

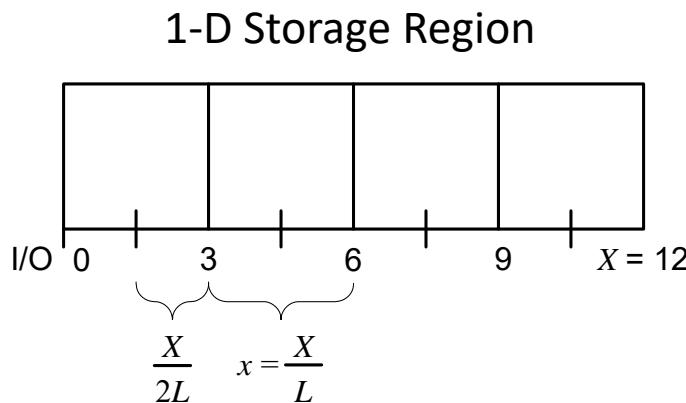
- Dual-command (DC):
- Combine storage with a retrieval:
  - store load in slot 1, travel empty to slot 2 to retrieve load
- Can reduce travel distance by a third, on average
- Also termed task “interleaving”

# Multi-Command S/R Cycle



- Multi-command: multiple loads can be carried at the same time
- Used in case and piece order picking
- Picker routed to slots
  - Simple VRP procedures can be used

# 1-D Expected Distance



$$TD_{1-way} = \sum_{i=1}^L \left( i \frac{X}{L} - \frac{X}{2L} \right) = \frac{X}{L} \sum_{i=1}^L i - \frac{X}{2L} (1)$$

$$= \frac{X}{L} \left( \frac{L(L+1)}{2} \right) - \frac{X}{2L} (L)$$

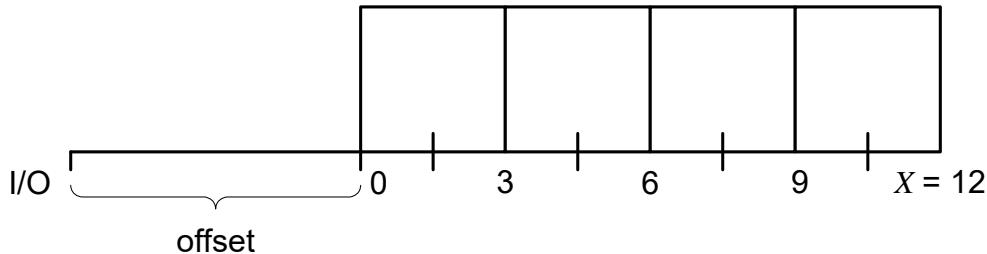
$$= \frac{XL + X - X}{2} = \frac{XL}{2}$$

$$ED_{1-way} = \frac{TD_{1-way}}{L} = \frac{X}{2}$$

$$d_{SC} = 2(ED_{1-way}) = X$$

- Assumptions:
  - All single-command cycles
  - Rectilinear distances
  - Each slot is region used with equal frequency (i.e., randomized storage)
- Expected distance is the average distance from I/O port to midpoint of each slot
  - e.g.,  $[2(1.5) + 2(4.5) + 2(6.5) + 2(10.5)]/4 = 12$

# Off-set I/O Port



- If the I/O port is off-set from the storage region, then 2 times the distance of the offset is added the expected distance within the slots

$$d_{SC} = 2(d_{\text{offset}}) + X$$

# 2-D Expected Distances

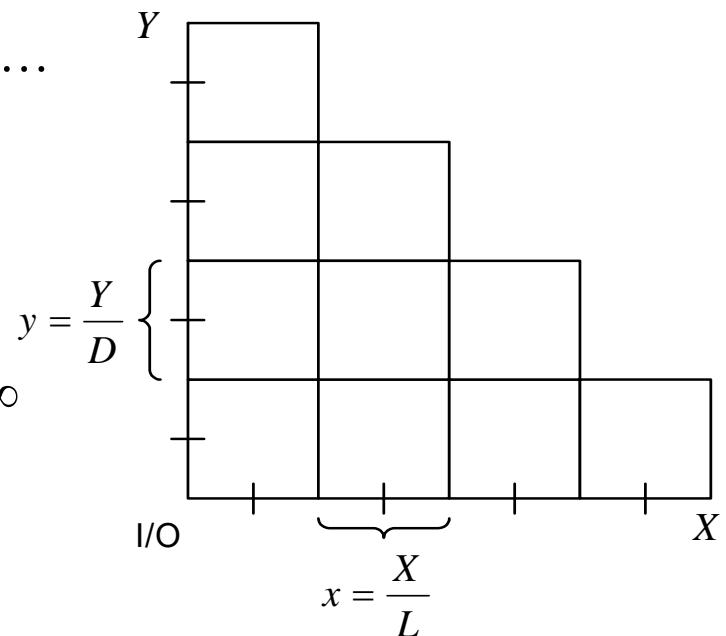
- Since dimensions  $X$  and  $Y$  are independent of each other for rectilinear distances, the expected distance for a 2-D rectangular region with the I/O port in a corner is just the sum of the distance in  $X$  and in  $Y$ :  $d_{SC}^{rect} = X + Y$
- For a triangular region with the I/O port in the corner:

$$TD_{1\text{-way}} = \sum_{i=1}^L \sum_{j=1}^{L-i+1} \left[ \left( i \frac{X}{L} - \frac{X}{2L} \right) + \left( j \frac{X}{L} - \frac{X}{2L} \right) \right] = \dots$$

$$= \frac{X}{6} (2L^2 + 3L + 1)$$

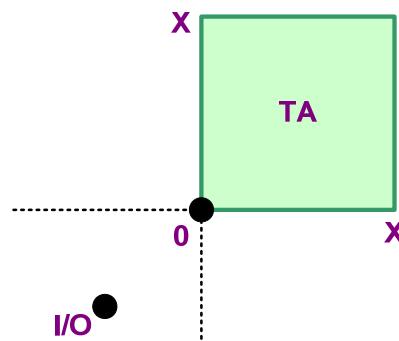
$$ED_{1\text{-way}} = \frac{TD_{1\text{-way}}}{L(L+1)} = \frac{2}{3} X + \frac{X}{3L} = \frac{2}{3} X, \quad \text{as } L \rightarrow \infty$$

$$d_{SC}^{tri} = 2 \left( \frac{2}{3} X \right) = 2 \left( \frac{1}{3} X + \frac{1}{3} Y \right) = \frac{2}{3} (X + Y)$$



# I/O-to-Side Configurations

Rectangular

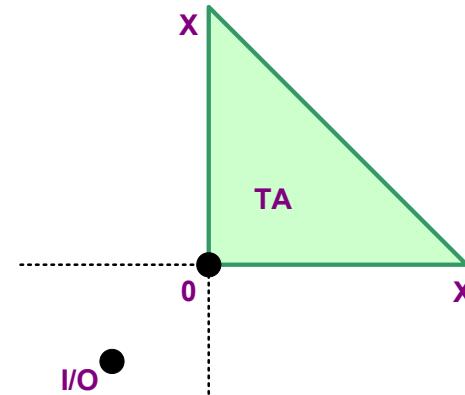


$$TA = X^2$$

$$\Rightarrow X = \sqrt{TA}$$

$$\Rightarrow d_{SC} = 2\sqrt{TA}$$

Triangular



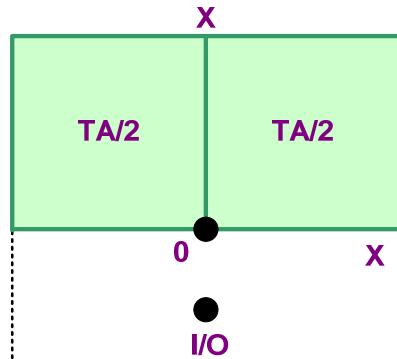
$$TA = \frac{1}{2} X^2$$

$$\Rightarrow X = \sqrt{2TA} = \sqrt{2}\sqrt{TA}$$

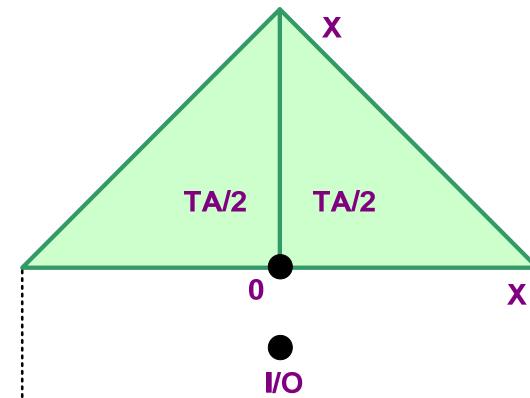
$$\Rightarrow d_{SC} = \frac{4}{3}\sqrt{2}\sqrt{TA} = 1.886\sqrt{TA}$$

# I/O-at-Middle Configurations

Rectangular



Triangular



$$\frac{TA}{2} = X^2$$

$$\Rightarrow X = \sqrt{\frac{TA}{2}} = \frac{\sqrt{TA}}{\sqrt{2}}$$

$$\Rightarrow d_{SC} = \sqrt{2} \sqrt{TA} = 1.414 \sqrt{TA}$$

$$\frac{TA}{2} = \frac{1}{2} X^2$$

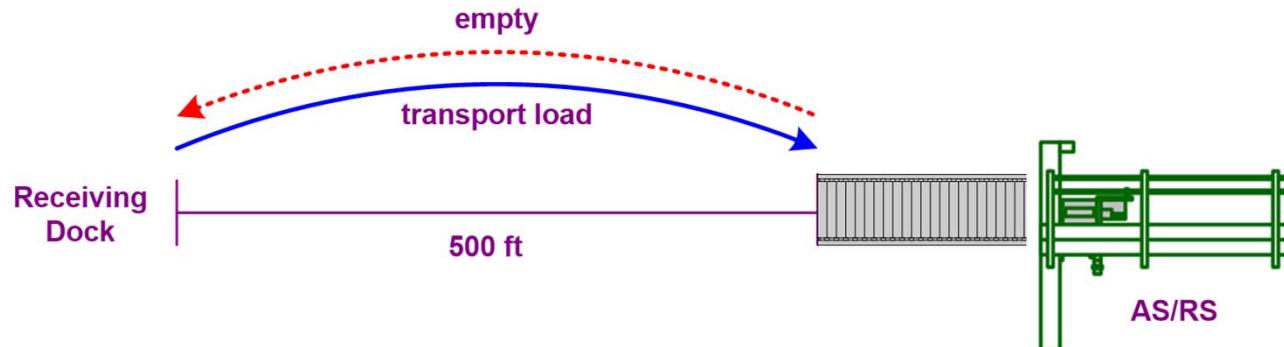
$$\Rightarrow X = \sqrt{TA}$$

$$\Rightarrow d_{SC} = \frac{4}{3} \sqrt{TA} = 1.333 \sqrt{TA}$$

# Example 3: Handling Requirements

Pallet loads will be unloaded at the receiving dock of a warehouse and placed on the floor. From there, they will be transported 500 feet using a dedicated pallet truck to the in-floor induction conveyor of an AS/RS. Given

- a. It takes 30 sec to load each pallet at the dock
- b. 30 sec to unload it at the induction conveyor
- c. There will be 80,000 loads per year on average
- d. Operator rides on the truck (because a pallet truck)
- e. Facility will operate 50 weeks per year, 40 hours per week



# Example 3: Handling Requirements

- Assuming that it will take 30 seconds to load each pallet at the dock and 30 seconds to unload it at the induction conveyor, what is the expected time required for each single-command S/R cycle?

$$d_{SC} = 2(500) = 1000 \text{ ft/mov}$$

$$\begin{aligned} t_{SC} &= \frac{d_{SC}}{v} + 2t_{L/U} = \frac{1000 \text{ ft/mov}}{616 \text{ ft/min}} + 2\left(\frac{30}{60}\right) \text{ min/mov} \\ &= 2.62 \text{ min/mov} = \frac{2.62}{60} \text{ hr/mov} \end{aligned}$$

(616 fpm because operator rides on a pallet truck)

## Example 3: Handling Requirements

2. Assuming that there will be 80,000 loads per year on average and that the facility will operate for 50 weeks per year, 40 hours per week, what is the minimum number of trucks needed?

$$r_{avg} = \frac{80,000 \text{ mov/yr}}{50(40) \text{ hr/yr}} = 40 \text{ mov/hr}$$

$$\begin{aligned}m &= \left\lfloor r_{avg} t_{SC} + 1 \right\rfloor \\&= \left\lfloor 40 \left( \frac{2.62}{60} \right) + 1 \right\rfloor = \left\lfloor 1.75 + 1 \right\rfloor \\&= 2 \text{ trucks}\end{aligned}$$

## Example 3: Handling Requirements

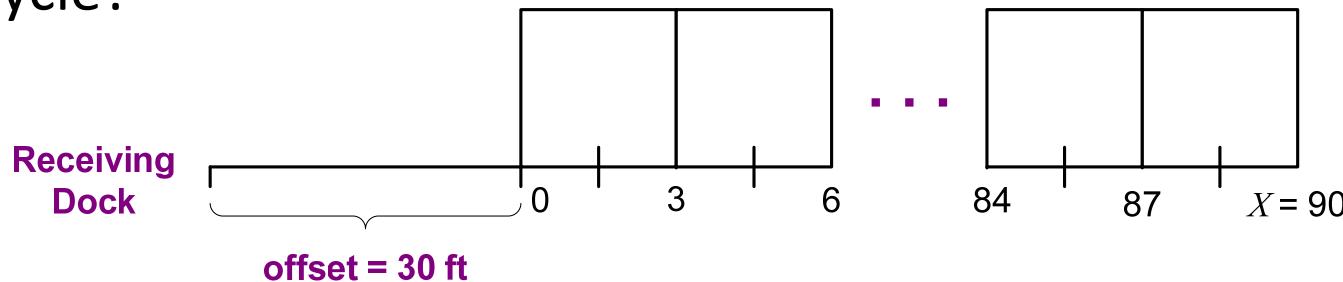
3. How many trucks are needed to handle a peak expected demand of 80 moves per hour?

$$r_{peak} = 80 \text{ mov/hr}$$

$$\begin{aligned}m &= \left\lfloor r_{peak} t_{SC} + 1 \right\rfloor \\&= \left\lfloor 80 \left( \frac{2.62}{60} \right) + 1 \right\rfloor = \left\lfloor 3.50 + 1 \right\rfloor \\&= 4 \text{ trucks}\end{aligned}$$

# Example 3: Handling Requirements

4. If, instead of unloading at the conveyor, the 3-foot-wide loads are placed side-by-side in a staging area along one side of 90-foot aisle that begins 30 feet from the dock, what is the expected time required for each single-command S/R cycle?



$$d_{SC} = 2(d_{\text{offset}}) + X = 2(30) + 90 = 150 \text{ ft}$$

$$t_{SC} = \frac{d_{SC}}{v} + 2t_{L/U} = \frac{150 \text{ ft/mov}}{616 \text{ ft/min}} + 2 \left( \frac{30}{60} \right) \text{ min/mov}$$

$$= 1.24 \text{ min/mov} = \frac{1.24}{60} \text{ hr/mov}$$

# Estimating Handling Costs

- Warehouse design involves the trade-off between building and handling cost.
- Maximizing the cube utilization of a storage region will help minimize building costs.
- Handling costs can be estimated by determining:
  1. Expected time required for each move based on an average of the time required to reach each slot in the region.
  2. Number of vehicles needed to handle a target *peak demand* for moves, e.g., moves per hour.
  3. *Operating costs per hour of vehicle operation*, e.g., labor, fuel (assuming the operators can perform other productive tasks when not operating a truck)
  4. Annual operating costs based on *annual demand* for moves.
  5. Total handling costs as the sum of the annual capital recovery costs for the vehicles and the annual operating costs.

# Example 4: Estimating Handing Cost

Expected Distance:  $d_{SC} = \sqrt{2} \sqrt{TA} = \sqrt{2} \sqrt{20,000} = 200$  ft

Expected Time:

$$\begin{aligned} t_{SC} &= \frac{d_{SC}}{v} + 2t_{L/U} \\ &= \frac{200 \text{ ft}}{200 \text{ fpm}} + 2(0.5 \text{ min}) = 2 \text{ min per move} \end{aligned}$$

TA = 20,000

I/O

Peak Demand:  $r_{\text{peak}} = 75$  moves per hour

Annual Demand:  $r_{\text{year}} = 100,000$  moves per year

Number of Trucks:  $m = \left\lfloor r_{\text{peak}} \frac{t_{SC}}{60} + 1 \right\rfloor = \left\lfloor 3.5 \right\rfloor = 3$  trucks

Handling Cost: 
$$\begin{aligned} TC_{\text{hand}} &= mK_{\text{truck}} + r_{\text{year}} \frac{t_{SC}}{60} C_{\text{labor}} \\ &= 3(\$2,500 / \text{tr-yr}) + 100,000 \frac{2}{60} (\$10 / \text{hr}) \\ &= \$7,500 + \$33,333 = \$40,833 \text{ per year} \end{aligned}$$

# Dedicated Storage Assignment (DSAP)

- The assignment of items to slots is termed *slotting*
  - With randomized storage, all items are assigned to all slots
- DSAP (dedicated storage assignment problem):
  - Assign  $N$  items to slots to minimize total cost of material flow
- DSAP solution procedure:
  1. *Order Slots*: Compute the expected cost for each slot and then put into nondecreasing order
  2. *Order Items*: Put the flow density (flow per unit of volume) for each item  $i$  into nonincreasing order

$$\frac{f_{[1]}}{M_{[1]}s_{[1]}} \geq \frac{f_{[2]}}{M_{[2]}s_{[2]}} \geq \dots \geq \frac{f_{[N]}}{M_{[N]}s_{[N]}}$$

3. *Assign Items to Slots*: For  $i = 1, \dots, N$ , assign item  $[i]$  to the first slots with a total volume of at least  $M_{[i]}s_{[i]}$

# 1-D Slotting Example

		A	B	C
Max units	M	4	5	3
Space/unit	s	1	1	1
Flow	f	24	7	21
Flow Density	f/(M x s)	6.00	1.40	7.00

Flow Density	1-D Slot Assignments	Expected Distance	Flow	Total Distance
$\frac{21}{3} = 7.00$	I/O [c, c, c] 0 3	$2(0) + 3 = 3 \times$	21	= 63
$\frac{24}{4} = 6.00$	I/O [-3, 0, 4] A A A A	$2(3) + 4 = 10 \times$	24	= 240
$\frac{7}{5} = 1.40$	I/O [-7, 0, 5] B B B B	$2(7) + 5 = 19 \times$	7	= 133
	I/O [c, c, c, A, A, A, B, B, B, B] 0 3 7 12			436

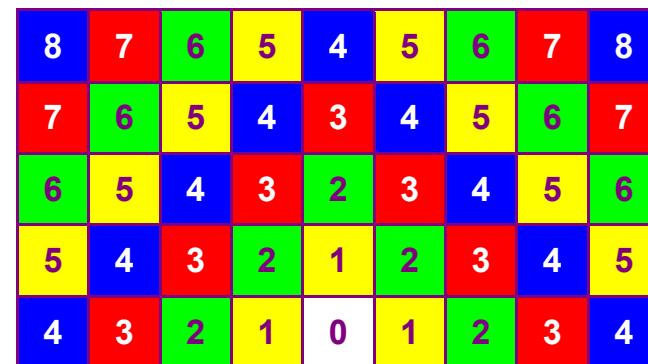
# 1-D Slotting Example (cont)

		Dedicated			Random		Class-Based		
		A	B	C	ABC	AB	AC	BC	
Max units	M	4	5	3	9	7	7	8	
Space/unit	s	1	1	1	1	1	1	1	
Flow	f	24	7	21	52	31	45	28	
Flow Density	f/(M x s)	6.00	1.40	7.00	5.78	4.43	6.43	3.50	

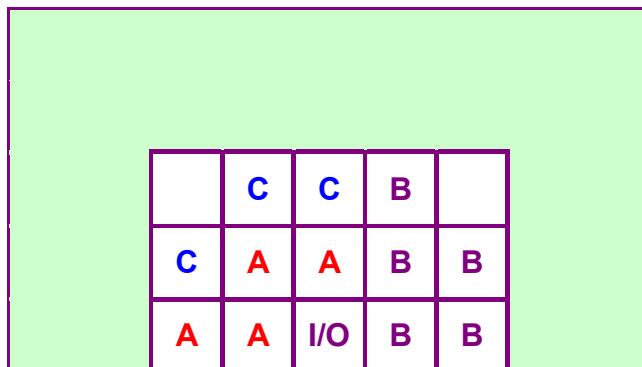
1-D Slot Assignments										Total Distance	Total Space	
Dedicated (flow density)	I/O	c	c	c	a	a	a	a	b	b	b	b
Dedicated (flow only)	I/O	a	a	a	a	c	c	c	b	b	b	b
Class-based	I/O	c	c	c	ab	ab	ab	ab	ab	ab		
Randomized	I/O	abc										

# 2-D Slotting Example

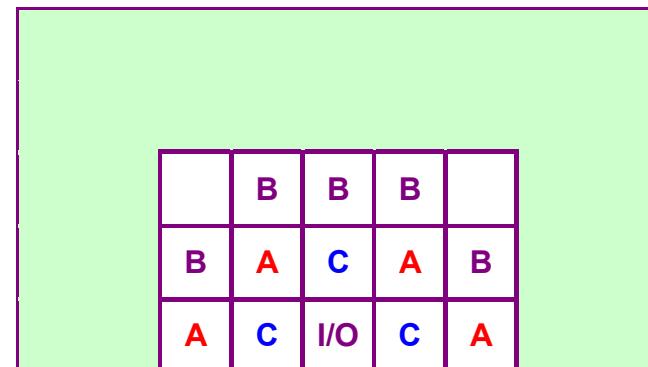
		A	B	C
Max units	M	4	5	3
Space/unit	s	1	1	1
Flow	f	24	7	21
Flow Density	f/(M x s)	6.00	1.40	7.00



Distance from I/O to Slot



Original Assignment (TD = 215)



Optimal Assignment (TD = 177)

# DSAP Assumptions

1. All SC S/R moves
2. For item  $i$ , probability of move to/from each slot assigned to item is the same
3. The *factoring assumption*:
  - a. Handling cost and distances (or times) for each slot are identical for all items
  - b. Percent of S/R moves of item stored at slot  $j$  to/from I/O port  $k$  is identical for all items
- Depending of which assumptions not valid, can determine assignment using other procedures

$$\begin{array}{c} \left( c_i x_{ij} \right) DSAP \subset LAP \subset LP \subset QAP \left( c_{ijkl} x_{ij} x_{kl} \right) \\ \left( c_{ij} x_{ij} \right) \qquad \qquad \qquad \stackrel{\cup}{TSP} \end{array}$$

## Example 5: 1-D DSAP

- What is the change in the minimum expected total distance traveled if dedicated, as compared to randomized, block stacking is used, where
  - a. Slots located on one side of 10-foot-wide down aisle
  - b. All single-command S/R operations
  - c. Each lane is three-deep, four-high
  - d.  $40 \times 36$  in. two-way pallet used for all loads
  - e. Max inventory levels of SKUs A, B, C are 94, 64, and 50
  - f. Inventory levels are uncorrelated and retrievals occur at a constant rate
  - g. Throughput requirements of A, B, C are 160, 140, 130
  - h. Single I/O port is located at the end of the aisle

# Example 5: 1-D DSAP

- Randomized:



$$M = \left\lfloor \frac{M_A + M_B + M_C}{2} + \frac{1}{2} \right\rfloor = \left\lfloor \frac{94 + 64 + 50}{2} + \frac{1}{2} \right\rfloor = 104$$

$$\begin{aligned} L_{rand} &= \left\lceil \frac{M + NH \left( \frac{D-1}{2} \right) + N \left( \frac{H-1}{2} \right)}{DH} \right\rceil \\ &= \left\lceil \frac{104 + 3(4) \left( \frac{3-1}{2} \right) + N \left( \frac{4-1}{2} \right)}{3(4)} \right\rceil = 11 \text{ lanes} \end{aligned}$$

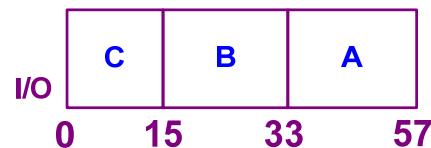
$$X = xL_{rand} = 3(11) = 33 \text{ ft}$$

$$d_{SC} = X = 33 \text{ ft}$$

$$TD_{rand} = (f_A + f_B + f_C)X = (160 + 140 + 130)33 = 14,190 \text{ ft}$$

# Example 5: 1-D DSAP

- Dedicated:



$$\frac{f_A}{M_A} = \frac{160}{94} = 1.7, \frac{f_B}{M_B} = \frac{140}{64} = 2.19, \frac{f_C}{M_C} = \frac{130}{50} = 2.6 \Rightarrow C < B < A$$

$$L_A = \left\lceil \frac{M_A}{DH} \right\rceil = \left\lceil \frac{160}{3(4)} \right\rceil = 8, L_B = \left\lceil \frac{M_B}{DH} \right\rceil = \left\lceil \frac{140}{3(4)} \right\rceil = 6, L_C = \left\lceil \frac{M_C}{DH} \right\rceil = \left\lceil \frac{130}{3(4)} \right\rceil = 5$$

$$X_C = xL_C = 3(5) = 15, X_B = xL_B = 3(6) = 18, X_A = xL_A = 3(8) = 24$$

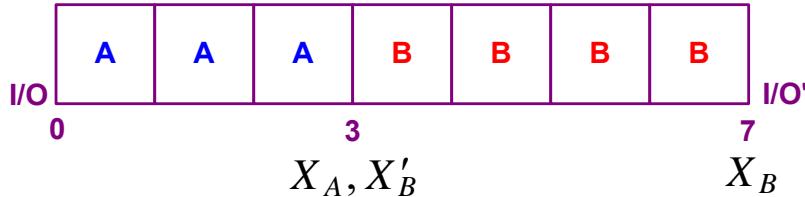
$$d_{SC}^C = X_C = 3(5) = 15 \text{ ft}$$

$$d_{SC}^B = 2(X_C) + X_B = 2(15) + 18 = 48 \text{ ft}$$

$$d_{SC}^A = 2(X_C + X_B) + X_A = 2(15 + 18) + 24 = 90 \text{ ft}$$

$$TD_{ded} = f_A d_{SC}^A + f_B d_{SC}^B + f_C d_{SC}^C = 160(90) + 140(48) + 130(15) = 23,070 \text{ ft}$$

# 1-D Multiple Region Expected Distance



$$d_{SC}^A = d_A = X_A = 3$$

$$d_B = 2d_{offset} + (X_B - X_A) = 2X_A + (X_B - X_A) = X_A + X_B = 10$$

$$= 2(d_{I/O \text{ to } I/O'}) - X'_B = 2(7) - 4 = 10$$

$$d_{AB} = 7$$

$$TA_A = X_A = 3, \quad TA_B = X_B - X_A = 4, \quad TA_{AB} = TA_A + TA_B = 7$$

$$TM_A = TA_A d_A, \quad TM_B = TA_B d_B$$

$$TM_{AB} = TA_{AB} d_{AB} = X_B^2 = (X_A + X_B - X_A)^2 = X_A^2 + (X_B - X_A)(2X_A + X_B - X_A)$$

$$= TA_A d_A + TA_B d_B = TM_A + TM_B$$

$$d_B = \frac{TM_B}{TA_B} = \frac{TM_{AB} - TM_A}{TA_B} = \frac{TA_{AB} d_{AB} - TA_A d_A}{TA_B} = \frac{7(7) - 3(3)}{4} = 10$$

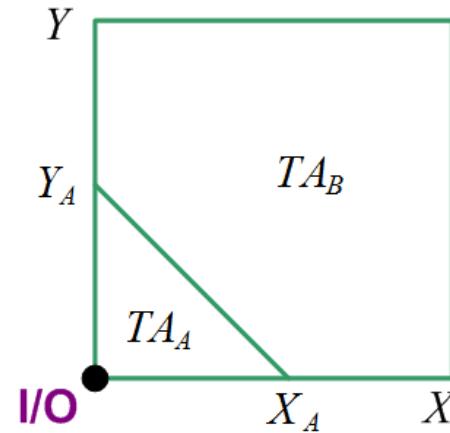
- In 1-D, easy to determine the offset
- In 2-D, no single offset value for each region

# 2-D Multiple Region Expected Distance

**Case:**  $TA_A \leq \frac{1}{2} TA_{AB}$

Let  $X = Y, X_A = Y_A, d = d_{sc}$

$$d_A = \frac{2}{3}(X_A + Y_A) = \frac{4}{3}X_A$$



$$d_B = \frac{TM_B}{TA_B} = \frac{TM_{AB} - TM_A}{TA_B} = \frac{TA_{AB}d_{AB} - TA_A d_A}{TA_B} = \frac{XY(X+Y) - \frac{1}{2}X_A Y_A \frac{2}{3}(X_A + Y_A)}{XY - \frac{1}{2}X_A Y_A}$$

$$= \frac{2}{3} \frac{3X^2Y + 3XY^2 - X_A^2Y_A - X_A Y_A^2}{2XY - X_A Y_A} = \frac{4}{3} \frac{3X^3 - X_A^3}{2X^2 - X_A^2}$$

If  $X = X_A \Rightarrow d_B = \frac{4}{3} \frac{X^2(2X)}{X^2} = \frac{8}{3}X$

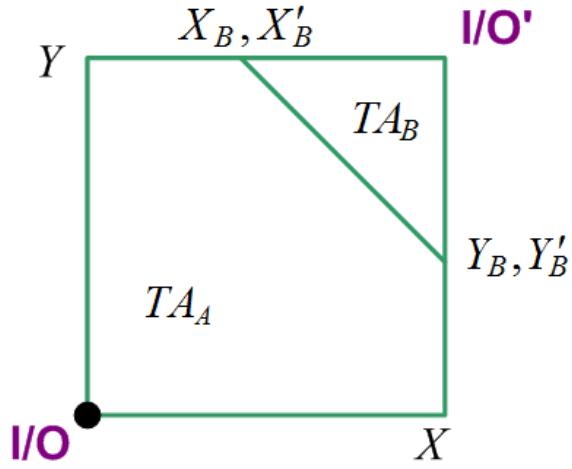
# 2-D Multiple Region Expected Distance

**Case:**  $TA_A \geq \frac{1}{2} TA_{AB}$

Let  $X'_B = X - X_B, Y'_B = Y - Y_B$

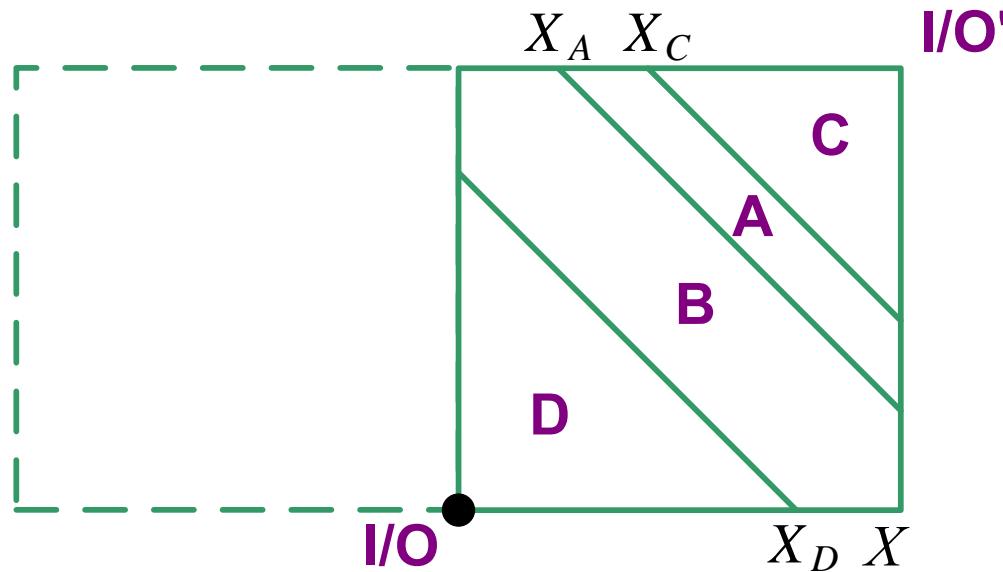
$$\begin{aligned}
 d_B &= 2(d_{I/O \text{ to } I/O'}) - \frac{2}{3}(X'_B + Y'_B) \\
 &= 2(X + Y) - \frac{2}{3}[(X - X_B) + (Y - Y_B)] \\
 &= \frac{4}{3}(X + Y) + \frac{2}{3}(X_B + Y_B) \\
 &= \frac{8}{3}X + \frac{4}{3}X_B, \text{ where } X = Y
 \end{aligned}$$

If  $X_B = 0 \Rightarrow d_B = \frac{8}{3}X$

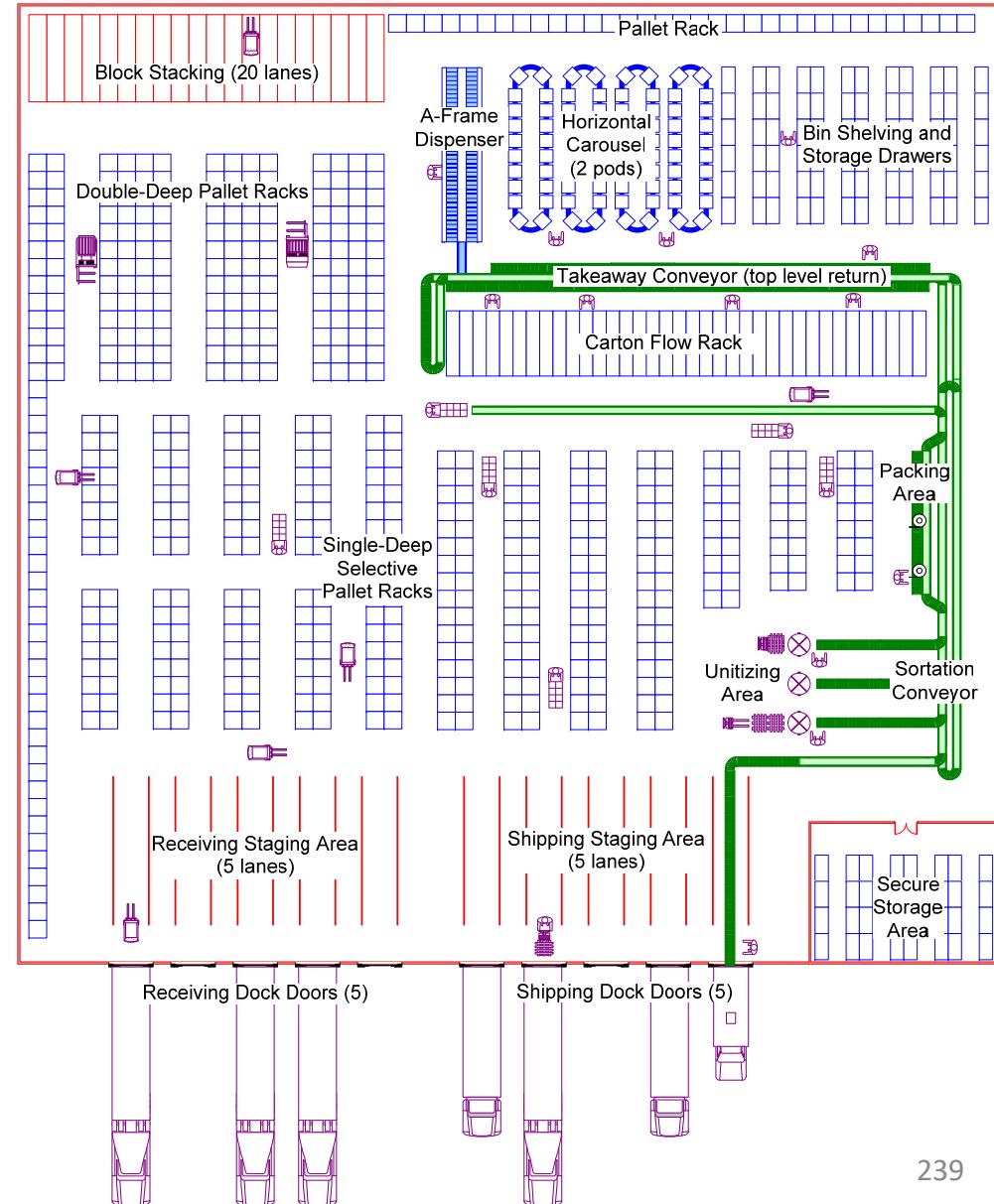
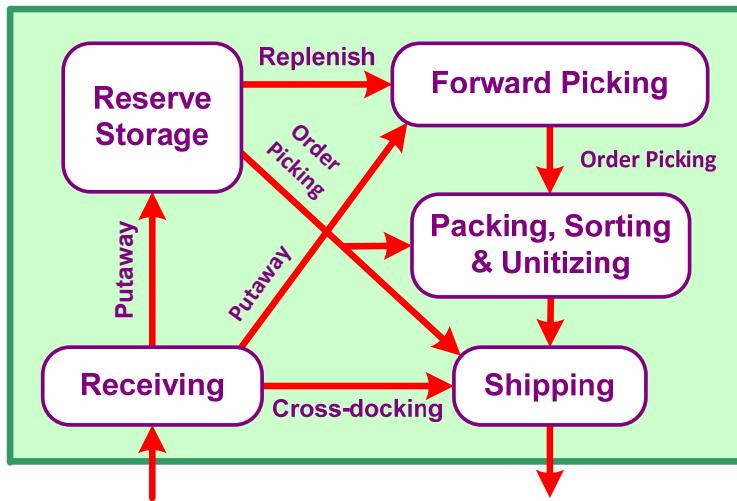


# 2-D Multiple Region Expected Distance

- If more than two regions
  - For regions below diagonal (D), start with region closest to I/O
  - For regions above diagonal (A+C), start with regions closest to I/O' (C)
  - For region in the middle (B), solve using whole area less other regions

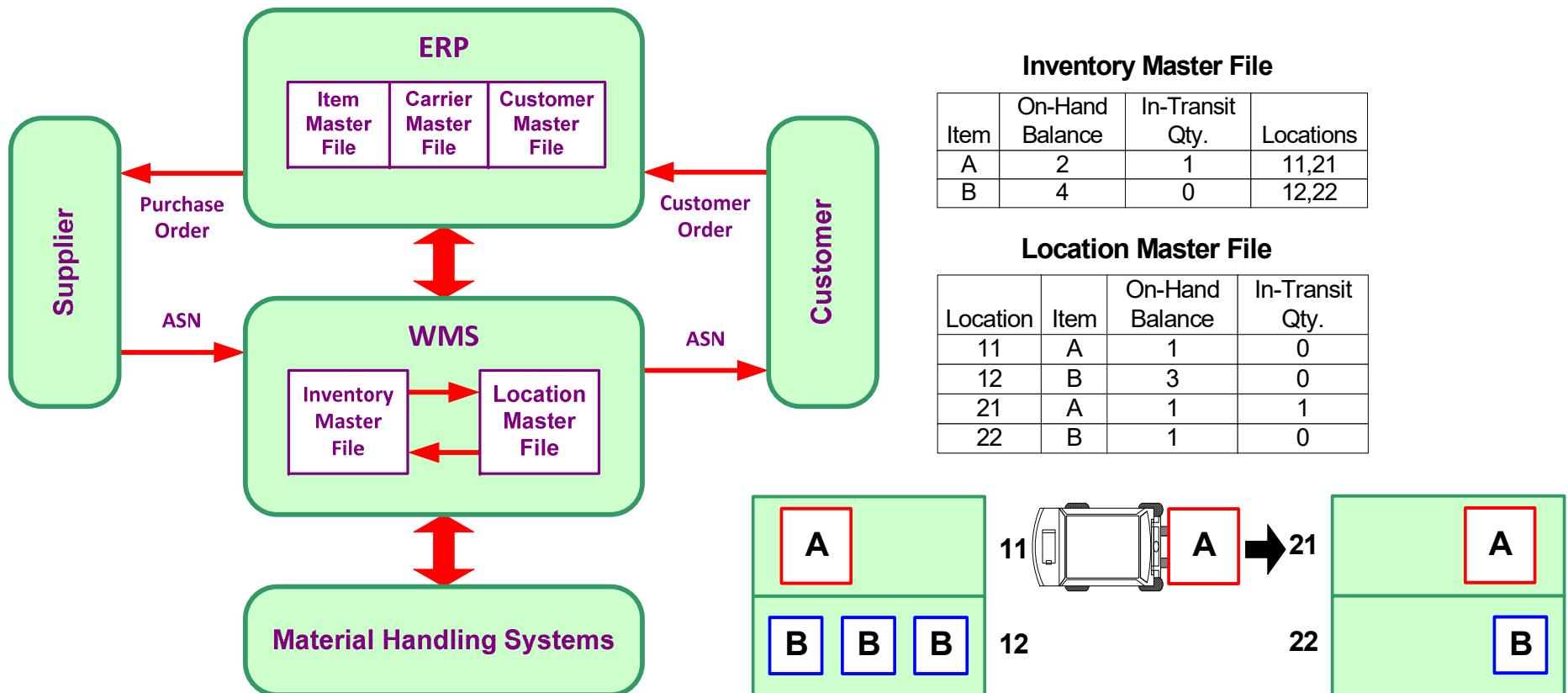


# Warehouse Operations



# Warehouse Management System

- WMS interfaces with a corporation's enterprise resource planning (ERP) and the control software of each MHS



- Advance shipping notice (ASN) is a standard format used for communications

# Logistics-related Codes

Commodity Code		Item Code	Unit Code
Level	Category	Class	Instance
Description	Grouping of similar objects	Grouping of identical objects	Unique physical object
Function	Product classification	Inventory control	Object tracking
Names	—	Item number, Part number, SKU, SKU + Lot number	Serial number, License plate
Codes	UNSPSC, GPC	GTIN, UPC, ISBN, NDC	EPC, SSCC

UNSPSC: United Nations Standard Products and Services Code

GPC: Global Product Catalogue

GTIN: Global Trade Item Number (includes UPC, ISBN, and NDC)

UPC: Universal Product Code

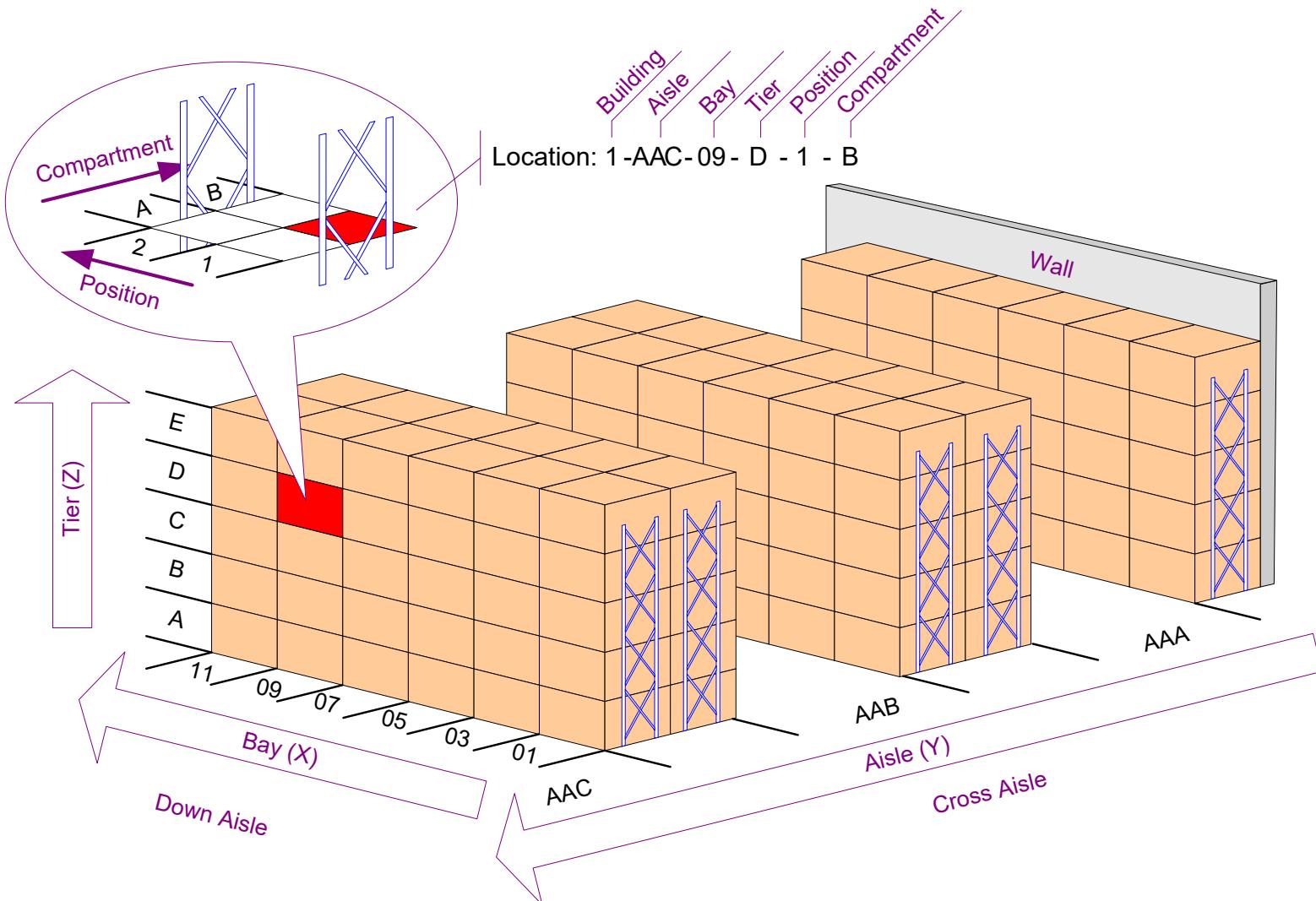
ISBN: International Standard Book Numbering

NDC: National Drug Code

EPC: Electronic Product Code (globally unique serial number for physical objects identified using RFID tags)

SSCC: Serial Shipping Container Code (globally unique serial number for identifying *movable units* (carton, pallet, trailer, etc.))

# Identifying Storage Locations



# Receiving



- Basic steps:
  1. Unload material from trailer.
  2. Identify supplier with ASN, and associate material with each moveable unit listed in ASN.
  3. Assign inventory attributes to movable unit from item master file, possibly including repackaging and assigning new serial number.
  4. Inspect material, possibly including holding some or all of the material for testing, and report any variances.
  5. Stage units in preparation for putaway.
  6. Update item balance in inventory master and assign units to a receiving area in location master.
  7. Create receipt confirmation record.
  8. Add units to putaway queue

# Putaway

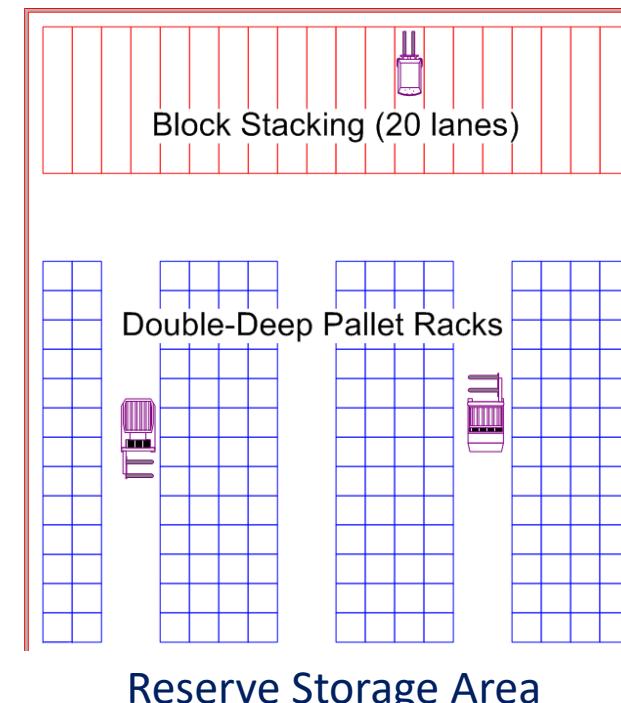


- A putaway algorithm is used in WMS to search for and validate locations where each movable unit in the putaway queue can be stored
- Inventory and location attributes used in the algorithm:
  - *Environment* (refrigerated, caged area, etc.)
  - *Container type* (pallet, case, or piece)
  - *Product processing type* (e.g., floor, conveyable, nonconveyable)
  - *Velocity* (assign to A, B, C based on throughput of item)
  - Preferred putaway zone (item should be stored in same zone as related items in order to improve picking efficiency)

# Replenishment



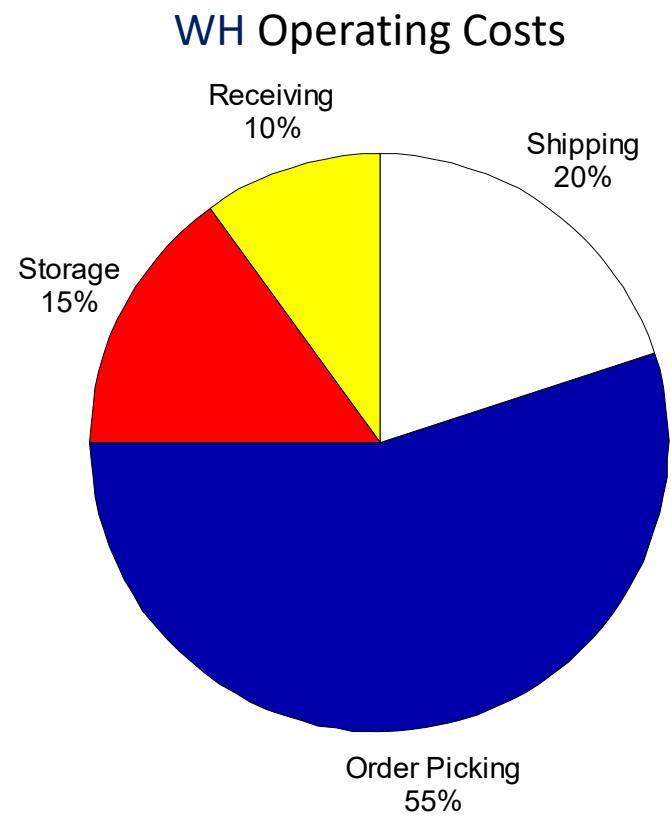
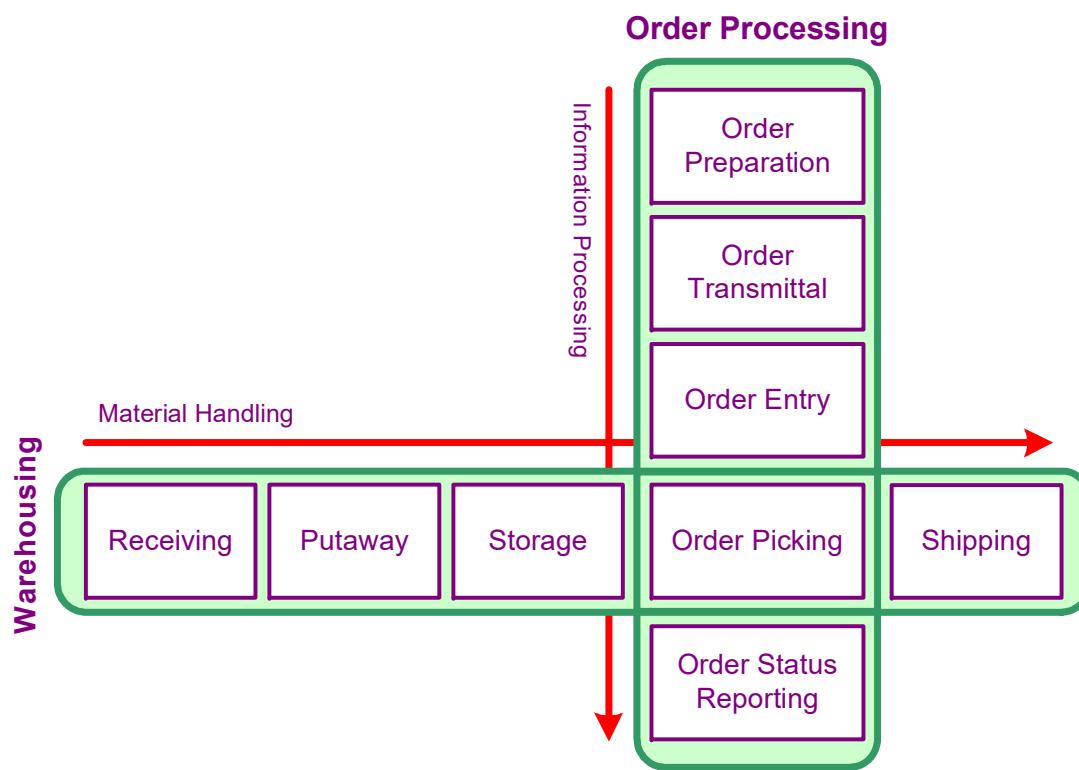
- Replenishment is the process of moving material from reserve storage to a forward picking area so that it is available to fill customer orders efficiently
- Other types of in-plant moves include:
  - Consolidation: combining several partially filled storage locations of an item into a single location
  - Rewarehousing: moving items to different storage locations to improve handling efficiency



# Order Picking



- Order picking is at the intersection of warehousing and order processing



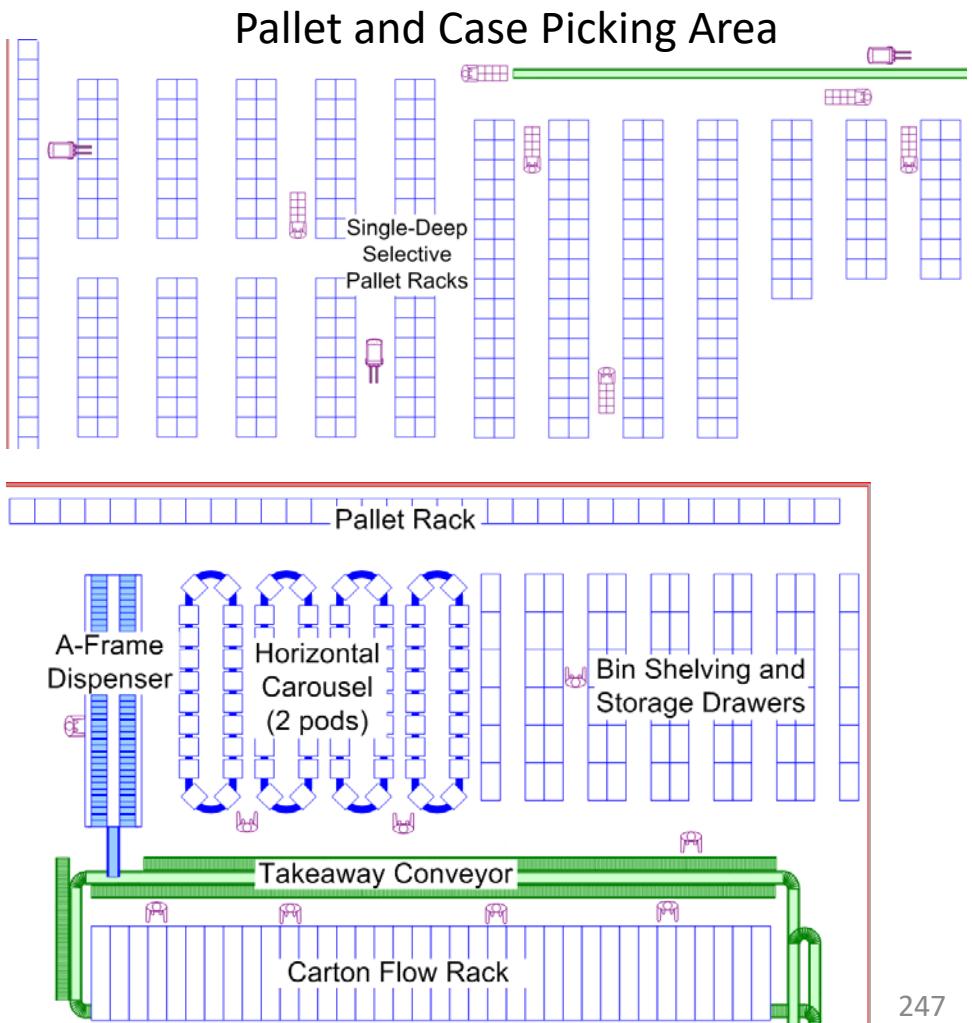
# Order Picking



## Levels of Order Picking



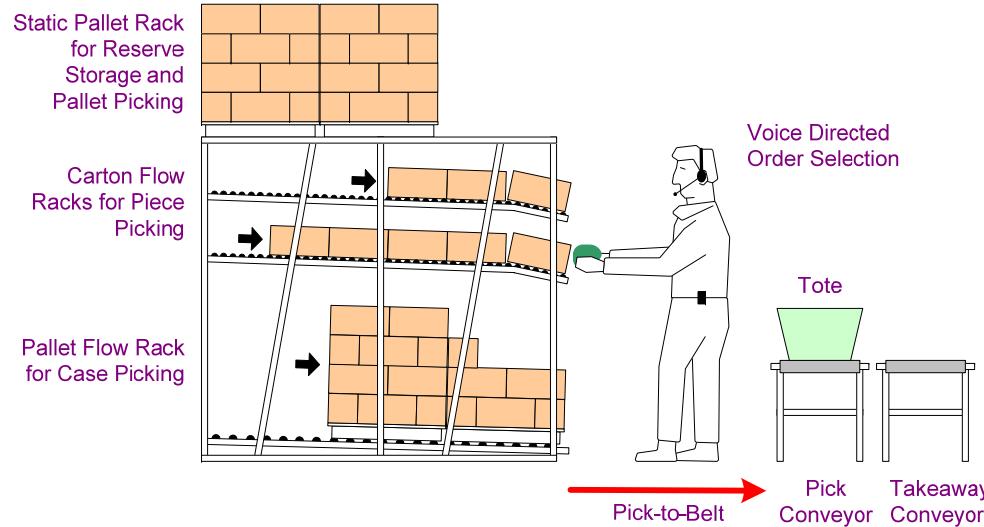
Forward Piece  
Picking Area



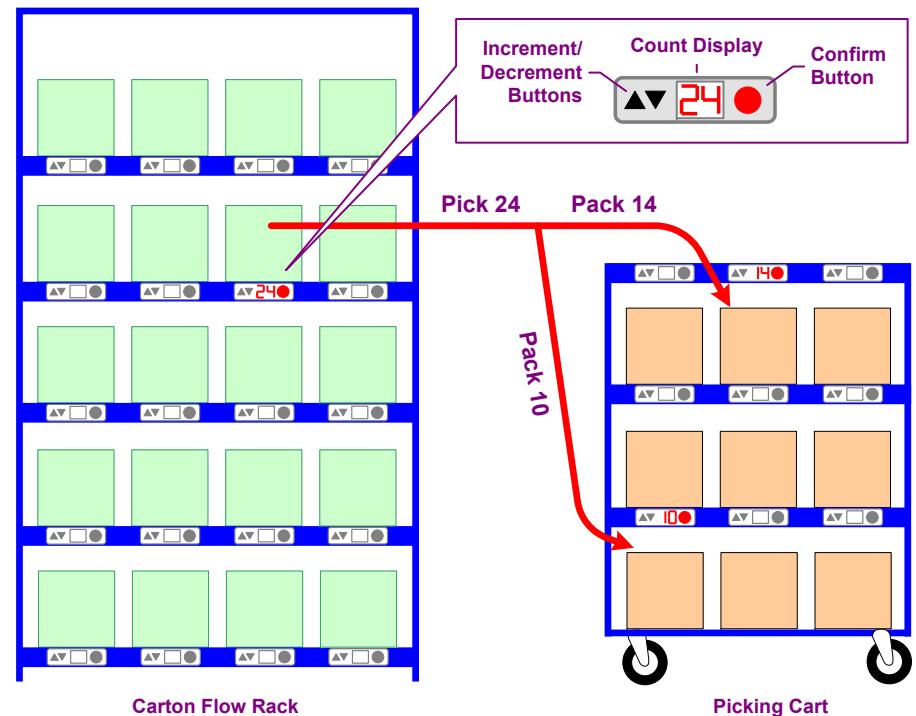
# Order Picking



Voice-Directed Piece and Case Picking



Pick-to-Light Piece Picking

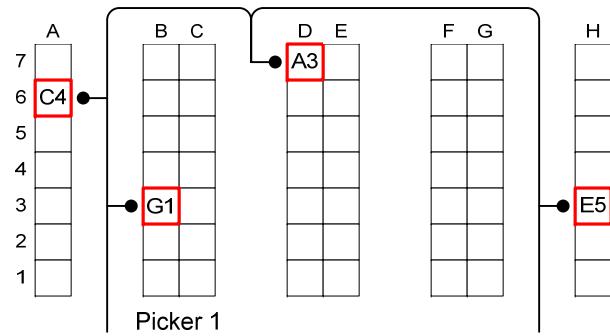


# Order Picking



## Methods of Order Picking

Discrete



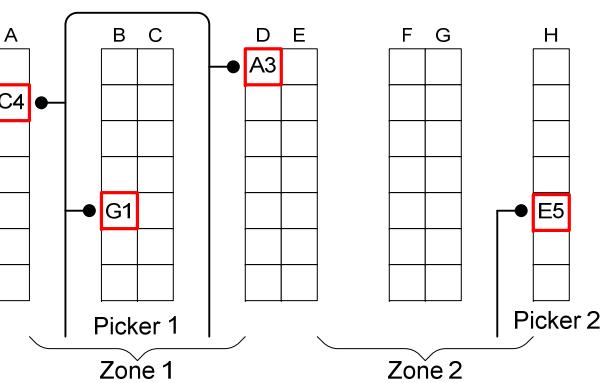
Method	Pickers per Order	Orders per Picker
Discrete	Single	Single

Zone	Multiple	Single
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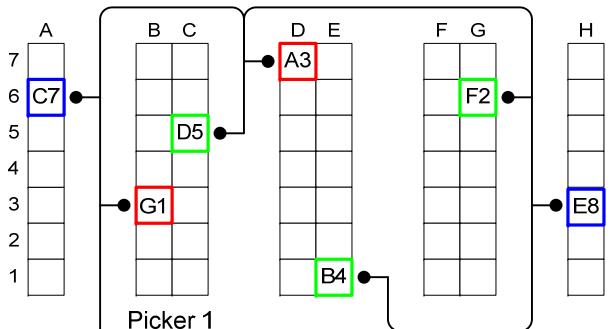
Batch	Single	Multiple
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Zone-Batch	Multiple	Multiple
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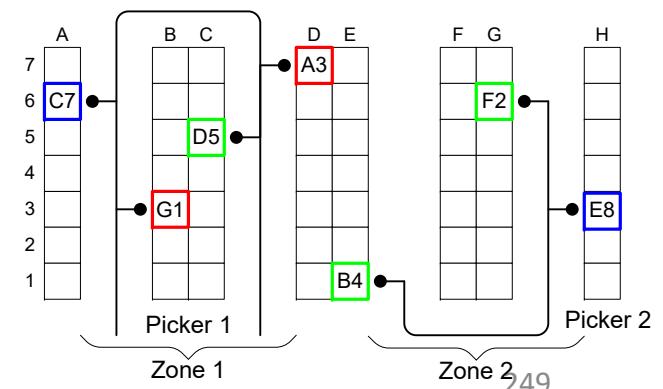
Zone



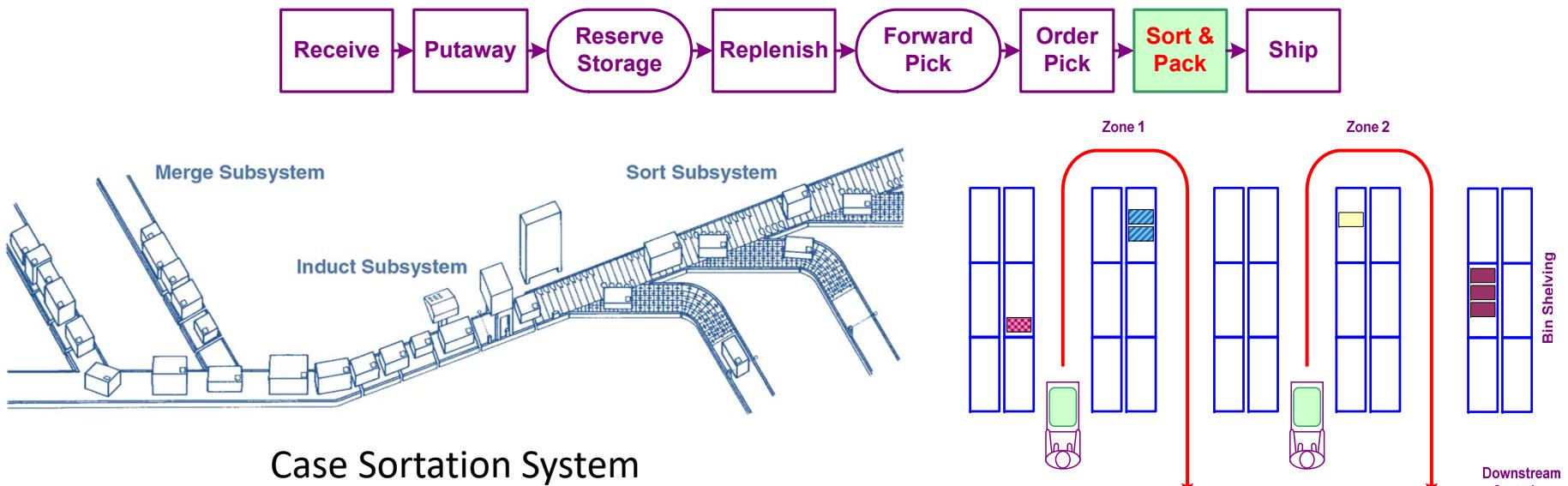
Batch



Zone-Batch



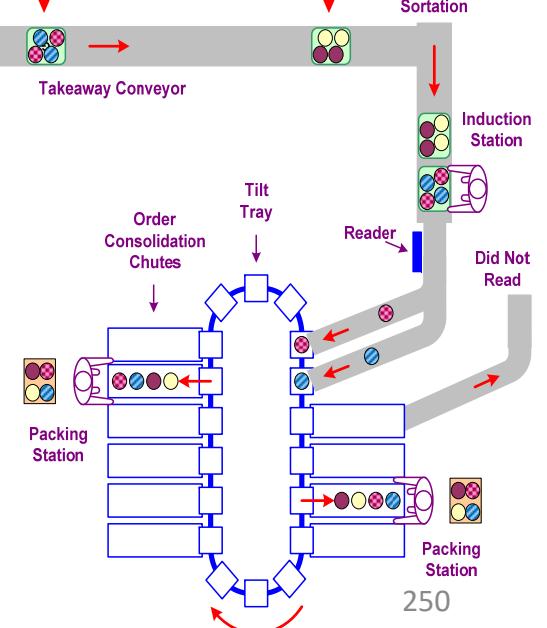
# Sortation and Packing



Case Sortation System



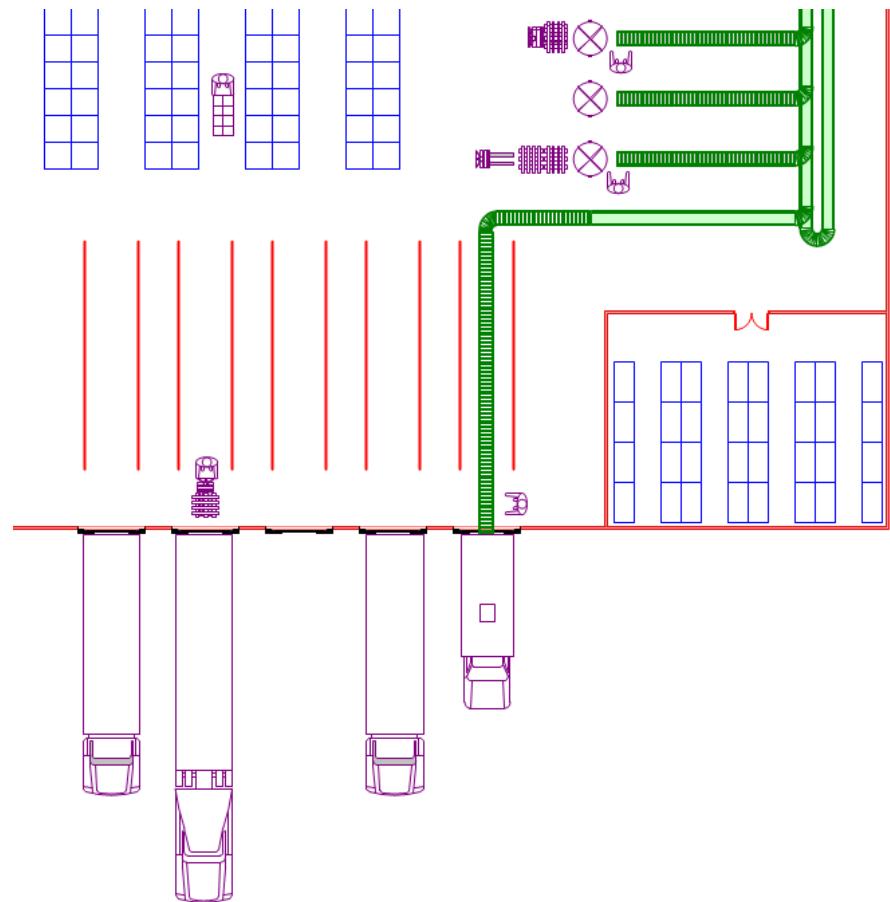
Wave zone-batch piece picking, including downstream tilt-tray-based sortation



# Shipping



- Staging, verifying, and loading orders to be transported
  - ASN for each order sent to the customer
  - Customer-specific shipping instructions retrieved from customer master file
  - Carrier selection is made using the rate schedules contained in the carrier master file



Shipping Area

# Activity Profiling

- *Total Lines*: total number of lines for all items in all orders
- *Lines per Order*: average number of different items (lines/SKUs) in order
- *Cube per Order*: average total cubic volume of all units (pieces) in order
- *Flow per Item*: total number of S/R operations performed for item
- *Lines per Item (popularity)*: total number of lines for item in all orders
- *Cube Movement*: total unit demand of item time x cubic volume
- *Demand Correlation*: percent of orders in which both items appear

**Customer Orders**

Order: 1	
SKU	Qty
A	5
B	3
C	2
D	6

Order: 2	
SKU	Qty
A	4
C	1

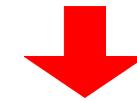
Order: 3	
SKU	Qty
A	2

Order: 4	
SKU	Qty
B	2

Order: 5	
SKU	Qty
C	1
D	12
E	6

**Item Master**

SKU	Length	Width	Depth	Cube	Weight
A	5	3	2	30	1.25
B	3	2	4	24	4.75
C	8	6	5	180	9.65
D	4	4	3	32	6.35
E	6	4	5	120	8.20



**Total Lines = 11**

**Lines per Order = 11/5 = 2.2**

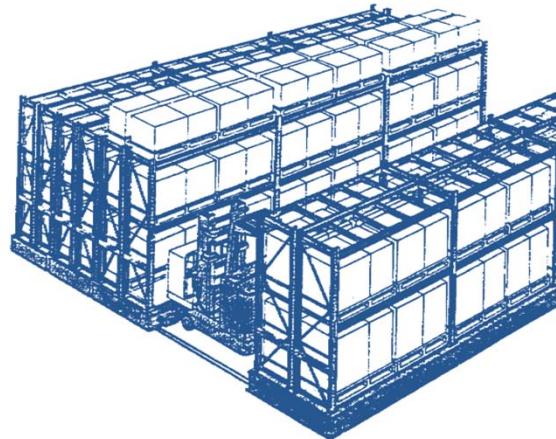
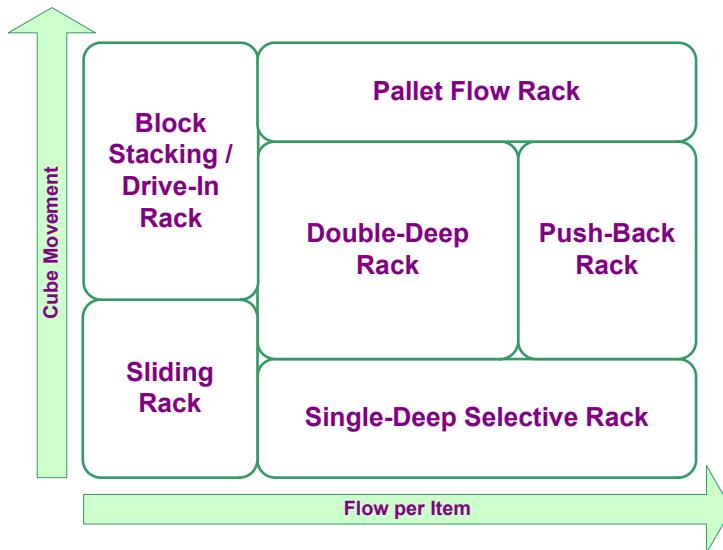
**Cube per Order = 493.2**

SKU	Flow per Item	Lines per Item	Cube Movement
A	11	3	330
B	5	2	120
C	4	3	720
D	18	2	576
E	6	1	720

**Demand Correlation Distribution**

SKU	A	B	C	D	E
A		0.2	0.4	0.2	0.0
B			0.2	0.2	0.0
C				0.4	0.2
D					0.2
E					

# Pallet Picking Equipment

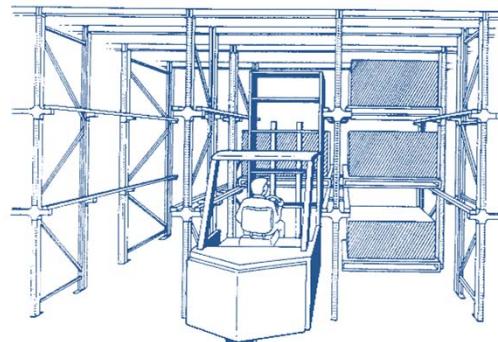


Sliding Rack

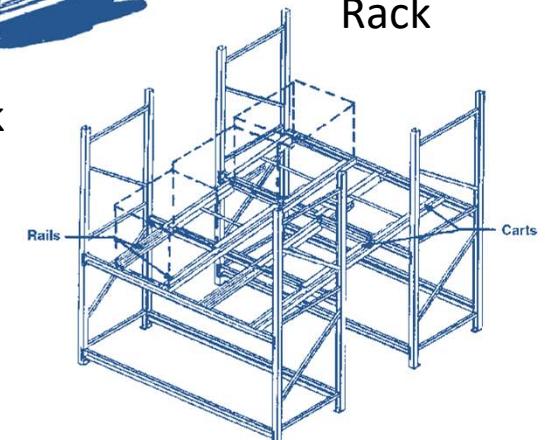


Push-Back Rack

Double-Deep Rack

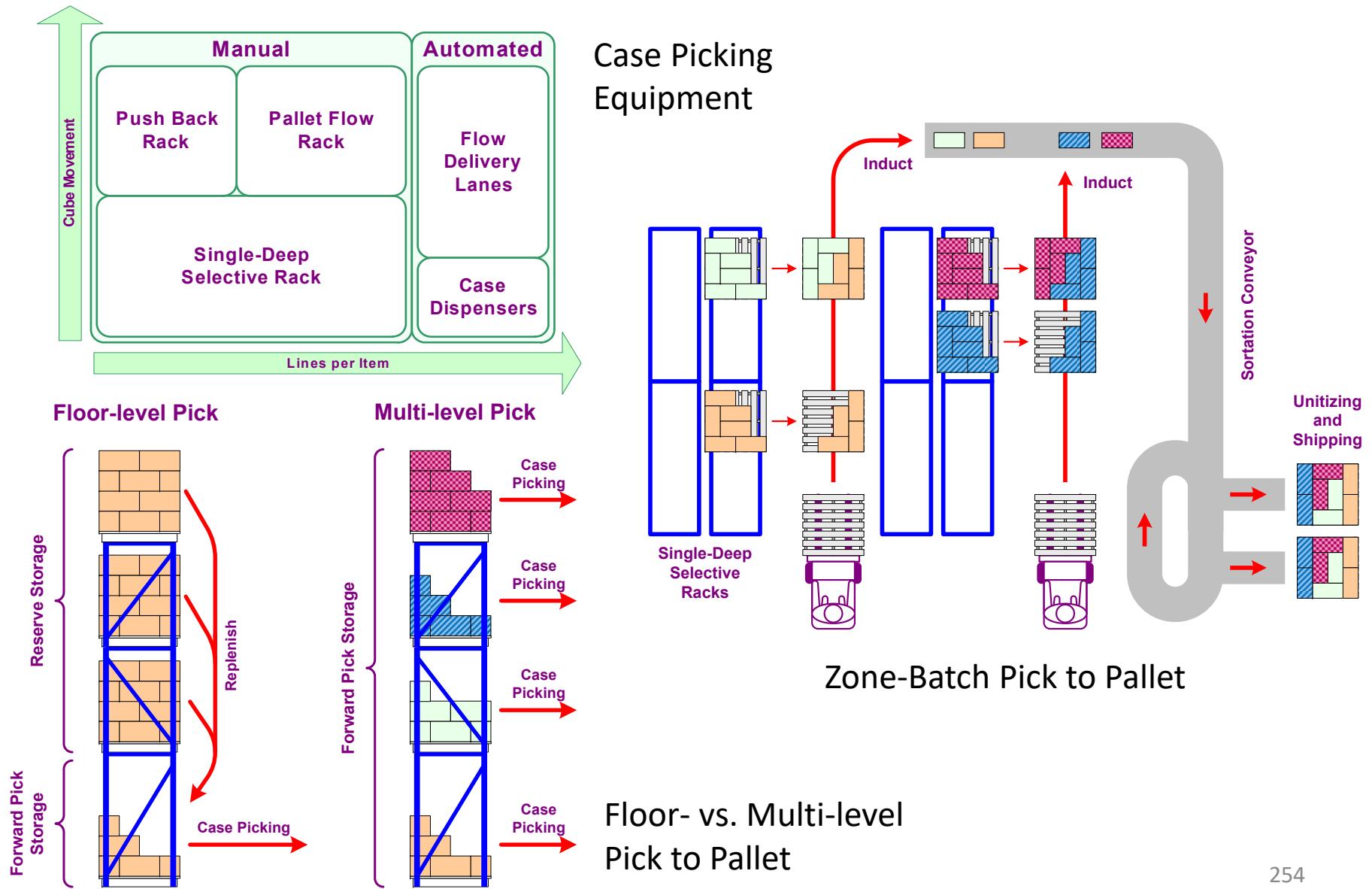


Drive-In Rack

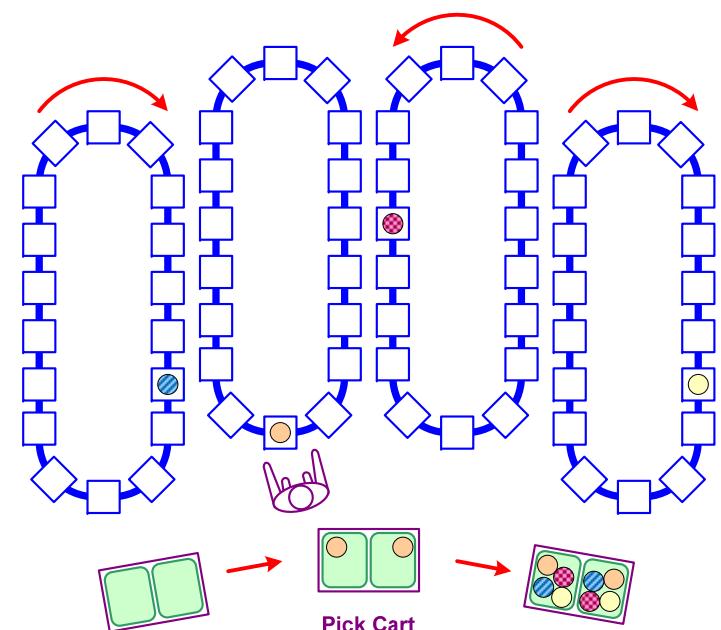
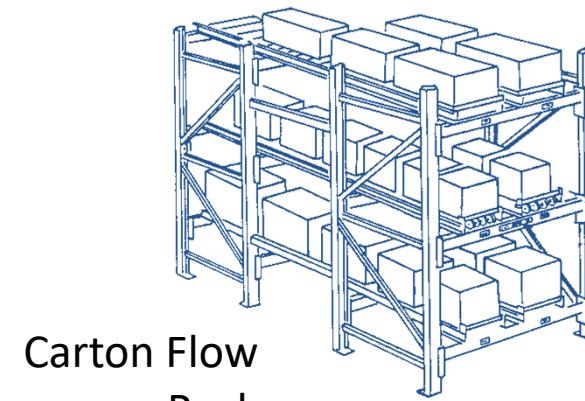
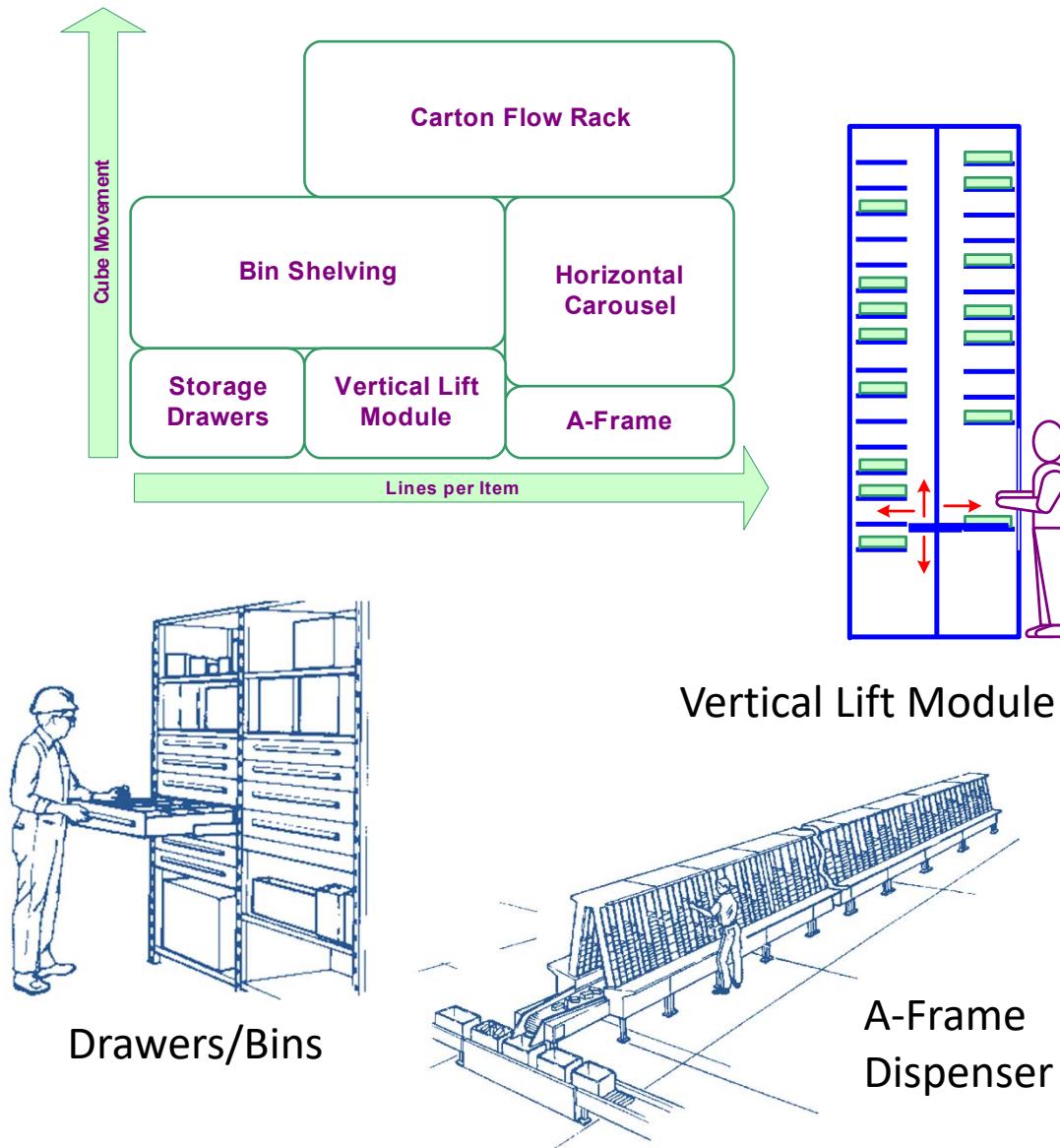


Single-Deep Selective Rack

# Case Picking



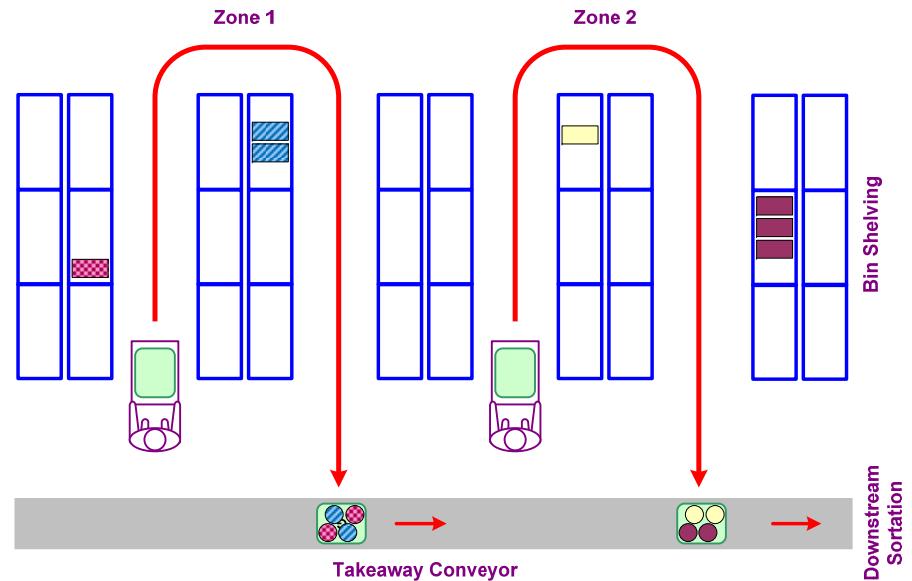
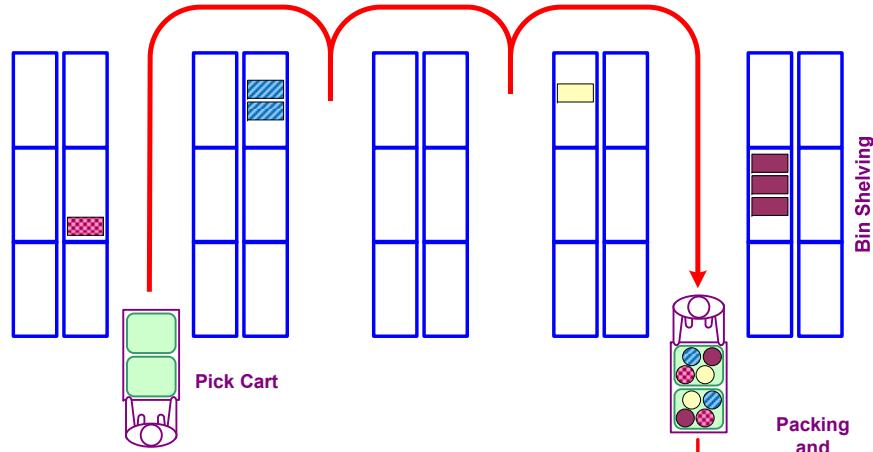
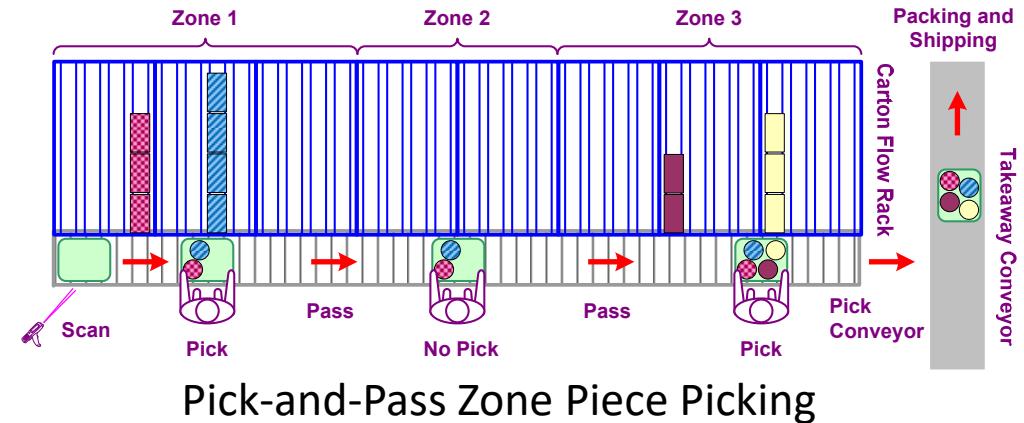
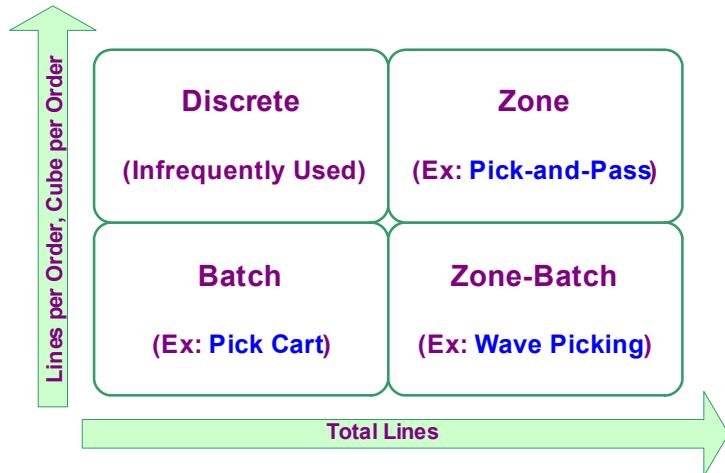
# Piece Picking Equipment



Carousel

255

# Methods of Piece Picking



# Warehouse Automation

- Historically, warehouse automation has been a craft industry, resulting highly customized, one-off, high-cost solutions
- To survive, need to
  - adapt mass-market, consumer-oriented technologies in order to realize economies of scale
  - replace mechanical complexity with software complexity
- How much can be spent for automated equipment to replace one material handler:

$$\$45,432 \left( \frac{1 - 1.017^{-5}}{1 - 1.017^{-1}} \right) = \$45,432(4.83) = \$219,692$$

- \$45,432: median moving machine operator annual wage + benefits
- 1.7% average real interest rate 2005-2009 (real = nominal – inflation)
- 5-year service life with no salvage (service life for Custom Software)

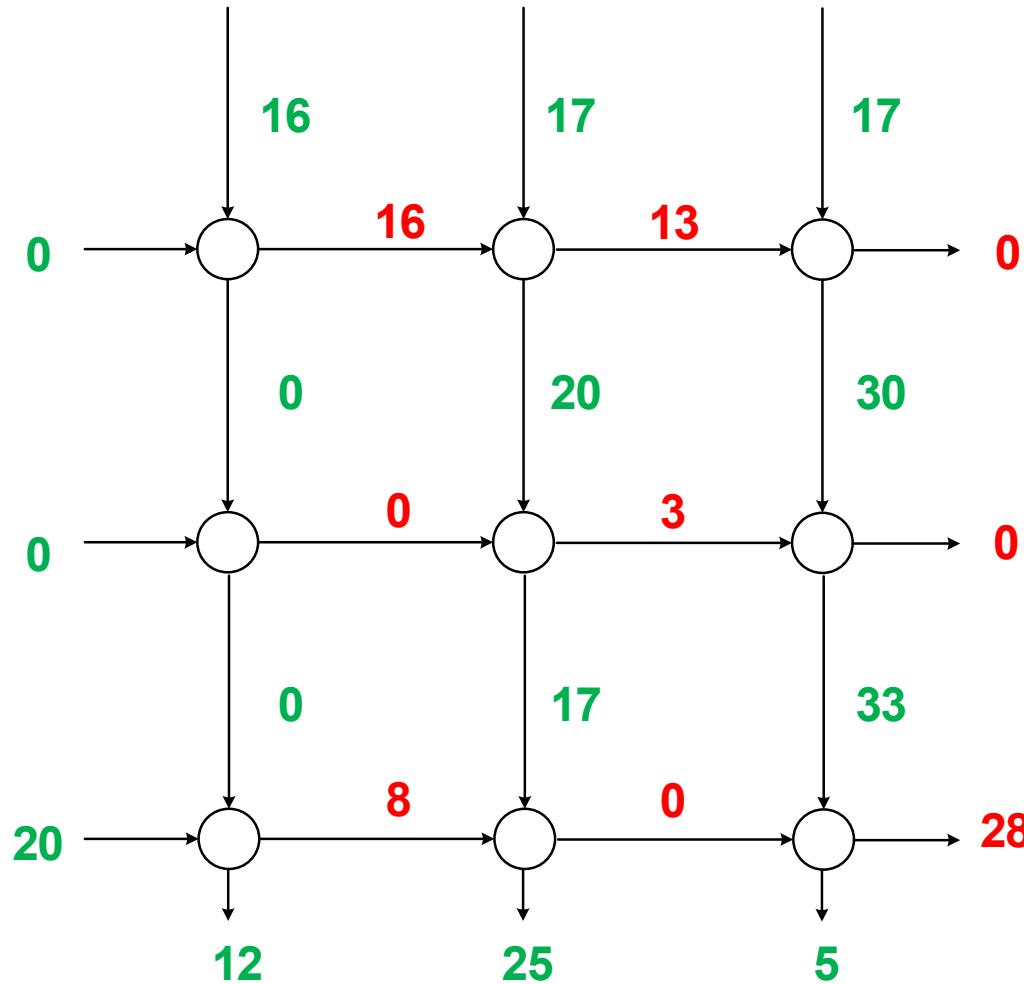
# KIVA Mobile-Robotic Fulfillment System

- Goods-to-man order picking and fulfillment system
- Multi-agent-based control
  - Developed by Peter Wurman, former NCSU CSC professor
- Kiva now called Amazon Robotics
  - purchased by Amazon in 2012 for \$775 million



# Study Guide for Final Exam

2. Second material: infer all flows from given data



# Study Guide for Final Exam

## 4. How $\min TLC$ determines TLC for a route:

$$q_{agg} = \sqrt{\frac{f_{agg} \max\{rd_{agg}, MC\}}{\alpha v_{agg} h}}$$

(no truck capacity constraints, only min charge)

$$q_i = q_{agg} \frac{f_i}{f_{agg}}$$

(allocate based on demand)

$$s_{L_j} = \sum_{i \in L_j} f_i \left/ \sum_{i \in L_j} s_i \right.$$

(aggregate density of shipments in load-mix  $L_j$ )

$$k = \min_{L_j} \left\{ 1, \frac{\min \left\{ K_{wt}, \frac{s_{L_j} K_{cu}}{2000} \right\}}{\sum_{i \in L_j} q_i} \right\}$$

(min ratio of max payload to size of shipment in load-mix)

$$q_i^* = k q_i$$

(apply truck capacity deduction factor)

$$TLC^* = \frac{f_{agg}}{\sum q_i^*} rd_{agg} + \alpha v_{agg} h \sum q_i^*$$

( $d_{agg}$  = distance of entire route)

# Study Guide for Final Exam

5. Order in which *twoopt* considers each sequence:

1:	1	2	3	4	5	6	1	38
2:	1	3	2	4	5	6	1	39
3:	1	4	3	2	5	6	1	32
4:	1	3	4	2	5	6	1	31
5:	1	4	3	2	5	6	1	32
6:	1	2	4	3	5	6	1	31
7:	1	5	2	4	3	6	1	21
8:	1	2	5	4	3	6	1	21
9:	1	4	2	5	3	6	1	32
10:	1	3	4	2	5	6	1	31
11:	1	5	4	2	3	6	1	12
12:	1	4	5	2	3	6	1	34
13:	1	2	4	5	3	6	1	40
14:	1	3	2	4	5	6	1	39
15:	1	5	2	4	3	6	1	21
16:	1	5	3	2	4	6	1	30
17:	1	5	6	3	2	4	1	31
18:	1	5	4	3	2	6	1	13
19:	1	5	4	6	3	2	1	18
20:	1	5	4	2	6	3	1	20

Local optimal sequence

Sequences considered at end to verify

local optimum:  $n$  nodes  $\Rightarrow$

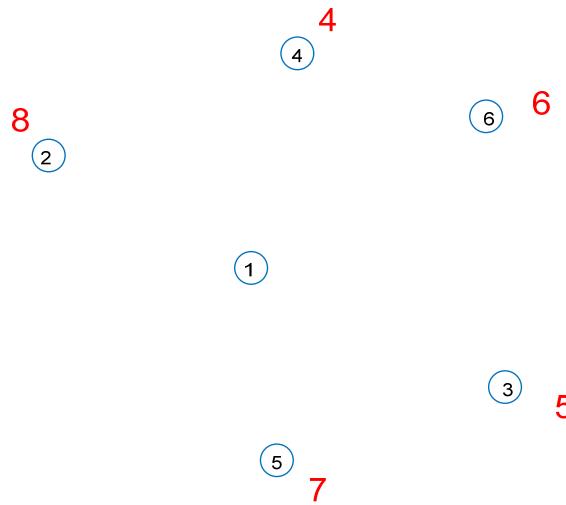
$$\sum_{i=1}^{n-2} \sum_{j=i+2}^{n-1} (l) = \frac{n^2 - 5n}{2} + 3 \text{ for } n = 6$$

# Study Guide for Final Exam

6. *vrpsavings* implements  
Clark-Wright procedure  
(truck capacity = 15)

$i$	$j$	$s_{ij}$
2	3	$40 + 48 - 87 = 1$
2	4	$40 + 38 - 46 = 32$
2	5	8
2	6	13
3	4	19
3	5	40
3	6	49
4	5	1
4	6	52
5	6	12

	1	2	3	4	5	6
1	0	40	48	38	33	48
2	40	0	87	46	65	75
3	48	87	0	67	41	47
4	38	46	67	0	70	34
5	33	65	41	70	0	69
6	48	75	47	34	69	0



# Study Guide for Final Exam

9.  $x = 3, y = 3, M = 100,000, N = 3600, D^* = 4,$   
 $L = 6710, TA = 410,652$
10.  $M = 210, D^* = 6, L_{ded} = 18, TD_{ded} = 3090,$   
 $L_{rand} = 11, TD_{rand} = 3630$
11.  $M = 50,000, D^* = 3, L = 4,195, TA = 163,605, TA =$   
 $188,146, d_{SC} = 613.43, t_{SC} = 2.00$

# Study Guide for Final Exam

13.

$$X = \sqrt{TA/2} = 31.62$$

$$X_D = \sqrt{2(TA_D / 2)} = 24.49$$

$$X'_C = \sqrt{TA_C} = 17.32 \Rightarrow X_C = X - X'_C$$

$$X'_A = \sqrt{TA_A + TA_C} = 24.49 \Rightarrow X_A = X - X'_A$$

