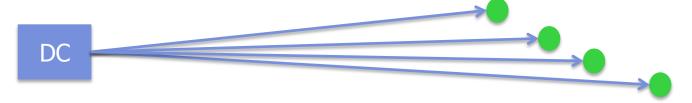
#### CTL.SC1x -Supply Chain & Logistics Fundamentals

# One to Many Distribution



### How can I distribute products?

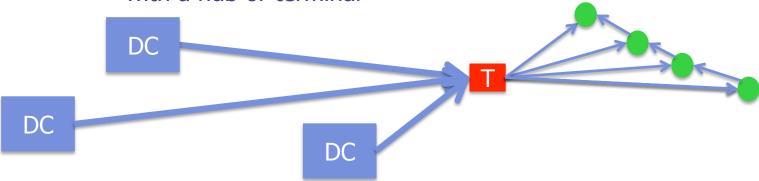
One-to-One – direct or point to point movements from origin to destination



**One-to-Many** – multi-stop moves from a single origin to many destinations

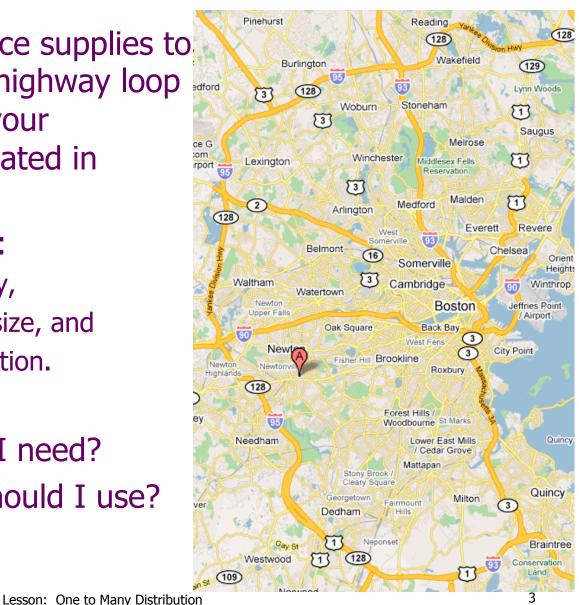


**Many-to-Many** – moving from multiple origins to multiple destinations usually with a hub or terminal



## Example: OfficeMin

- Your firm delivers office supplies to firms within the I-95 highway loop around Boston from your distribution center located in Newton.
- You want to estimate:
  - Expected cost per day,
  - 2. Expected truck fleet size, and
  - 3. Sensitivity of the solution.
- What information do I need?
- What methodology should I use?



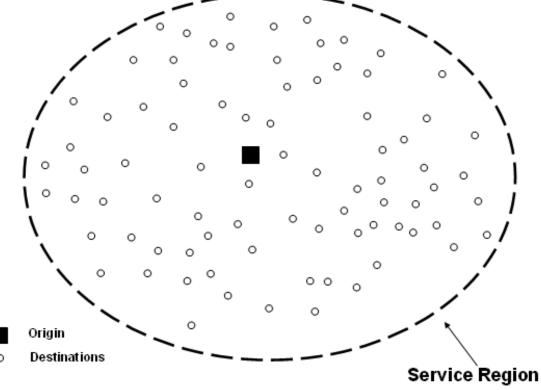
# **Defining Delivery Districts**

#### **Single Distribution Center:**

- Products originate from one origin
- Products are demanded at many destinations
- All destinations are within a specified Service Region
- Ignore inventory (same day delivery)

#### **Assumptions:**

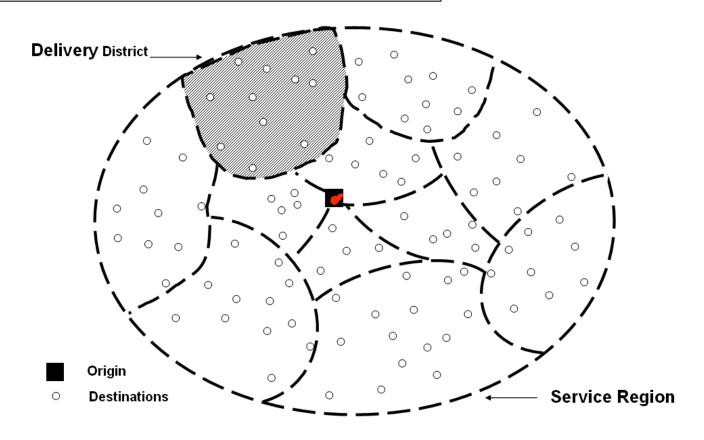
- Vehicles are homogenous
- Same capacity, Q<sub>MAX</sub>
- Fleet size is constant



Case adapted from Hernandez Lopez, J.J. (2003) "Evaluation of Bulk and Packaged Distribution Strategies in a Specialty Chemical Company," MIT Supply Chain Management Program Thesis.

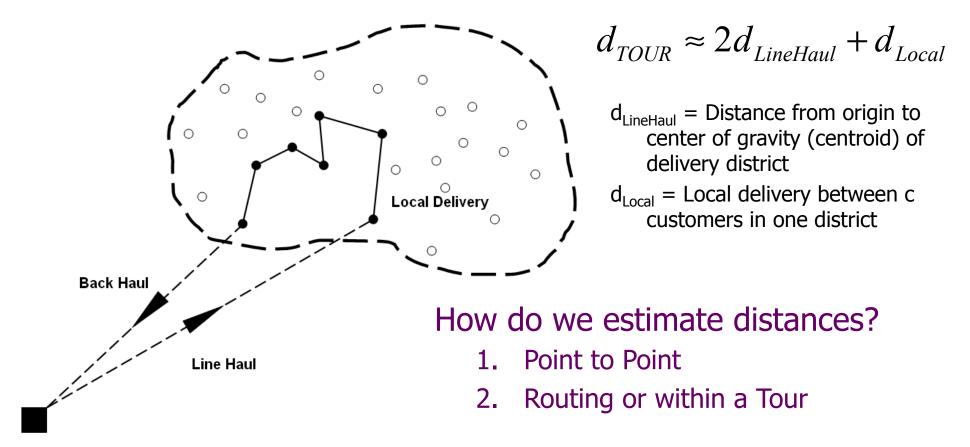
#### Finding the estimated total distance:

- Divide the Service Region into Delivery Districts
- Estimate the distance required to service each district



#### **Route to serve a specific district:**

- Line haul from origin to the 1st customer in the district
- Local delivery from 1<sup>st</sup> to last customer in the district
- Back haul (empty) from the last customer to the origin

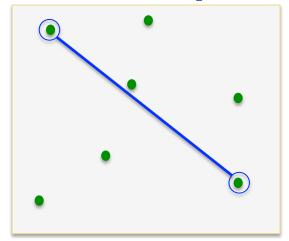


### **Estimating Point to Point Distances**

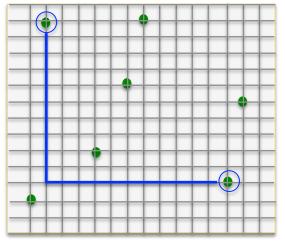
#### Distance Estimation: Point to Point

- Why bother?
- How to do it?
  - Depends on the topography of the underlying region
    - Euclidean Space:  $d_{A-B} = \sqrt{((x_A x_B)^2 + (y_A y_B)^2)}$
    - Grid:  $d_{A-B} = |x_A x_B| + |y_A y_B|$
    - Random Network: different approach

Euclidean Space (L<sub>2</sub> Metric)

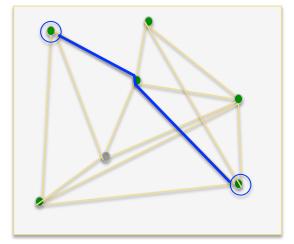


Manhattan Metric / Grid (L<sub>1</sub> Metric)



Lesson: One to Many Distribution

Random Network



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#### Distance Estimation: Point to Point

- For Random (real) Networks use: D<sub>A-B</sub> = k<sub>CF</sub> d<sub>A-B</sub>
- Find d<sub>A-B</sub> the "as crow flies" distance.
  - Euclidean: for really short distances
    - $d_{A-B} = SQRT((x_A-x_B)2+(y_A-y_B)2)$
  - Great Circle: for locations within the same hemisphere
    - $d_{A-B} = 3959(arccos[sin[LAT_A]sin[LAT_B] + cos[LAT_A] cos[LAT_B]cos[LONG_A-LONG_B]])$ Where:
    - LAT<sub>i</sub> = Latitude of point i in radians
    - LONG<sub>i</sub> = Longitude of point i in radians
    - Radians = (Angle in Degrees)(π/180°)
- Apply an appropriate circuity factor (k<sub>CF</sub>)
  - How do you get this value?
  - What do you think the ranges are?
  - What are some cautions for this approach?

# Selected Values of k<sub>CF</sub>

Country	$\mathbf{k}_{CF}$	StdDev	Country	<b>k</b> <sub>CF</sub>	StdDev
Argentina	1.22	0.15	Japan	1.41	0.15
Australia	1.28	0.17	Mexico	1.46	0.43
Belarus	1.12	0.05	New Zealand	2.05	1.63
Brazil	1.23	0.11	Poland	1.21	0.09
Canada	1.30	0.10	Russia	1.37	0.26
China	1.33	0.34	Saudi Arabia	1.34	0.19
Egypt	2.10	1.96	South Africa	1.23	0.12
Europe	1.46	0.58	Thailand	1.42	0.44
England	1.40	0.66	Turkey	1.36	0.34
France	1.65	0.46	Ukraine	1.29	0.12
Germany	1.32	0.95	United States	1.20	0.17
Italy	1.18	0.10	Alaska	1.79	0.87
Spain	1.58	0.80	US East	1.20	0.16
Hungary	1.35	0.25	US West	1.21	0.17
India	1.31	0.21			
Indonesia	1.43	0.34			

Source: Ballou, R. (2002) "Selected country circuity factors for road travel distance estimation," Transportation Research Part A, p 843-848.

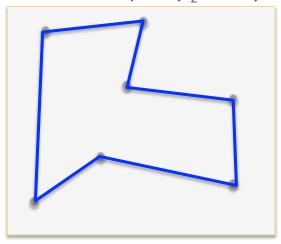
### **Estimating Route Distances**

## Distance Estimation: Routing

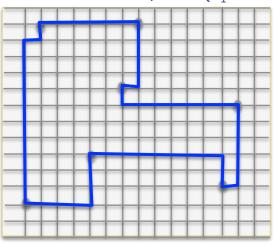
$$d_{TSP} = k_{TSP} \sqrt{(nA)}$$

- Traveling Salesman Problem
  - Starting from an origin, find the minimum distance required to visit each destination once and only once and return to origin.
  - The expected TSP distance,  $d_{TSP}$ , is proportional to  $\sqrt{(nA)}$  where n= number of stops and A=area of district
  - The factor  $(k_{TSP})$  is a function of the topology

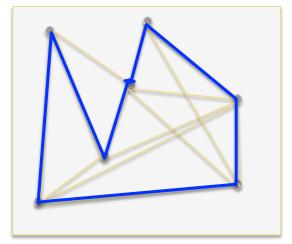
Euclidean Space (L<sub>2</sub> Metric)



Manhattan Metric / Grid (L<sub>1</sub> Metric)



Random Network



 What can we say about the expected TSP distance to cover <u>n stops</u> in district with an <u>area of A?</u> A=Area of district
n=Number of stops in district
δ=Density (# stops/Area)
k=VRP network factor (unitless)

d<sub>TSP</sub>=Traveling Salesman Distance d<sub>stop</sub>=Average distance per stop

 A good approximation, assuming a "fairly compact and fairly convex" area, is:

$$E\left[d_{TSP}\right] \approx k_{TSP} \sqrt{nA} = k_{TSP} \sqrt{n\left(\frac{n}{\delta}\right)} = \frac{k_{TSP}n}{\sqrt{\delta}}$$

- What values of k<sub>TSP</sub> should we use?
  - Lots of research on this for L<sub>1</sub> and L<sub>2</sub> networks depends on district shape, approach to routing, etc.
  - Euclidean (L<sub>2</sub>) Networks
    - $k_{TSP} = 0.57$  to 0.99 depending on clustering & size of N (MAPE~4%, MPE~-1%)
    - k<sub>TSP</sub>=0.765 commonly used
  - Grid (L<sub>1</sub>) Networks
    - $k_{TSP} = 0.97$  to 1.15 depending on clustering and partitioning of district

## **Estimating Tour Distances**

### **Estimating Tour Distance**

- Finding the total distance traveled on all tours, where:
  - I = number of tours
  - c = number of customer stops per tour and
  - n=total number of stops = c\*l

$$E[d_{TOUR}] = 2d_{LineHaul} + \frac{ck_{TSP}}{\sqrt{\delta}}$$

$$E[d_{AllTours}] = lE[d_{TOUR}] = 2ld_{LineHaul} + \frac{nk_{TSP}}{\sqrt{\delta}}$$

Minimize number of tours by maximizing vehicle capacity

$$l = \left[\frac{D}{Q_{MAX}}\right]^{+}$$

$$E\left[d_{AllTours}\right] = 2\left[\frac{D}{Q_{MAX}}\right]^{+} d_{LineHaul} + \frac{nk_{TSP}}{\sqrt{\delta}}$$

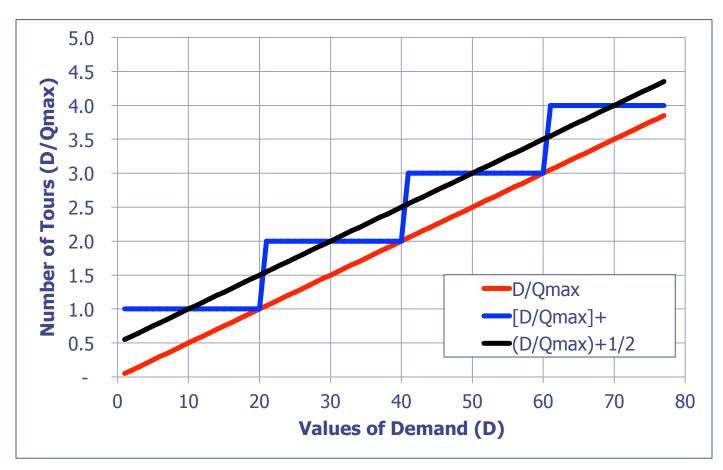
 $[x]^+$  = lowest integer value > x. This is a step function

Estimate this with continuous function:

$$E([x]^+) \sim E(x) + \frac{1}{2}$$

# **Continuous Approximation**

In this example,  $Q_{MAX}$ =20. The number of tours, I, would be  $[D/Q_{MAX}]^+$  which is a step function. Step functions are not continuous – lets create a continuous approximation of this function that we can use.



So that expected distance for all tours becomes:

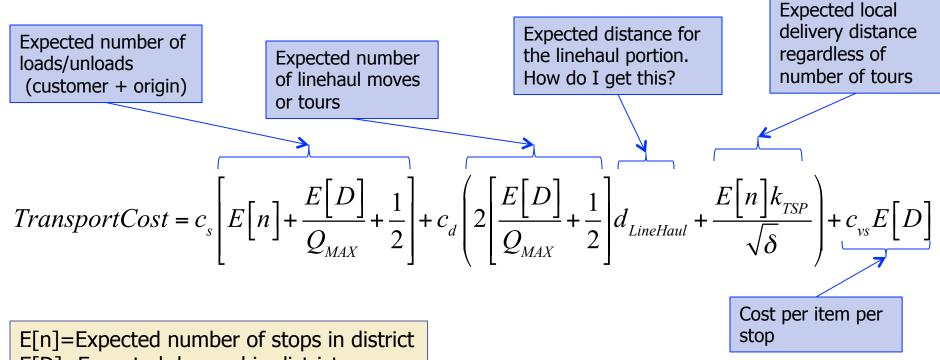
$$E\left[d_{AllTours}\right] = 2\left[\frac{E\left[D\right]}{Q_{MAX}}\right]^{+}d_{LineHaul} + \frac{E\left[n\right]k_{TSP}}{\sqrt{\delta}} = 2\left[\frac{E\left[D\right]}{Q_{MAX}} + \frac{1}{2}\right]d_{LineHaul} + \frac{E\left[n\right]k_{TSP}}{\sqrt{\delta}}$$

Note that if each delivery district has a different density, then:

$$E\left[d_{AllTours}\right] = 2\sum_{i} \left[\frac{E\left[D_{i}\right]}{Q_{MAX}} + \frac{1}{2}\right] d_{LineHaul_{i}} + k_{TSP} \sum_{i} \frac{E\left[n_{i}\right]}{\sqrt{\delta_{i}}}$$

# Putting it all together

For identical districts, the approximate transportation cost to deliver to each customer becomes:



Lesson: One to Many Distribution

E[D]=Expected demand in district

Q<sub>MAX</sub>=Capacity of each truck

c<sub>s</sub>=Cost per stop (\$/stop)

c<sub>d</sub>=Cost per distance (\$/mile)

c<sub>vs</sub>=Cost per unit per stop (\$/item-stop)

 $\delta$ =Density (# stops/Area)

 $k_{TSP}$  =TSP network factor (unitless)

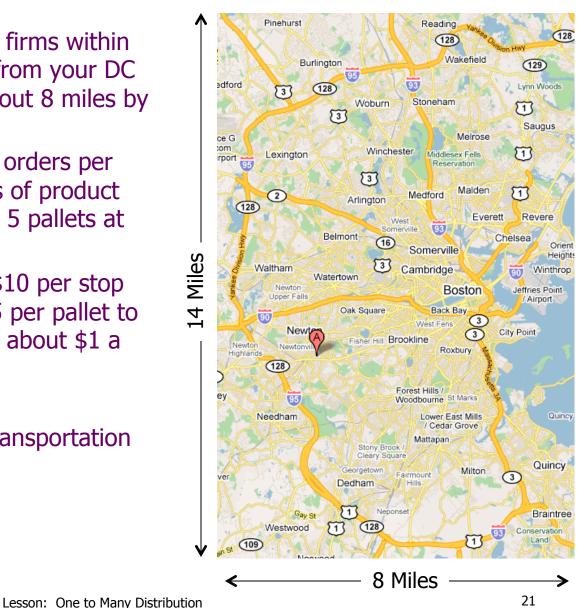
d<sub>TSP</sub>=Traveling Salesman Distance

d<sub>stop</sub>=Average distance per stop

#### Solution OfficeMin

#### OfficeMin Problem

- You deliver office supplies to firms within the I95 loop around Boston from your DC in Newton. This region is about 8 miles by 14 miles.
- You expect ~ 100 customer orders per day – for about 1 to 2 pallets of product each. Local vans can handle 5 pallets at most.
- You estimate it costs about \$10 per stop (to load or unload), about \$5 per pallet to deliver to end customer, and about \$1 a mile for driving.
- What is the expected daily transportation cost?



#### OfficeMin

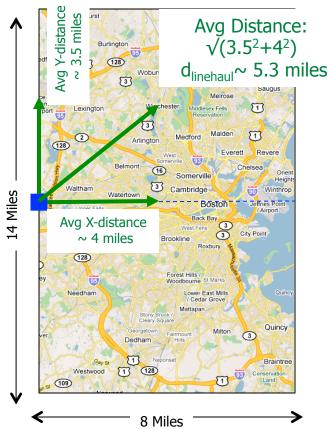
$$TransportCost = c_{s} \left[ E[n] + \frac{E[D]}{Q_{MAX}} + \frac{1}{2} \right] + c_{d} \left( 2 \left[ \frac{E[D]}{Q_{MAX}} + \frac{1}{2} \right] d_{LineHaul} + \frac{E[n]k_{TSP}}{\sqrt{\delta}} \right) + c_{vs} E[D]$$

#### What do we know?

- $c_s = 10 \$/stop$
- $c_d = 1 \text{ } /mile$
- $\mathbf{c}_{vs} = 5 \text{ } \text{/pallet}$
- E[n] = 100
- E[D] = 150
- $Q_{MAX} = 5$  pallets

#### What do we need to find?

- k = 1.15 (estimate)
- $\delta = 100/(8)(14) = 0.89 \sim 1$
- $\mathbf{d}_{linehaul} = ??$



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#### OfficeMin

$$\begin{array}{lll} c_{s} = 10 \; \text{\$/stop} & E[n] = 100 & k = 1.15 \\ c_{d} = 1 \; \text{\$/mile} & E[D] = 150 & \delta = 1 \\ c_{vs} = 5 \; \text{\$/pallet} & Q_{\text{MAX}} = 5 \; \text{pallets} & d_{\text{linehaul}} = 5 \end{array}$$

$$TransportCost = c_{s} \left[ E[n] + \frac{E[D]}{Q_{MAX}} + \frac{1}{2} \right] + c_{d} \left( 2 \left[ \frac{E[D]}{Q_{MAX}} + \frac{1}{2} \right] d_{LineHaul} + \frac{E[n]k_{TSP}}{\sqrt{\delta}} \right) + c_{vs} E[D]$$

Estimated number of tours per day:

$$l = \frac{E[D]}{Q_{MAX}} + \frac{1}{2} = \frac{150}{5} + .5 = 30.5$$

Estimated stop (load/unload) cost per day:

$$c_s [E[n] + E[l]] = 10(100 + 30.5) = $1305$$

Estimated distance (driving) cost per day:

$$c_d \left( 2E[l]d_{LineHaul} + \frac{E[n]k_{TSP}}{\sqrt{\delta}} \right) = 1\left( 2(30.5)(5) + \frac{100(1.15)}{\sqrt{1}} \right) = 305 + 115 = $420$$

Estimated stop-pallet costs per day:

$$c_{vs}E[D] = 5(150) = $750$$

Estimated total daily cost ~ \$2400 to \$2500

# Estimating Fleet Size [OPTIONAL – MORE ADVANCED]

## Estimating the Fleet Size

- Find minimum number of vehicles required based on the amount of required work time each day where
  - M = minimum number of vehicles needed in fleet
  - t<sub>w</sub> = available worktime for each vehicle per period
  - W = required amount of work time each day
  - s = average vehicle speed
  - I = number of shipments per period
  - t<sub>1</sub> = loading time per shipment
  - t<sub>s</sub> = unloading time per stop

$$Mt_{_{\scriptscriptstyle{W}}} \geq W$$

$$W = \frac{d_{AllTours}}{s} + lt_l + nt_s$$

$$W = \frac{\left(2E\left[l\right]d_{LineHaul} + \frac{E\left[n\right]k_{TSP}}{\sqrt{\delta}}\right)}{S} + E\left[l\right]t_{l} + E\left[n\right]t_{s}$$

$$W = \left(\frac{2d_{LineHaul}}{s} + t_l\right) E[l] + E[n] \left(\frac{k_{TSP}}{s\sqrt{\delta}} + t_s\right)$$

#### Fleet Size

Note that W is a linear combination of two random variables, **n** and **D**. But, they are not independent, in fact,  $D = nD_c$  where  $D_c$  is the number of pallets per customer

of pallets per customer 
$$W = \left(\frac{2d_{LineHaul}}{S} + t_l\right) \left[\frac{D}{Q_{MAX}} + \frac{1}{2}\right] + n\left(\frac{k_{TSP}}{S\sqrt{\delta}} + t_s\right) \qquad Var[D_c] = \frac{\sum_{i=1}^{n} \left(x_i - \overline{x}\right)^2}{n} = 0.25$$

$$a = \left(\frac{2d_{LineHaul}}{S} + t_l\right) \left[\frac{1}{Q_{MAX}}\right] = 0.144 \, hrs$$

$$W = a$$

$$W = a$$

$$W = a$$

$$b = \left(\frac{k_{TSP}}{s\sqrt{\delta}} + t_s\right) = 0.525 \, hrs$$

$$c = \frac{1}{2} \left( \frac{2d_{Linehaul}}{s} + t_l \right) = 0.361 \, hrs$$

$$+ t_{s}$$
  $Var[D_{c}] = \frac{\sum_{i=1}^{n} (x_{i} - \overline{x})^{2}}{n} = 0.25$ 

$$W = aD + bn + c$$

$$W = aD_{c}n + bn + c$$

$$W = (aD_{c} + b)n + c$$

Setting  $X=aD_c+b$ , W is now a function of random variables, X and n.

$$W = Xn + c$$

$$E[W] = E[X]E[n] + c$$

$$Var[W] = E[n]Var[X] + E[X]^{2}Var[n]$$

#### Fleet Size

$$W = \left(\frac{2d_{LineHaul}}{s} + t_l\right) \left[\frac{D}{Q_{MAX}} + \frac{1}{2}\right] + n\left(\frac{k_{TSP}}{s\sqrt{\delta}} + t_s\right)$$

$$a = 0.144 hrs$$

$$b = 0.525 hrs$$

$$c = 0.361 hrs$$

$$E[D_c] = 1.5 \ pallets$$
$$Var[D_c] = 0.25$$

$$W = (aD_c + b)n + c = (0.144D_c + 0.525)n + 0.361 = Xn + c$$

$$E[X] = aE[D_c] + b = 0.144(1.5) + 0.525 = 0.741$$
  $E[X] = a^2Var[D_c] = 0.021(0.25) = 0.00525$   $Var[X] = a^2Var[D_c] = 0.021(0.25) = 0.00525$ 

$$E[n] = 100 \ customers$$
  
 $Var[n] = 400 \ (assume \ \sigma_n = 20)$ 

$$E[W] = E[X]E[n] + c$$

$$E[W] = (0.741)(100) + 0.361 = 74.46$$

$$Var[W] = E[n]Var[X] + E[X]^{2}Var[n]$$

$$Var[W] = (100)(0.00525) + (0.741)^{2}(400) = 220$$

Lesson: One to Many Distribution

Distribution of required daily work hours:  $\mu_W \sim 75 \text{ hrs}$   $\sigma_W \sim 15 \text{ hrs}$ 

#### Fleet Size

- Daily distribution of required work time  $\sim N(75, 15)$
- Set the fleet size (M) to match our level of risk how?

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Note: This is not the TSP k!
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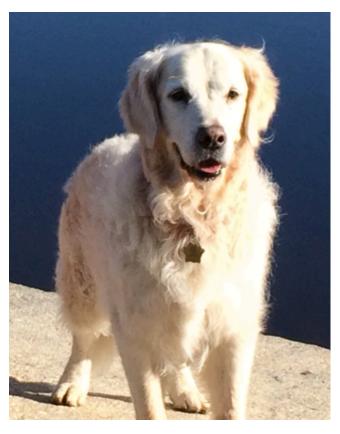
- Select a cycle service level (CSL) equal to P[W<Mt<sub>w</sub>]
  - Set M=  $(\mu_W + (k_{CSL})\sigma_W) / t_w$
  - M(80%) = (75 hrs + 0.84(15 hrs))/(10 hrs/veh) = 8.00 = 8
  - M(90%) = (75 hrs + 1.28(15 hrs))/(10 hrs/veh) = 9.42 = 10
  - M(95%) = (75 hrs + 1.64(15 hrs))/(10 hrs/veh) = 9.96 = 10
  - M(99%) = (75 hrs + 2.33(15 hrs))/(10 hrs/veh) = 10.99 = 11
- Using very few, very rough estimates of input values, we can get a feel for the trade-offs between costs and service.
- Approximations can be used for sensitivity analysis.

# **Key Take Aways**

# Key Take Aways

- Many forms of product flow: One-to-One, One-to-Many, Many-to-One, and Many-to-Many
- Approximations are good "first steps"
  - Require minimal data
  - Allow for fast sensitivity analysis
  - Enables quick scoping of the solution space
- Optimal methods usually require tremendous amounts of detailed information
- Planning problems usually have lots of uncertainty the actual conditions are unknown
- Before deciding to spend the time and energy to find an optimal solution, it is helpful to see if it is worth it.

# CTL.SC1x -Supply Chain & Logistics Fundamentals Questions, Comments, Suggestions? Use the Discussion!



"Dexter – continuously approximating" Yankee Golden Retriever Rescued Dog (www.ygrr.org)

