

# Amazon: optimising the journey of a package

## Examples of optimization and machine learning applications

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Some slides are from other Amazonians' Informs presentations  
All photos are from [www.aboutamazon.com](http://www.aboutamazon.com)



# Agenda

- A glimpse of Amazon fulfilment
- Examples working backwards from the customer
  - Reinforced learning example in the **last mile**
  - Optimization example in the **middle mile**
  - Optimization example in the **first mile**

# The journey of a package at a glimpse

What makes ordering special for you at amazon.com?

For me... **the delivery estimate.**

Give fast **delivery estimate**, then **fulfil to delivery estimate**

Raise the bar of customer's expectations

Customer **expects even faster delivery**

Customer looks at an item

Amazon gives a delivery estimate

Customer orders

Amazon decides where to fulfil from – decide what units to pack together if ordering multiple items

Amazon decides the route to take – Amazon 1P fulfilment / 3P fulfilment, what route, when

# Fulfillment at Amazon



\$24<sup>20</sup> (\$0.61 / Count)

✓prime

FREE delivery **Tomorrow, 23 August.**  
Order within **7 hrs 47 mins.** [Details](#)

Add to Cart

Buy Now

Customer  
places  
order

Initial  
fulfillment plan

12:00 FC1 → DS1  
16:00 AMZL last-  
mile

Picking



Packing



Final  
fulfillment plan

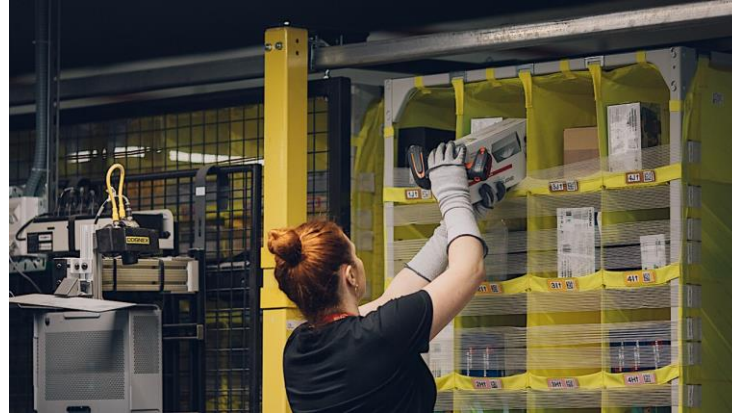
16:00 FC1 → DS1  
05:00 AMZL last-  
mile

Delivered



Daniel Chen, Informs 2023

## Inbound and stow



Older (orange) drives

## Outbound, pick, package, SLAM – Ship Label Apply Manifest



Docks and loading to linehaul trucks



ROBIN – ROBotic INduction



Sub-same day station



AR facility



Newer drives

# Selection is in a different order of magnitude



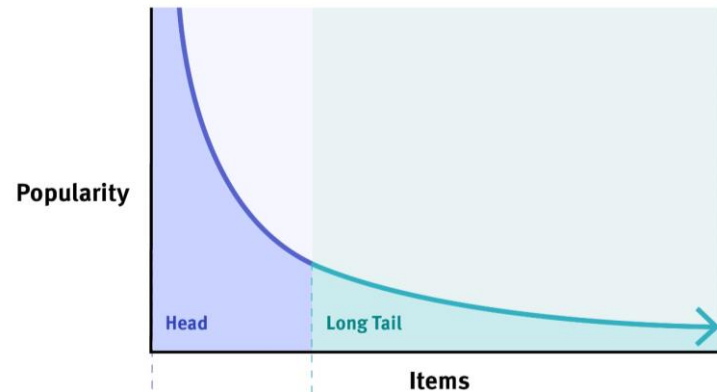
$\sim 10^5$  unique products

vs.



$\sim 10^7$  unique products

... and demand is very skewed



Also known as *the Long Tail*.<sup>1</sup>

<sup>1</sup>Anderson, C. *The Long Tail: Why the Future of Business is Selling Less of More*, 2006.

Vendors / suppliers

1000s



**First mile:**  
inbound network,  
distribute supply to  
fulfilment centres.

Fulfilment centres.  
Holds inventory.

100s of buildings  
10s millions of unique items in  
inventory

10s millions of packages  
1000s trucks  
100s flights



**Middle mile:**  
outbound network,  
Maximise speed,  
minimize cost

**Last mile:** delivery to customer

Delivery stations

Customer



10, 000s vehicles

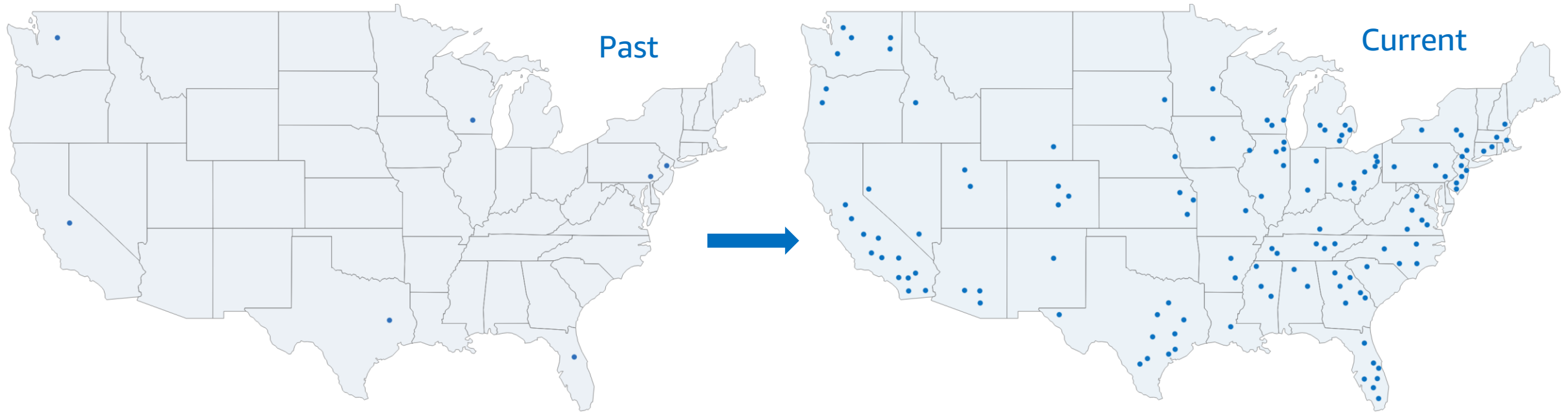
10s millions of  
orders

[www.aboutamazon.com](http://www.aboutamazon.com)



# The Everything Store Problem

- Start with few Fulfillment Centers (FCs), with thick lanes
- Fast-forward to 2024, 200+ FCs that can connect to any customer
- Network complexity spirals - hard to operate and get good signals for inventory placement



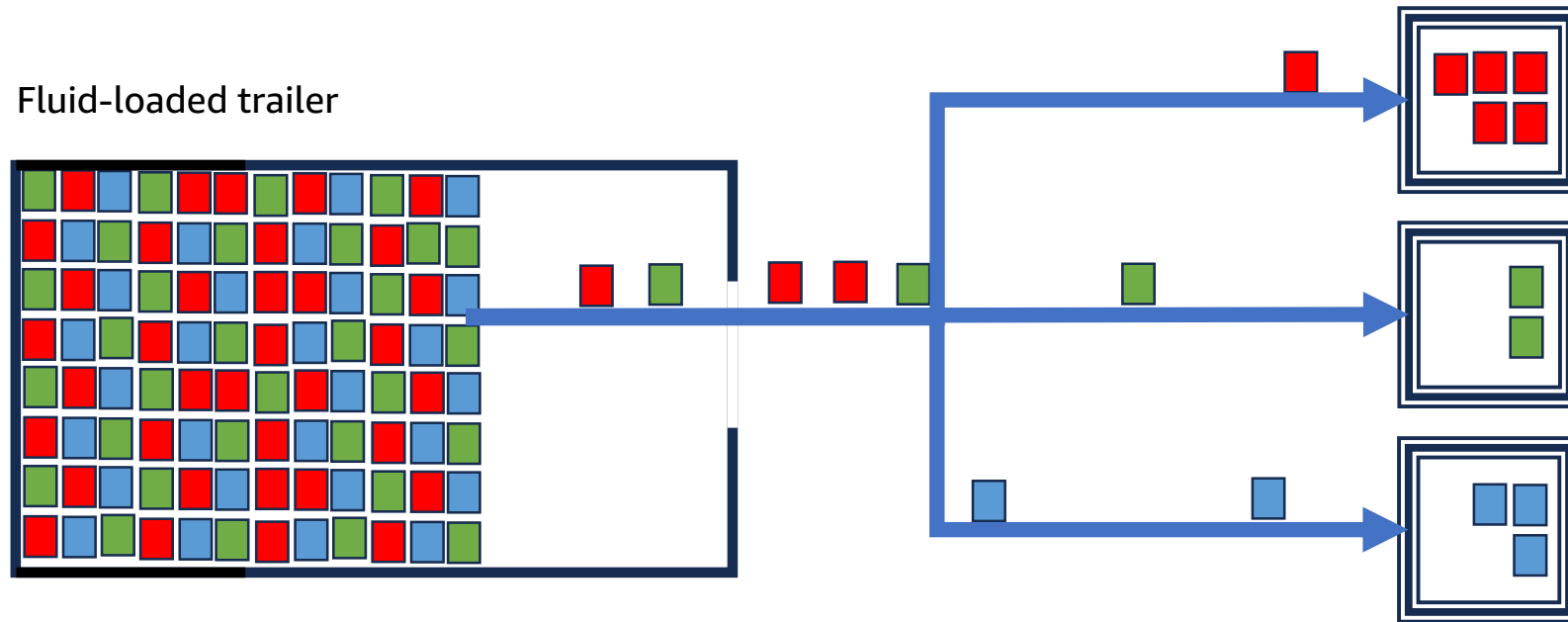
Not actual FC locations. Plot is only for illustration purpose.

- On the bright side, we now have many copies of FCs with fungible inventories
- A perfect opportunity to divide and conquer: **Regionalization**

Yuan Li, Informs 2023

# Sortation centers (Inbound Cross Dock and Middle Mile Sort Centers)

Each Sort is a package level sortation process that sorts each package to its assigned destination



Last mile: delivery to customer

Delivery stations  Customer

Tactical:

- Vehicle routing problem
- Labour / driver demand

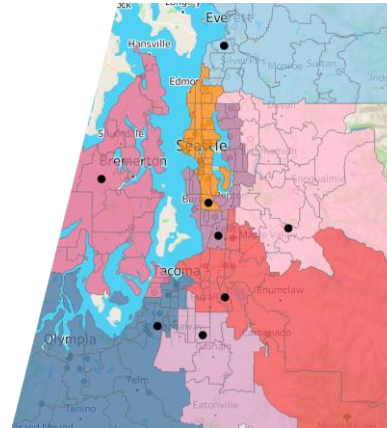


10, 000s vehicles

10s millions of orders

Strategic:

- Number and placement of delivery stations
- Throughput capacity
- Jurisdictions of delivery stations
  - how many customers
  - expected demand / workload
  - resources

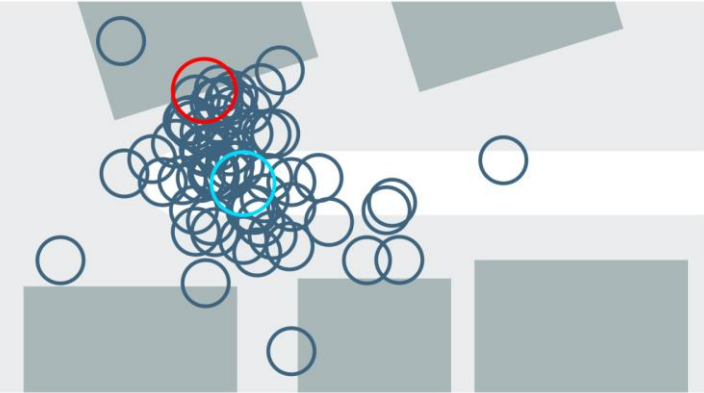


# Last mile example machine learning problem

Getting Your Package to the Right Place: Supervised Machine Learning for Geolocation

Amazon Last Mile delivers millions of packages daily

The geolocation of each address is needed for partitioning and optimizing routes among vehicles



*Learning to rank or machine-learned ranking is the application of machine learning, typically supervised, semi-supervised or reinforcement learning, in the construction of ranking models for information retrieval systems. (Wikipedia)*

Label	Machine Learning	find	*4251106TH AVE NE BELLEVUE*					
IF Auto-advance after labeling	AB-A							
Select next unlabeled address	P9							
Select parent xor child	P12							
events	DP1.moved	DP1.loss	DP2.loss	Δ.loss	Label	Score	bank?	
40	4.3	3.5	7.8	-4.3	manual+Point...	0.010		
6	6.9	1.0	7.7	-6.7	manual+Point...	0.018		
49	0.0	2.2	2.2	0.0	manual+Point...	0.017		
55	3.4	2.6	1.6	1.0	manual+Point...	0.017		
68	9.2	0.5	9.3	-8.8	manual+Point...	0.008		
53	0.8	3.9	4.0	-0.0	manual+Point...	0.017		
78	6.4	1.7	6.2	-4.5	manual+Point...	0.017		
20	6.8	6.5	2.2	4.4	P@only1; W6S...	0.010		
5	6.3	5.5	2.0	3.5	manual+Point...	0.034		
79	5.4	4.5	4.7	-0.2	manual+Point...	0.017		
11	3.6	3.3	2.8	0.5	P@only1; W6S...	0.017		
66	5.4	5.0	0.9	4.1	P@only1; W6S...	0.018		
2	16.4	14.2	9.0	5.1	manual+Point...	0.017		
4251106TH AVE NE BELLEVUE KING WASH	511	9.8	17.4	46.1	-28.8	manual+Point...	0.258	
BELLEVUE KING WASH	2	5.2	17.1	19.9	-2.8	manual+Point...	0.037	
VE BELLEVUE KING WASH	2	10.4	34.2	26.7	-7.5	manual+Point...	0.017	
BELLEVUE KING WASH	6	18.1	18.8	5.4	11.4	manual+Point...	0.036	
BELLEVUE KING WASH	2	5.6	19.7	25.3	-5.6	manual+Point...	0.017	
EI BELLEVUE KING WASH	5	17.4	17.4	2.5	14.9	manual+Point...	0.055	
IKING WASH	13	89.4	NaN	NaN	NaN X scattered ma...	0.045		
MSHSHNOCHMISH WASH	3	5.2	NaN	NaN	NaN X too few event...	0.018		
SHMISHSHNOCHMISH WASH	120	1.9	NaN	NaN	NaN X no buildings ...	0.011		
JAH KING WASH	4	728.4	NaN	NaN	NaN X too few event...	0.170		
UE KING WASH	2	NaN	NaN	NaN	NaN X too few event...	0.219		
IKING WASH	75	6.2	7.2	1.1	6.1	manual+Point...	0.024	
ING WASH	94	5.0	3.2	2.1	1.2	P@only1; W6S...	0.010	
SHSHNOCHMISH WASH	2	NaN	NaN	NaN	NaN X too few event...	0.018		
IKING WASH	17	3.1	2.7	9.4	2.4	P@only1; W6S...	0.009	
IKING WASH	30	4.0	3.9	0.5	3.4	P@only1; W6S...	0.017	
AYINORTH BEND KING WASH	75	4.3	NaN	NaN	NaN X no buildings ...	0.018		



# Middle mile

## Tactical:

- where to fulfil from, including what to package together
- route selection, including 1p / 3p
- number of trucks (short / middle / long term)
- capacity management
- minimize fulfilment cost
- maximize speed

## Strategic:

- where to place facilities
- capacities
- connectivity



Since early 2024: +Florida

100s of buildings  
10s millions of unique items in inventory

Fulfilment centres.  
Holds inventory.

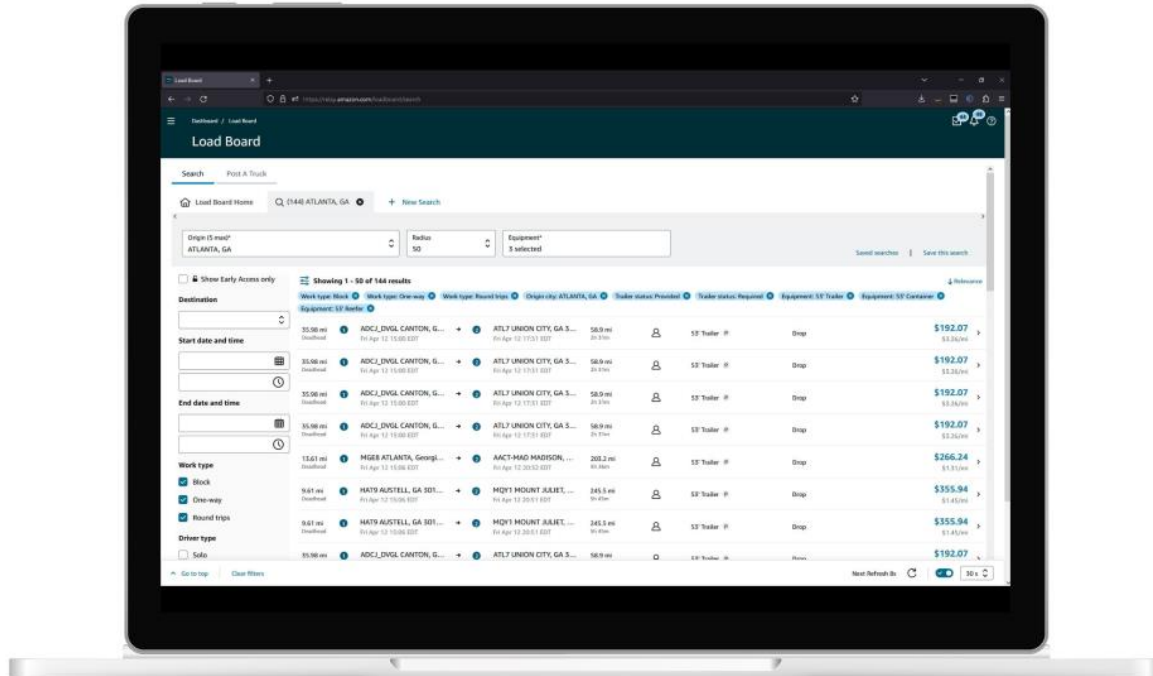


10s millions of packages  
1000s trucks  
100s flights

**Middle mile:**  
outbound network,  
Maximise speed,  
minimize cost

Delivery stations

# Middle mile example optimization and forecasting problem: The relay load board and the spot market



## What does Relay offer?

### For Carriers

A technology toolset with everything you need to haul loads for Amazon.

### Short-Term Contracts

Secure full work-weeks for your drivers, and grow your fleet by locking in revenue for provided trucks with single or multi-week contracts several weeks in advance.

### Auctions

Bid on contracts with full transparency into the timing remaining in the auction and the current lowest bid.

### Loadboard

Enjoy exclusive access to Amazon's spot work and book loads with all-in pricing at the click of a button.

### Post A truck

Reduce empty miles and idle time by sharing when and where your trucks are available. Relay will automatically book loads matching your criteria so you can focus your time on growing the business.

### Rewards

Access exclusive discounts on fuels, tires, maintenance and more.



Join the Amazon Freight Partner program for long-term, consistent work.

[Learn more](#)



Pictures from relay.amazon.com

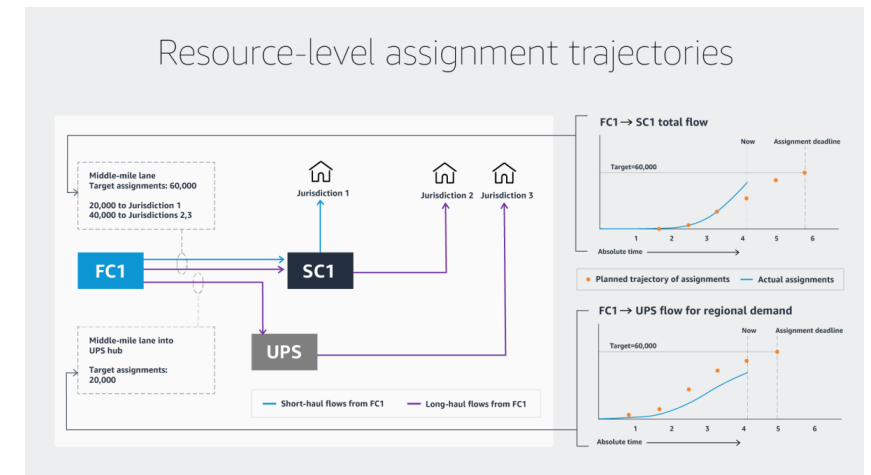
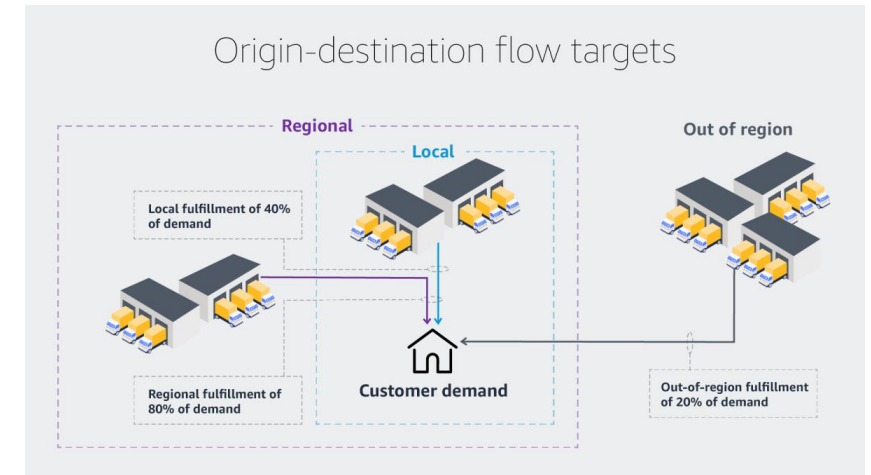
# Middle mile example optimization problem: Origin and route selection with capacity management

- A simplified example:

	Path A	Path B
Shipment 1 Costs	1	2
Shipment 2 Costs	1	10
Capacity	1	infinity

- Greedy online solution: shipment 1 -> Path A, shipment 2 -> path B
- Optimal solution: shipment 1 -> Path B, shipment 2-> Path A.

Special case: the cost is the same, but we can deliver faster in aggregate if we preserve capacity.



# Lagrangian decomposition

$$\begin{aligned}
 & \text{Minimize} && \sum_{s \in S} \sum_{p \in P_s} c_{s,p} x_{s,p} + \sum_{r \in R} c_r o_r \\
 & \text{s.t.} && \\
 & \text{(upper-bound)} && \sum_{s \in S} \sum_{p \in P_s \cap P_r} \beta_{s,r} x_{s,p} \leq U_{rt} + o_r \quad \forall r \in R^+ \\
 & \text{(lower-bound)} && \sum_{s \in S} \sum_{p \in P_s \cap P_r} \beta_{s,r} x_{s,p} \geq L_{rt} - o_r \quad \forall r \in R^- \\
 & \text{(demand)} && \sum_{p \in P_s} x_{s,p} = 1 \quad \forall s \in S \\
 & && 0 \leq x_{s,p} \leq 1, o_r \geq 0
 \end{aligned}$$



$$\begin{aligned}
 & \text{Minimize} && \sum_{s \in S} \sum_{p \in P_s} \left( c_{s,p} + \sum_{r \in R_p \cap R^+} \alpha_r \beta_{s,r} - \sum_{r \in R_p \cap R^-} \alpha_r \beta_{s,r} \right) x_{s,p} \\
 & && + \sum_{r \in R} (c_r - \alpha_r) o_r - \sum_{r \in R^+} \alpha_r U_{rt} + \sum_{r \in R^-} \alpha_r L_{rt} \\
 & \text{s.t.} && \\
 & \text{(demand)} && \sum_{p \in P_s} x_{s,p} = 1 \quad \forall s \in S \\
 & && 0 \leq x_{s,p} \leq 1, o_r \geq 0
 \end{aligned}$$

Statistic	Problem 1	Problem 2	Problem 3
Number of Resource Constraints	~ 1.6 million	~ 1.6 million	~ 1.6 million
Number of Shipments	~ 3.5 million	~ 3.6 million	~4.3 million
XPRESS Time (seconds)	8,794	15,034	25,748
LD Optimality Gap at 1800 seconds	4.6%	3.4%	4.7%



# Middle mile - topology design

## Demand variability resiliency: the case for consolidation

- Comparison of **two** network design principles.

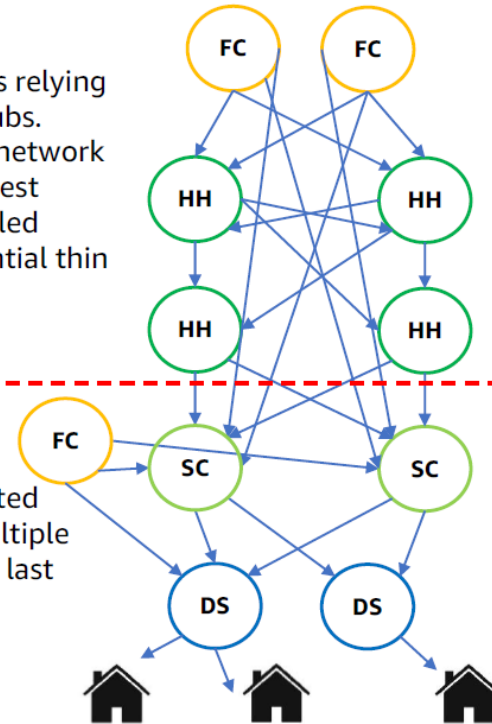
### [BAU] Business as Usual – shortest path

#### National

Multi-SC paths relying on national hubs. Complex hub network favorizing lowest distance traveled through potential thin lanes.

#### Regional

Mix of direct and consolidated paths with multiple injections into last mile nodes.



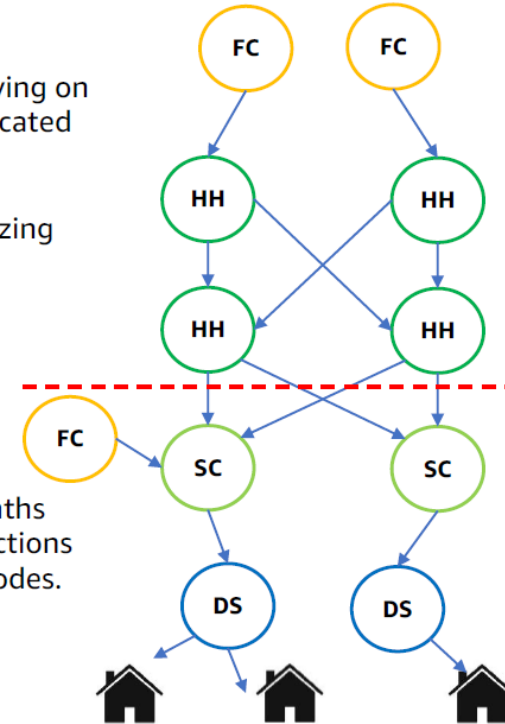
### [Consolidation] Simplified connectivity - SC

#### National

3-SC paths relying on hubs with dedicated operations. Simplified hub network favorizing thick lanes

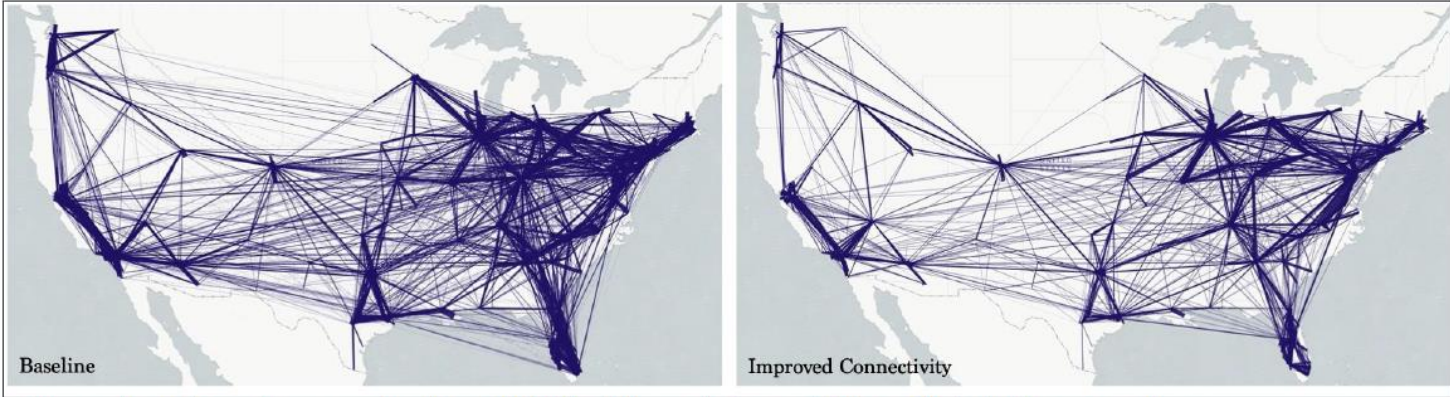
#### Regional

High frequency consolidated paths with single injections into last mile nodes.



# Middle mile – topology design

## Consolidated network: less lanes with higher volume



Map showing the arcs in the *BAU* (left) and *Consolidated* (right) networks.

- More consolidated design leads to **high Hub network utilization**, both in terms of number of arcs and in volume.
- Overall reduction in number of arcs, which are also shorter, translating into a **reduction in 'last' trucks and long distance trucks.**
- **Increase in number of touches.**

Arc type	Arc count	Pkg count
FC->DS	-71.6%	-41.7%
FC->SC	-37.3%	7.8%
SC->DS	-37.5%	13.6%
<b>SC-&gt;SC</b>	<b>25.4%</b>	<b>168.4%</b>
<b>Total</b>	<b>-37.0%</b>	<b>16.9%</b>

Arc count weighed package count

Difference between *Consolidated* and *BAU* networks.

Vendors /  
suppliers

1000s



**First mile:**  
inbound network,  
distribute supply to  
fulfilment centres.

Fulfilment centres.  
Holds inventory.

### Tactical

- Inventory replenishment
- Inventory placement / distribution

### Strategic:

- Where to build warehouses
- Inbound facilities
- Inbound network connectivity

# First mile example optimization problem: Topology design, vendor and FC connectivity, inventory distribution

## Logistics Network Design Problem

Output of an  
inventory  
placement model

- Multicommodity Capacitated fixed-charge network design problem (mcFND)
- Commodities  $(o_i, d_i)$  with  $n_i$  packages and volume  $cb_i$  to be routed via path  $P_i$  along arcs  $a$  with  $u_a$  units of capacity. Variable cost  $c_p$  per path and fixed charge cost  $f_a$  per unit of arc capacity.
- Assign commodities to feasible path so as to minimize cost
- NP-hard problem

$$\begin{aligned} \min \quad & \sum_{a \in A} f_a n_a + \sum_{i \in \mathcal{O}, p \in \mathcal{P}_i} c_p f_p \quad (\text{IP}) \\ \text{s.t.} \quad & \sum_{p \in \mathcal{P}_i} f_p = n_i \quad (i \in \mathcal{O}) \\ & \sum_{p \in \mathcal{P}_a} cb_p f_p \leq u_a n_a \quad (a \in A) \\ & f \geq 0, n \in \mathbb{Z}_+^A \end{aligned}$$

Path descriptions  
include under the  
roof constraints like  
sortation

Thank you

