

TRANSIT REDESIGNS AND COVID-19 RECOVERY PLANS USING LOCUS

presented by

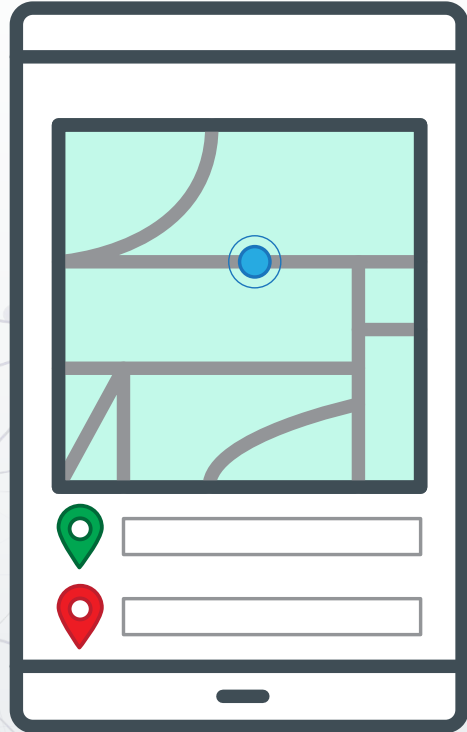
Cambridge Systematics, Inc.



**CAMBRIDGE
SYSTEMATICS**

June 2020

What are LBS Data?



- Smartphones
- Location data collection by Apps
- GPS quality
- User permission



LOCATION-BASED SERVICES DATA

DATA CHARACTERISTICS

- Large sample size
- Messy datasets
- Spatial precision
- Persistent ID
- Anonymized

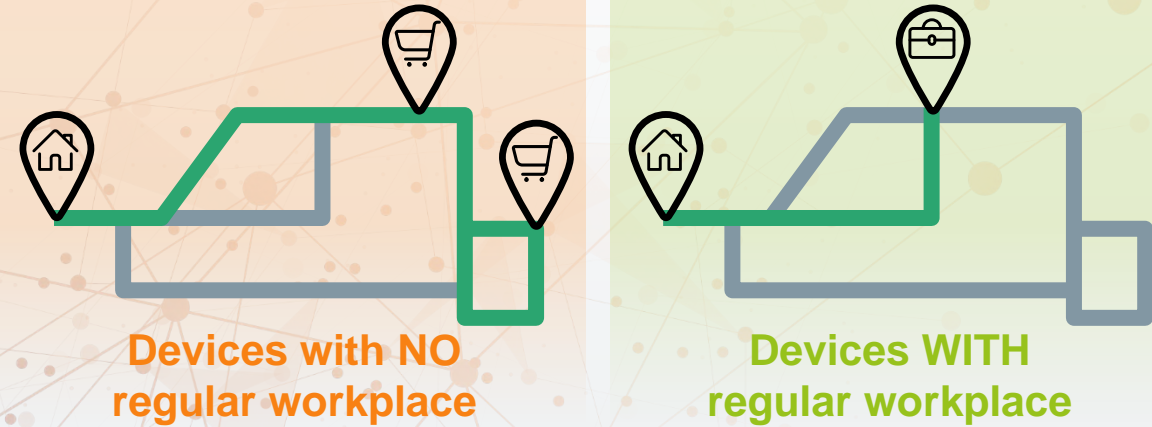
Four Core Algorithms

IDENTIFY TRIP ENDS

Process Activity Stays

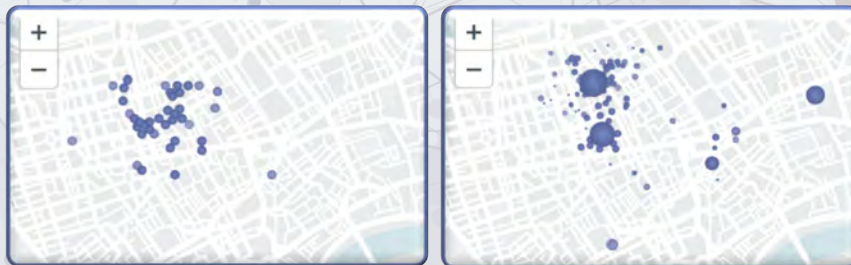


EXPAND THE DATA



IDENTIFY HOME & WORK LOCATIONS

Determine home and work locations



SCALE TRAVEL PATTERNS

Normalize the Trips



Benchmark/Calibrate Mobility Metrics



Sample Size

Market Penetration
Valid Devices
Valid Days
Tract-level Stats



Activity Patterns

Duration
Start & End times
Activity Type



Trip Statistics

Trip Rates
ToD Distribution
Trip Length
Travel Purpose

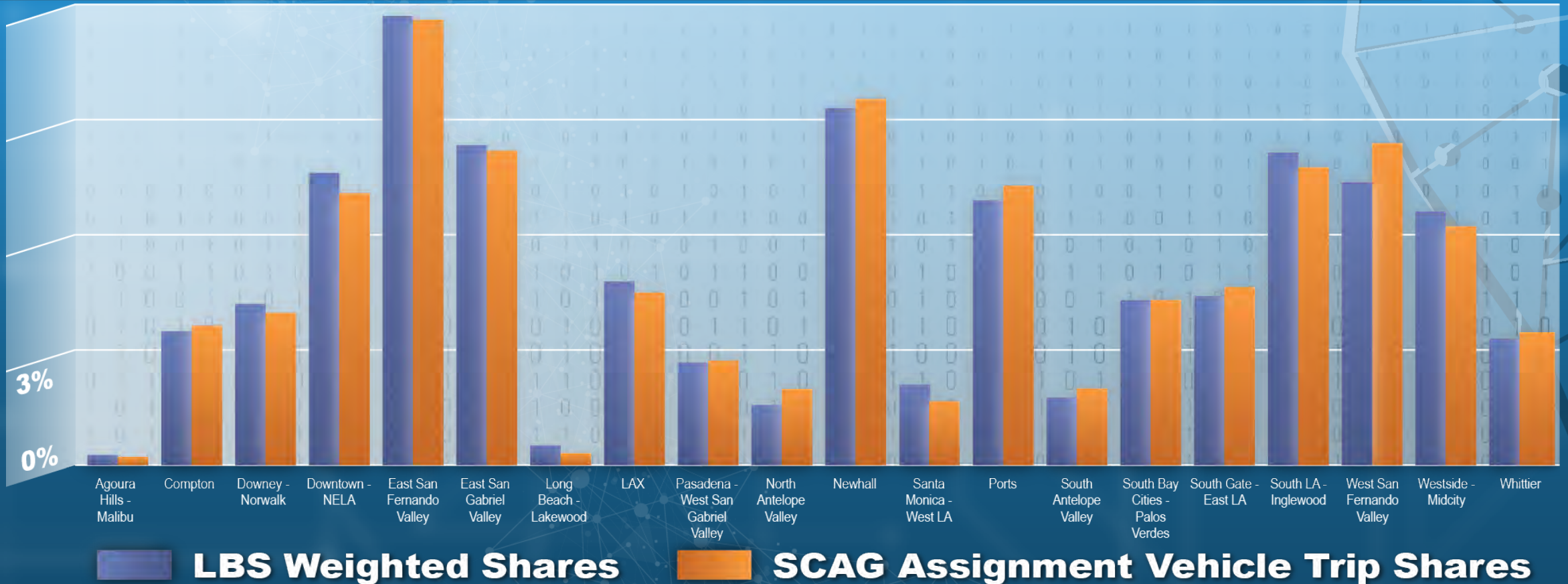


Travel Flows

Screen lines
Traffic Counts
VMT
RMSE

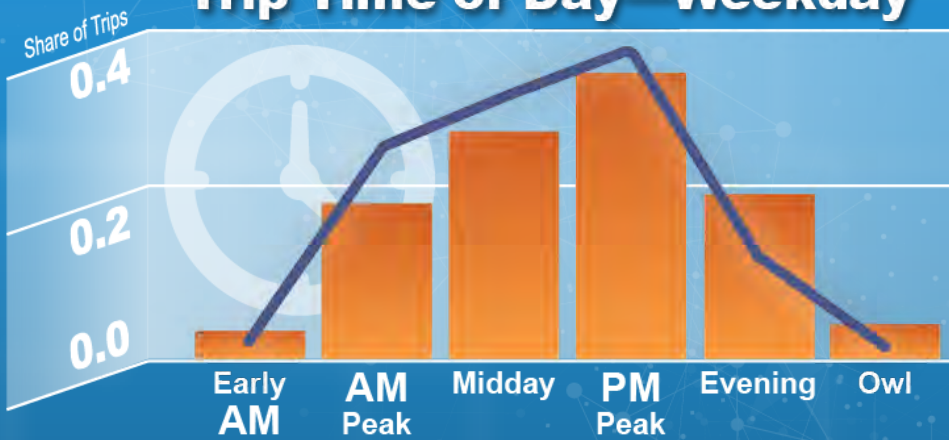
Trip Origins Compared to Regional Model

Trips by Origin District – LBS shares vs. SCAG model (for LA County)

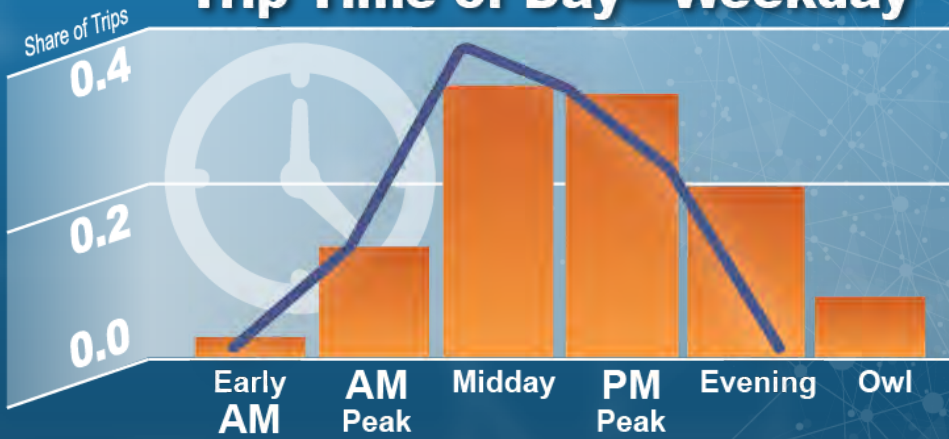


Temporal Segmentation vs. NHTS

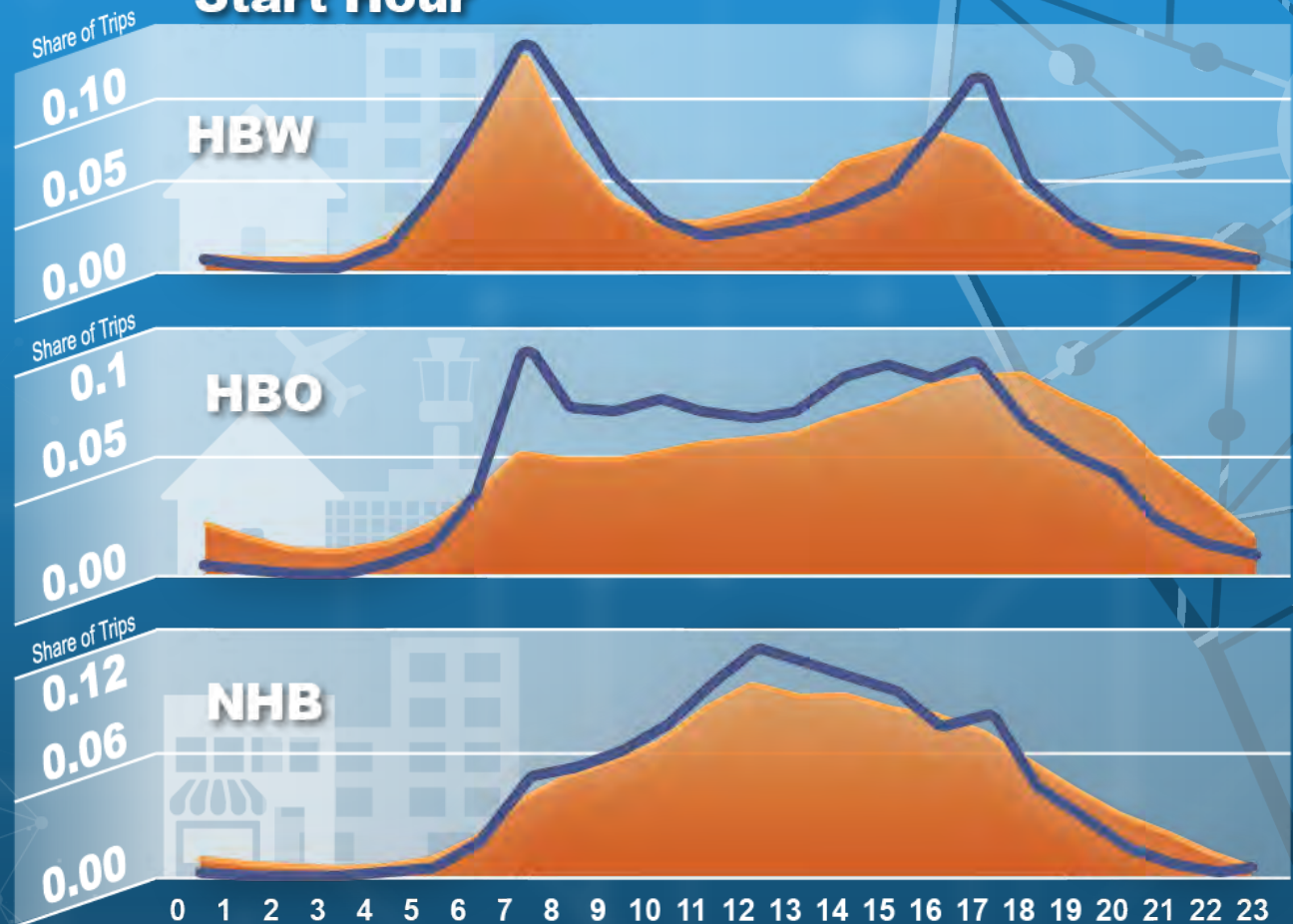
Trip Time of Day—Weekday



Trip Time of Day—Weekday



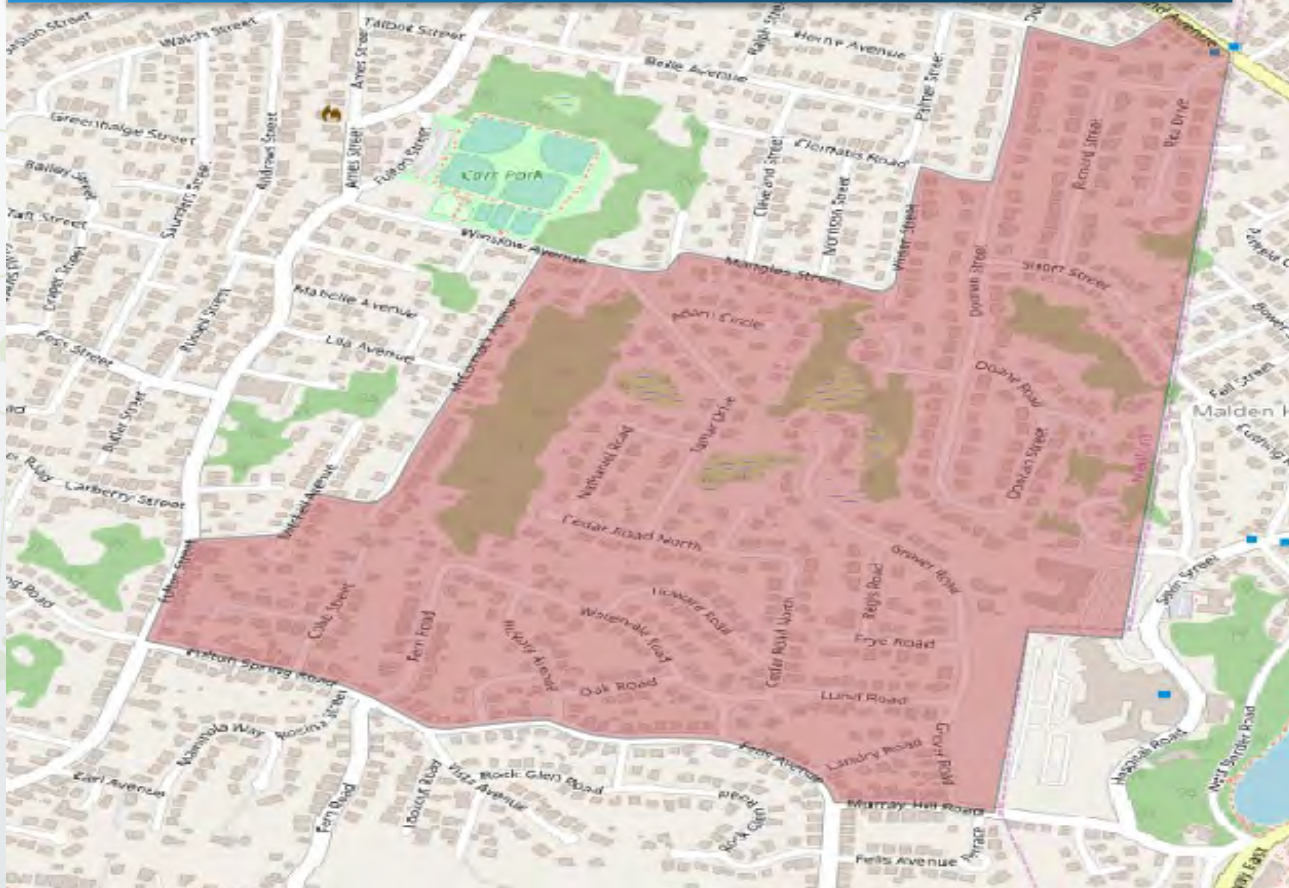
Start Hour



LBS Data **NHTS Data**

Activity Duration Analysis – Variation by Land Use Features

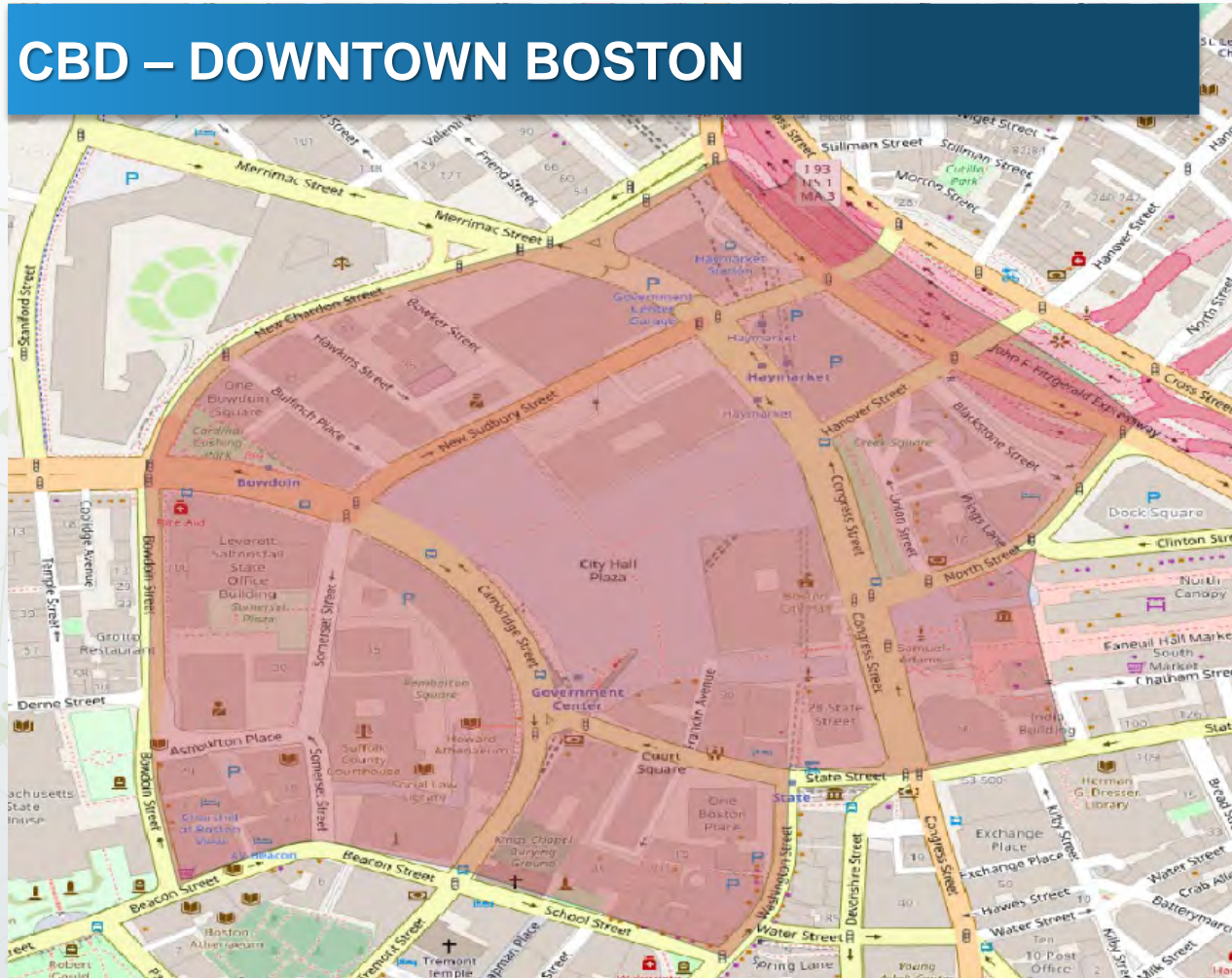
RESIDENTIAL NEIGHBORHOOD



Duration Bin mins	Activity Type		
	Home	Work/School	Other
[0, 5)	0%	0%	2%
[5, 10)	0%	0%	1%
[10, 15)	1%	0%	1%
[15, 20)	1%	0%	1%
[20, 30)	1%	0%	2%
[30, 40)	1%	0%	1%
[40, 50)	1%	0%	1%
[50, 60)	1%	0%	1%
[60, 90)	3%	0%	2%
[90, 120)	2%	0%	1%
[120, 150)	2%	0%	1%
[150, 180)	2%	0%	1%
[180, 210)	1%	0%	1%
[210, 240)	1%	0%	0%
[240, 300)	2%	0%	1%
[300, 360)	2%	0%	1%
[360, 420)	1%	0%	1%
[420, 480)	2%	0%	1%
8 hours or more	52%	2%	2%
Grand Total	76%	4%	21%

Activity Duration Analysis – Variation by Land Use Features

CBD – DOWNTOWN BOSTON



Duration Bin mins	Activity Type		
	Home	Work/School	Other
[0, 5)	0%	0%	2%
[5, 10)	0%	0%	1%
[10, 15)	0%	0%	5%
[15, 20)	0%	0%	3%
[20, 30)	0%	0%	4%
[30, 40)	0%	0%	3%
[40, 50)	0%	0%	2%
[50, 60)	0%	0%	2%
[60, 90)	0%	1%	5%
[90, 120)	0%	1%	3%
[120, 150)	0%	1%	2%
[150, 180)	0%	1%	2%
[180, 210)	0%	1%	1%
[210, 240)	0%	2%	1%
[240, 300)	0%	3%	2%
[300, 360)	0%	3%	1%
[360, 420)	0%	4%	1%
[420, 480)	0%	9%	1%
8 hours or more	3%	23%	3%
Grand Total	4%	51%	44%



LA Metro NextGen Bus Plan



Transit Competitiveness Analysis

STEP 1

Use farecard data to explain **TRANSIT** travel market



STEP 3

Compare **TRANSIT** travel to **TOTAL** travel in each market



STEP 2

Use cell phone data to explain **TOTAL** travel market






STEP 4

Use trip planners to **compare transit and driving** travel times



Transit Riders Frequency of Travel

USING 4 MONTHS OF FARECARD DATA

	USAGE FREQUENCY	# OF FARECARDS	COUNT OF BOARDINGS
 Frequent	>150 Transactions	152,532 5%	43,680,893 52%
	50 – 150 Transactions	248,851 9%	22,027,882 26%
 Occasional	10 – 50 Transactions	552,374 20%	12,585,194 15%
 Infrequent	<10 Transactions	1,905,501 66%	5,614,072 7%
	TOTAL	2,859,258 100%	83,908,041 100%

Source: TAP data - Metro and Municipal Operators (July through October, 2017)

Competitiveness of Relative Travel Times

Transit Market Share

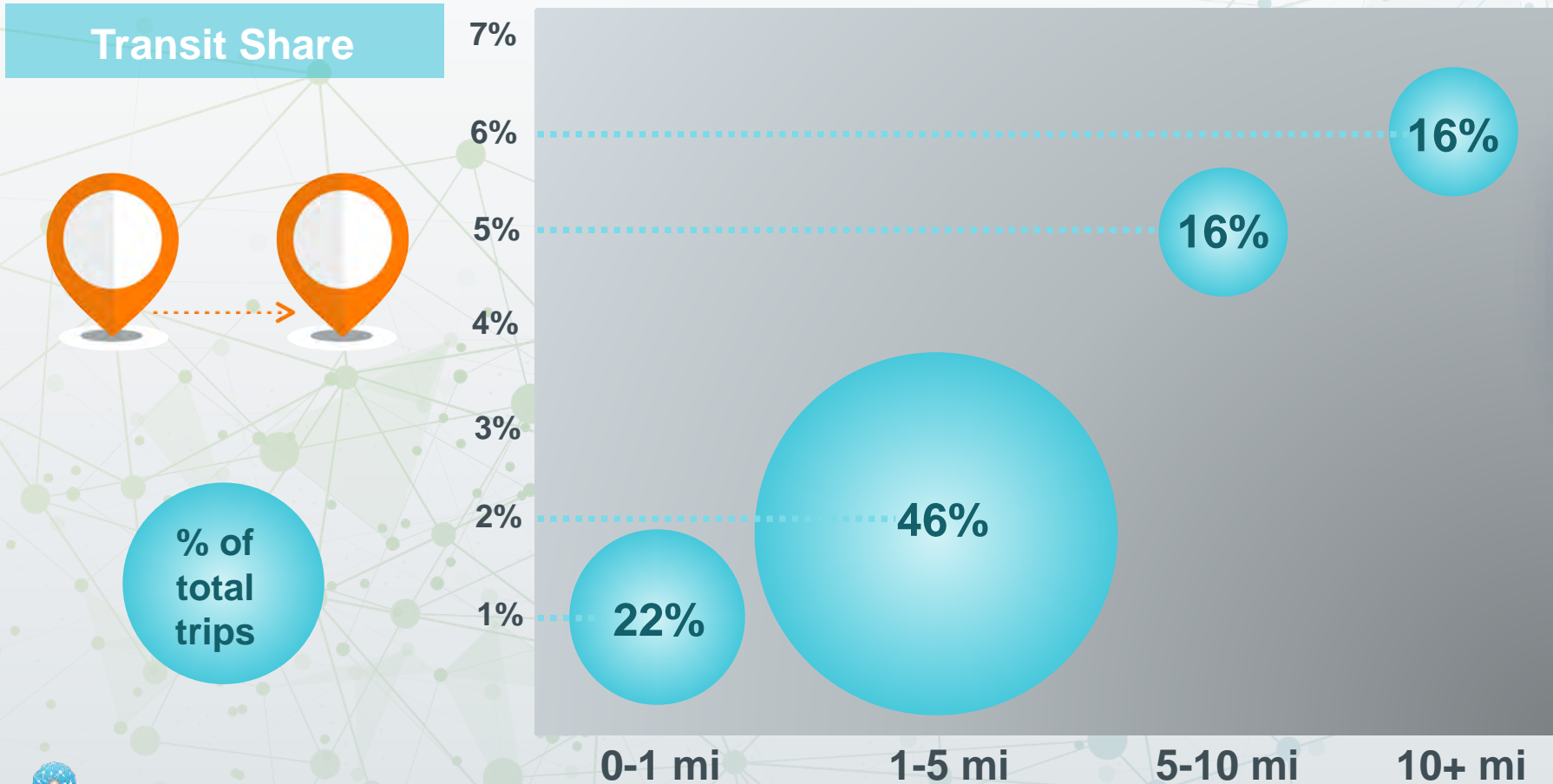


When driving is over twice as fast, transit is less competitive

Transit to Drive Time Ratio

Long-Distance Commute Trips and Transit

TRANSIT SHARE BY DISTANCE & PERCENT OF TOTAL TRIPS



Short trips represent the largest potential market

Redesign in Action

Sylmar/San Fernando Key Facts

Trips: 380,000 trips

Market Share: 1.4% market share

Mileage: 76% of trips under 2.5 miles

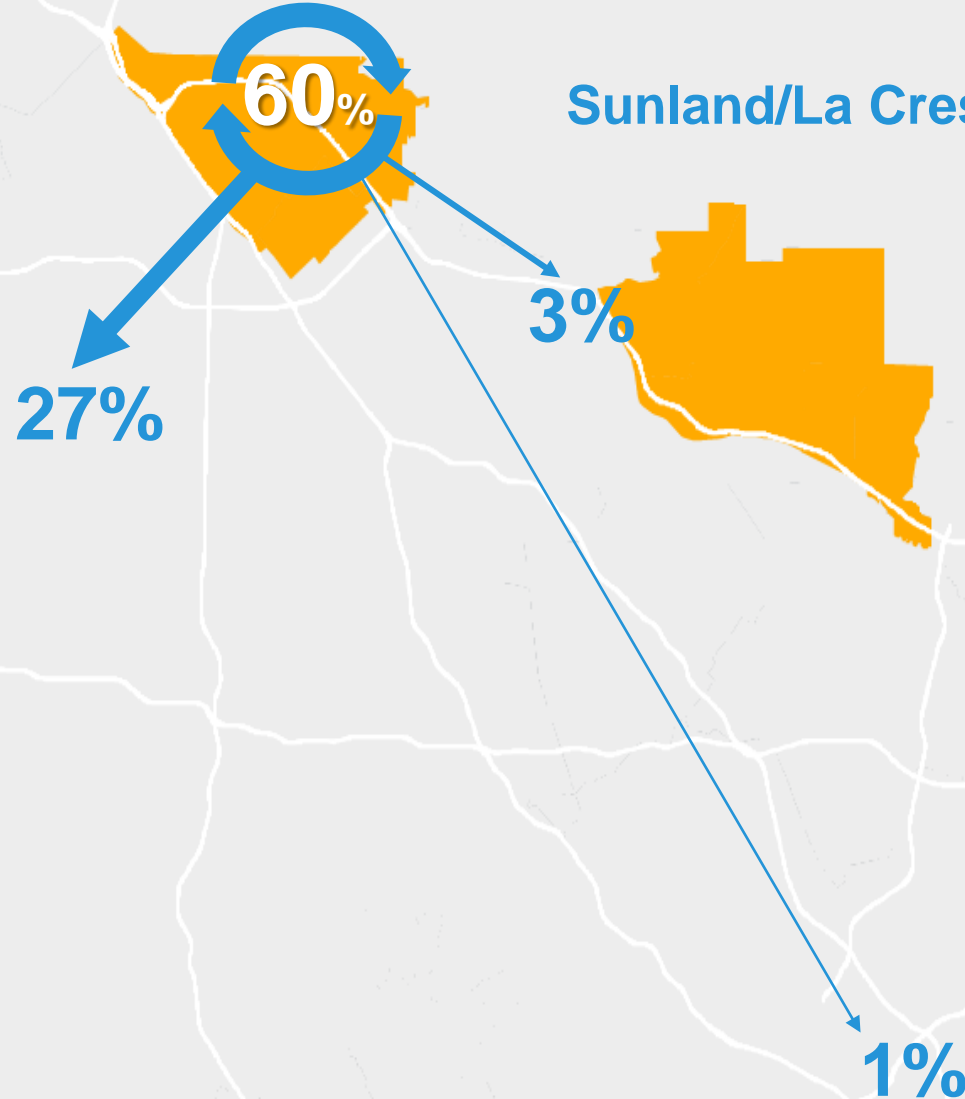
Travel time

competitiveness: 3.25-3.50

60% of trips occur within the area
27% of trips are to the Valley
1% of trips are to Downtown LA
3% of trips are to Sunland/La Crescenta

Sylmar/San Fernando

Sunland/La Crescenta



Percentages do not equal 100%. Additional trips dispersed throughout the County.

Redesign in Action

Sunland/La Crescenta Key Facts

Trips: 325,000 trips

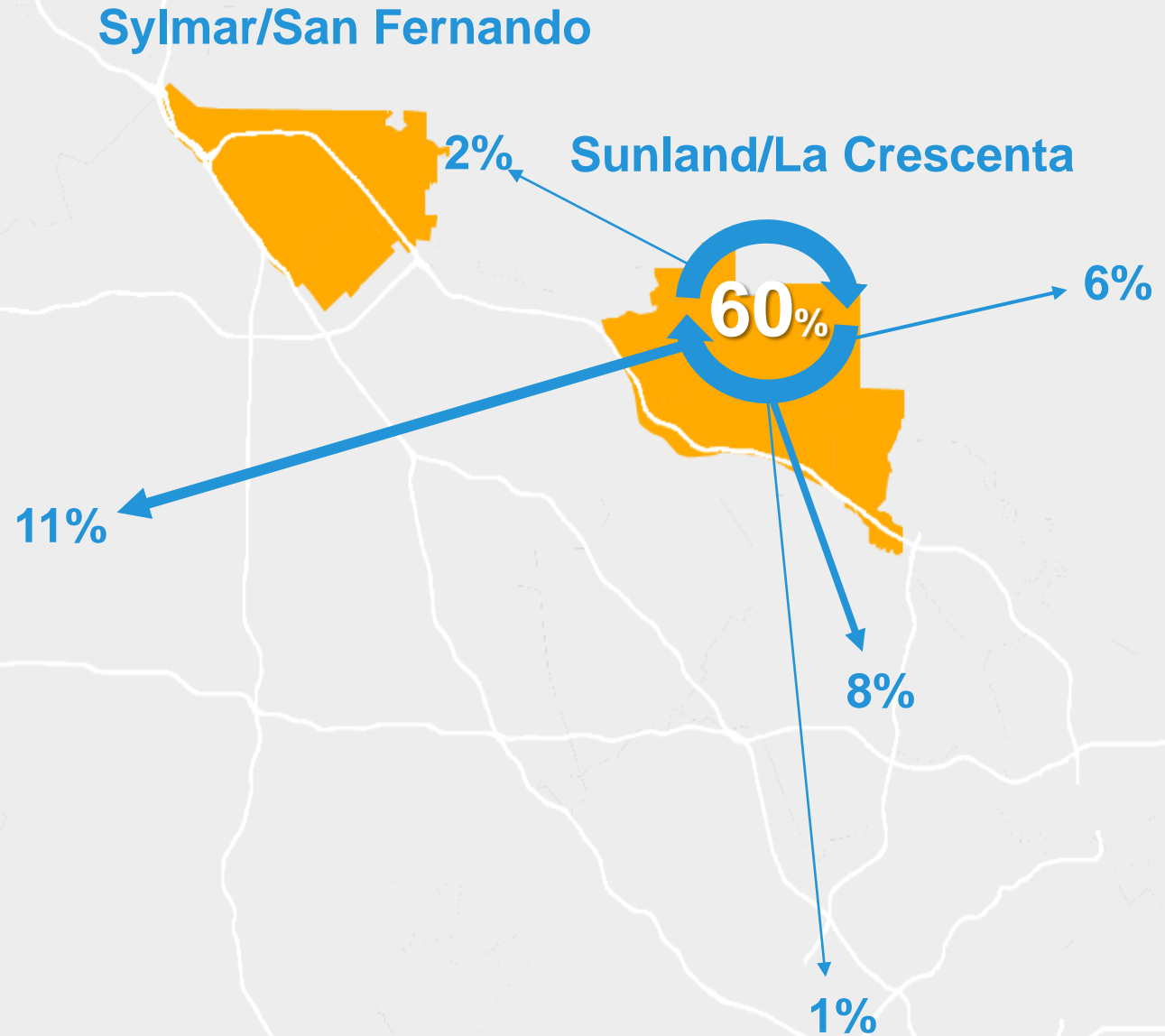
Market Share: 0.7% market share

Mileage: 44% of trips under 2.5 miles

Travel time

competitiveness: 3.00-3.25

60% of trips occur within the area
2% of trips are to Sylmar/San Fernando
11% of trips are to the Valley
1% of trips are to Downtown LA
8% of trips are to Glendale
6% of trips are to Pasadena



Percentages do not equal 100%. Additional trips dispersed throughout the County.

Redesign in Action

Existing Transit

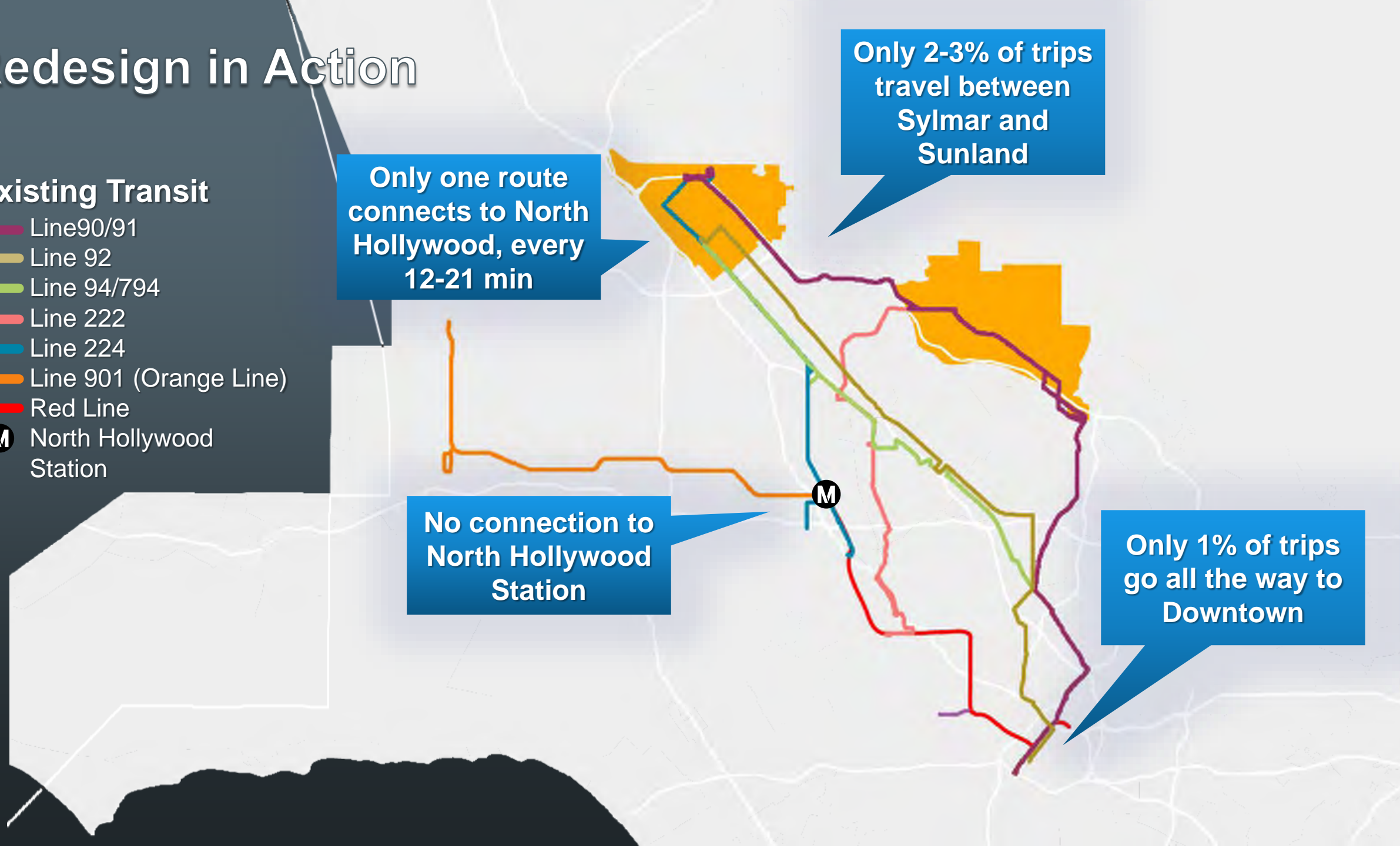
- Line 90/91
- Line 92
- Line 94/794
- Line 222
- Line 224
- Line 901 (Orange Line)
- Red Line
- M** North Hollywood Station

Only one route connects to North Hollywood, every 12-21 min

Only 2-3% of trips travel between Sylmar and Sunland

No connection to North Hollywood Station

Only 1% of trips go all the way to Downtown



Redesign in Action

Modified Service

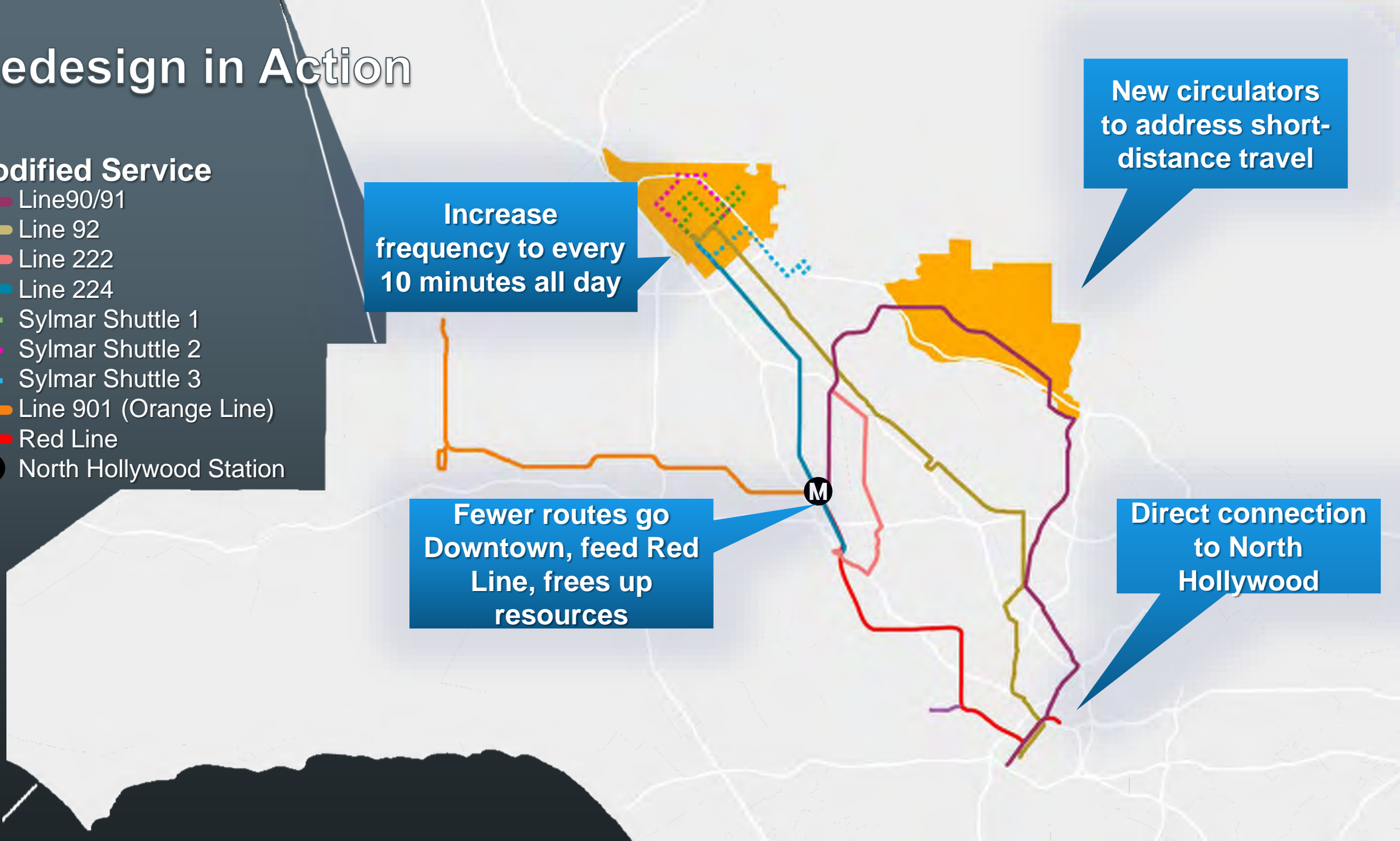
- Line 90/91
- Line 92
- Line 222
- Line 224
- Sylmar Shuttle 1
- Sylmar Shuttle 2
- Sylmar Shuttle 3
- Line 901 (Orange Line)
- Red Line
- M** North Hollywood Station

Increase frequency to every 10 minutes all day

New circulators to address short-distance travel

Fewer routes go Downtown, feed Red Line, frees up resources

Direct connection to North Hollywood





COVID-19 Recovery Planning for Transit



To Help with COVID-19 Recovery for Transit

1. What is the impact of work-from-home?
2. Have travel patterns stabilized?
3. Are automobile sales going up?
4. Are equity-focused communities impacted disproportionately?
5. What role does social distancing play?

Our Data Approach

To help agencies assess the changes in travel behavior as a result of COVID-19, we have developed a suite of near real-time LOCUS products powered by Location Based Services (LBS) data.

Traffic Footfall Tracker

Measures store visitation trends by location, day of week, brand, and category

Travel Tracker

Measures resident travel patterns – travel sheds, time-of-day of travel, VMT – to study changes in behavior.

1. Impacts on City Centers/Downtowns – New York

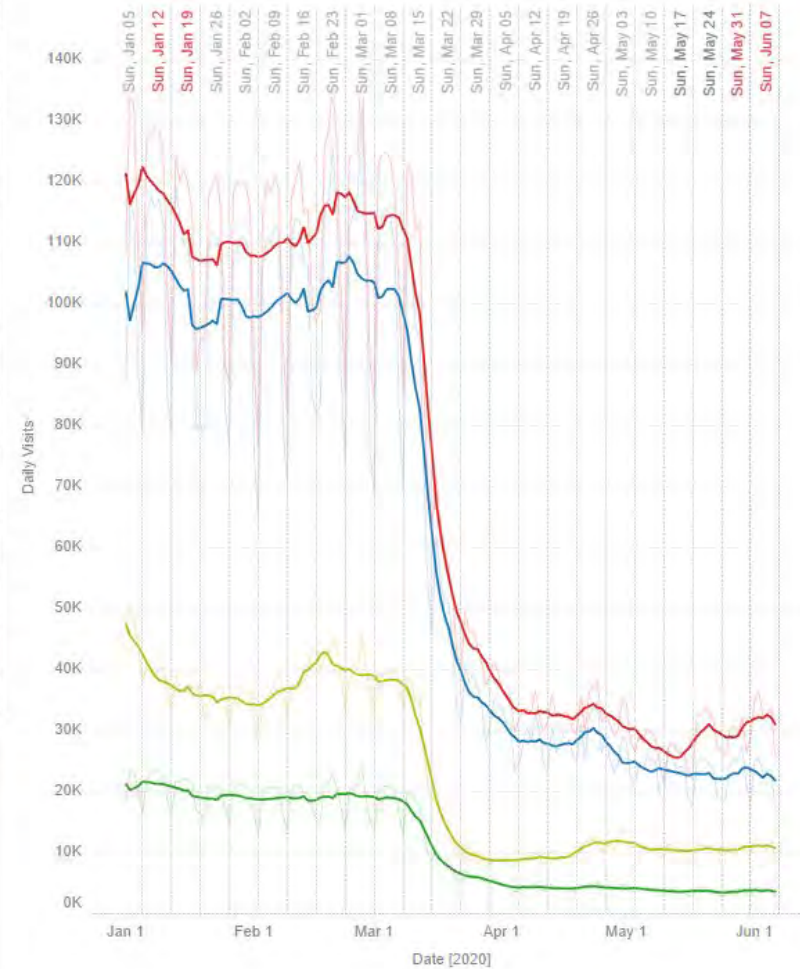
Traffic Footfall Tracker

Markets **4** Submarkets **31** Brands **196** Outlets **1,458**



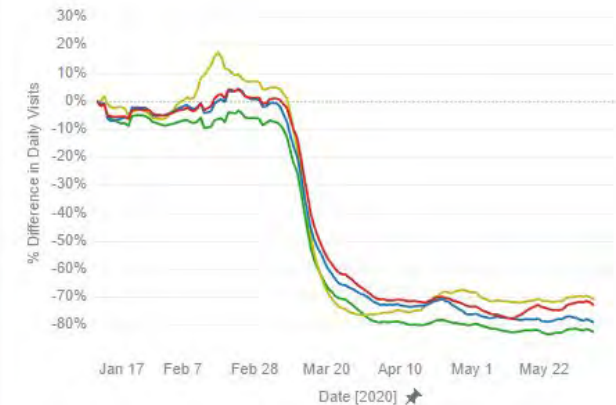
Market Type	Brands	Locations
Retail	83 brands	631 locations
Dining	43 brands	559 locations
Travel	60 brands	187 locations
Entertainment	10 brands	95 locations

7-day Moving Average of Daily Visits by Market Type



Market: (Multiple values)
 Submarket: (All)
 Brand: (All)
 State: NY
 County: (All)
 Valid Days Threshold: (Multiple values)

Percent Difference in 7-day Moving Average From Jan 14, 2020



[Go to Disaggregate Summaries](#)

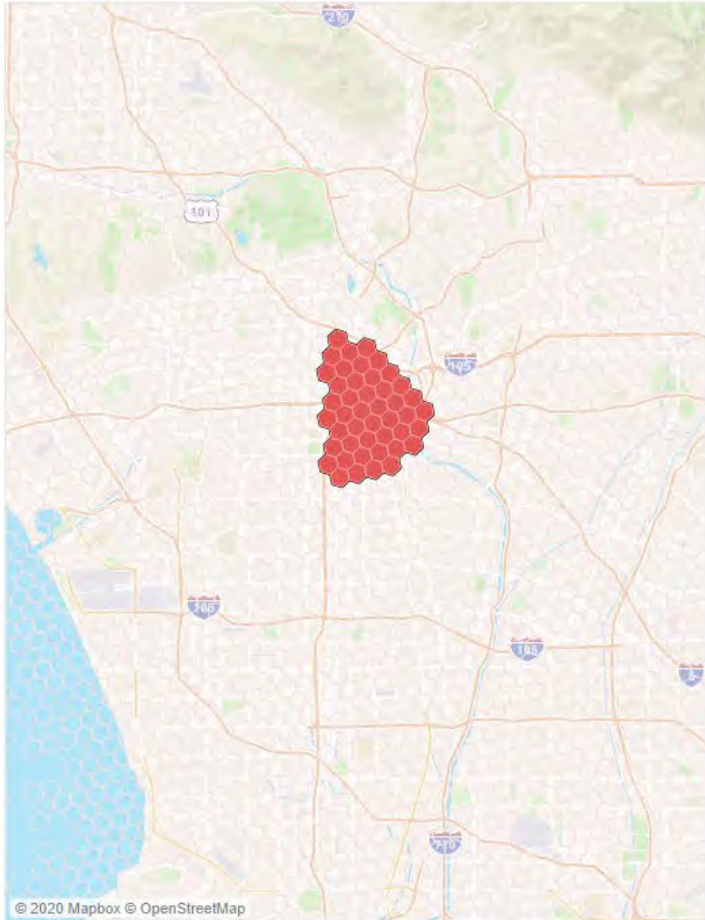


1. Impacts on City Centers/Downtowns – Los Angeles

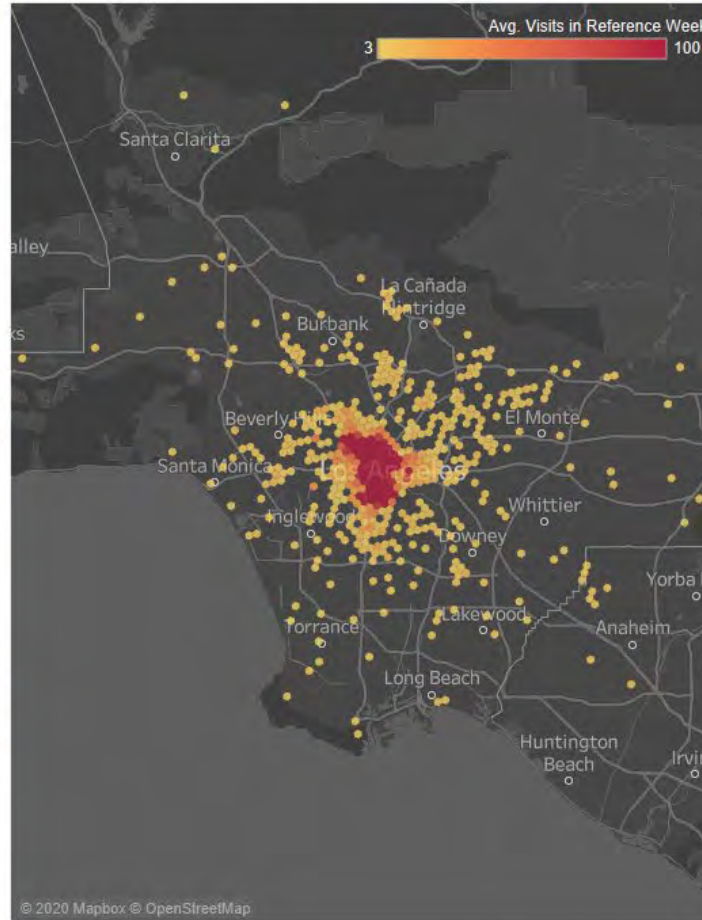
Los Angeles, CA | Travel Tracker

Avg. Visits in Reference Week 48,924
Avg. Visits in Recent Week 22,172

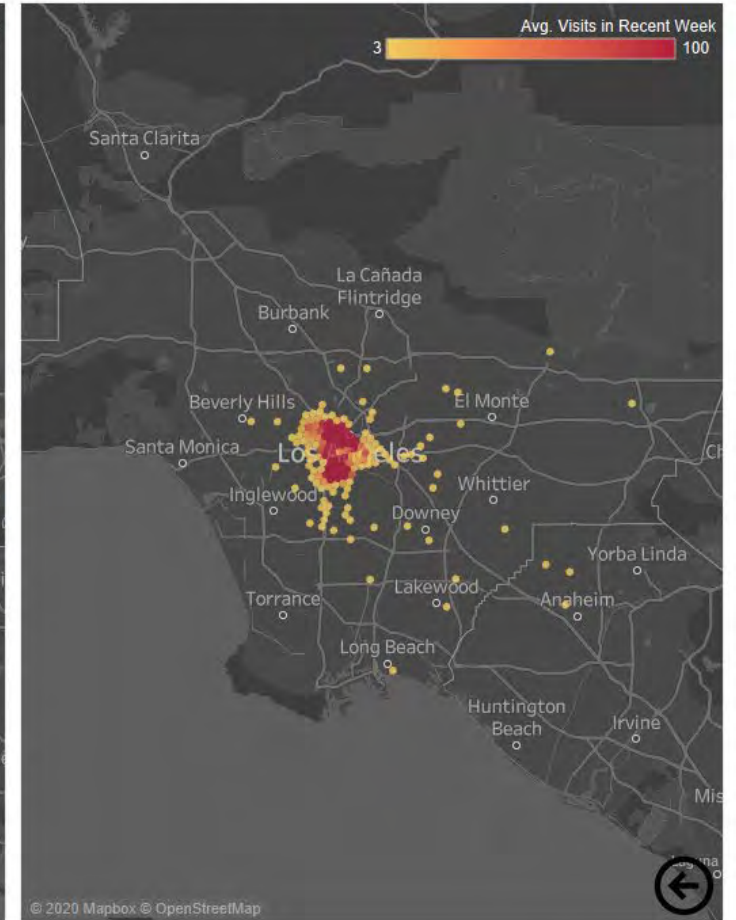
Visited Hexes



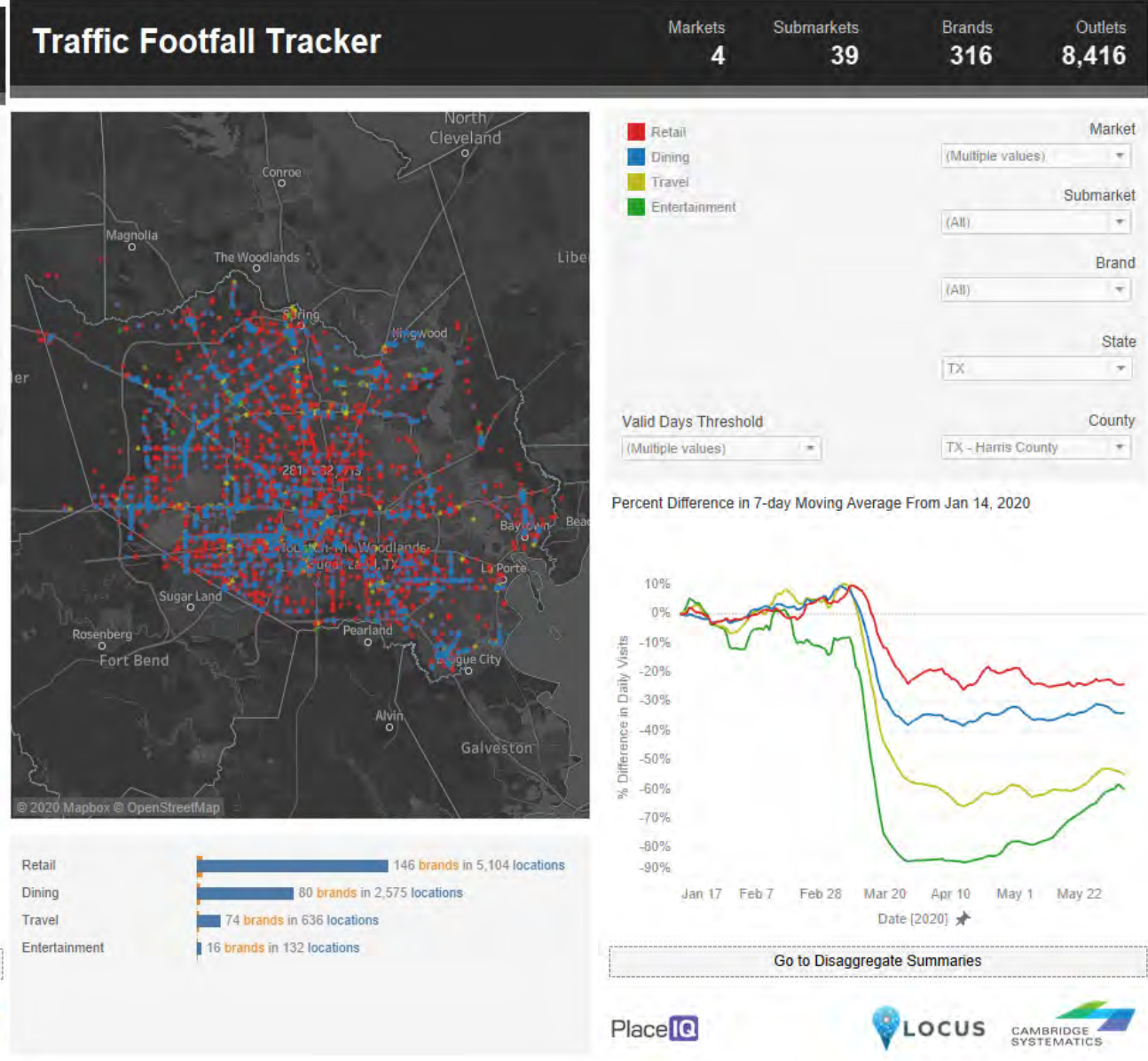
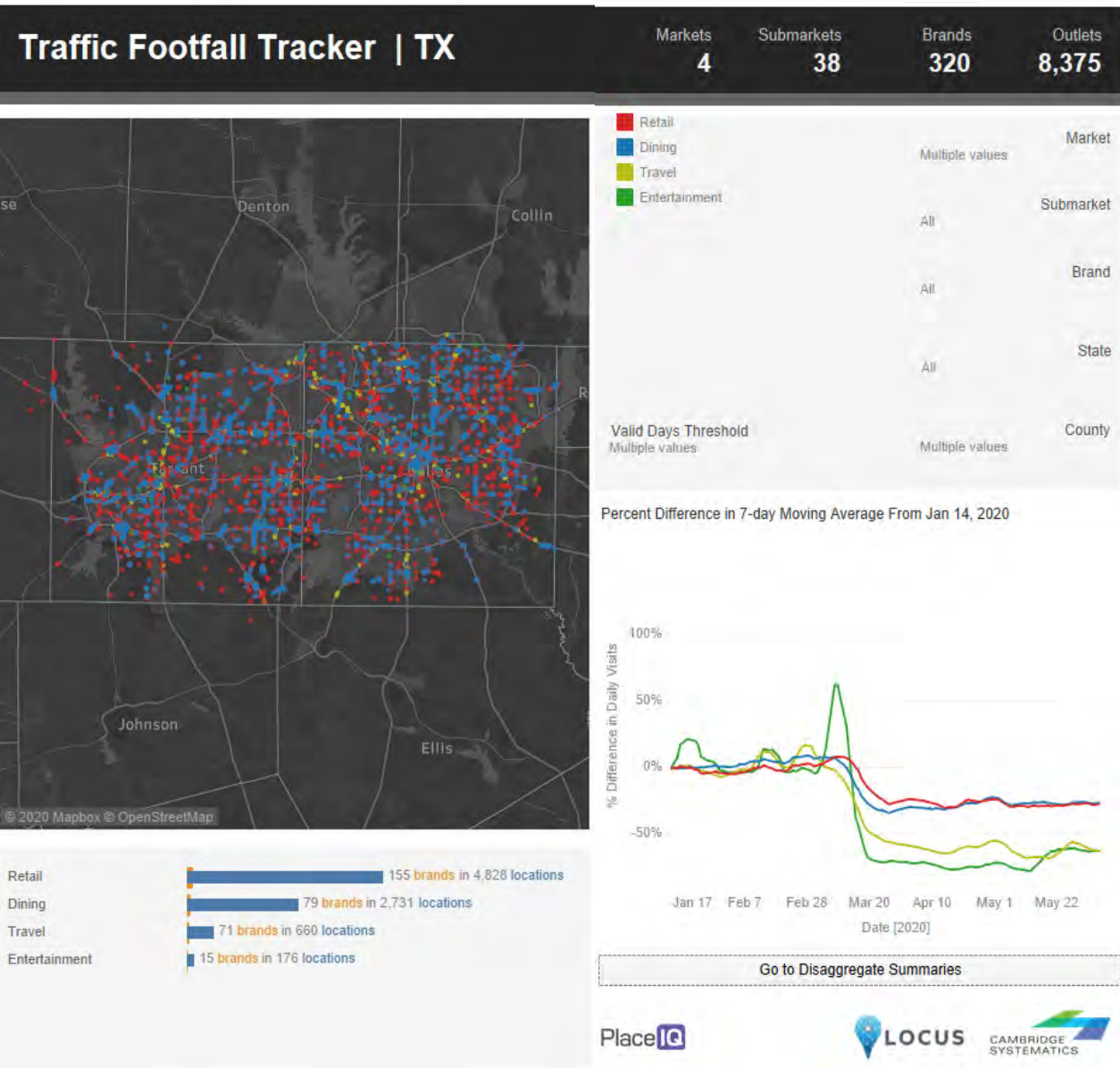
Home Locations in Reference Week



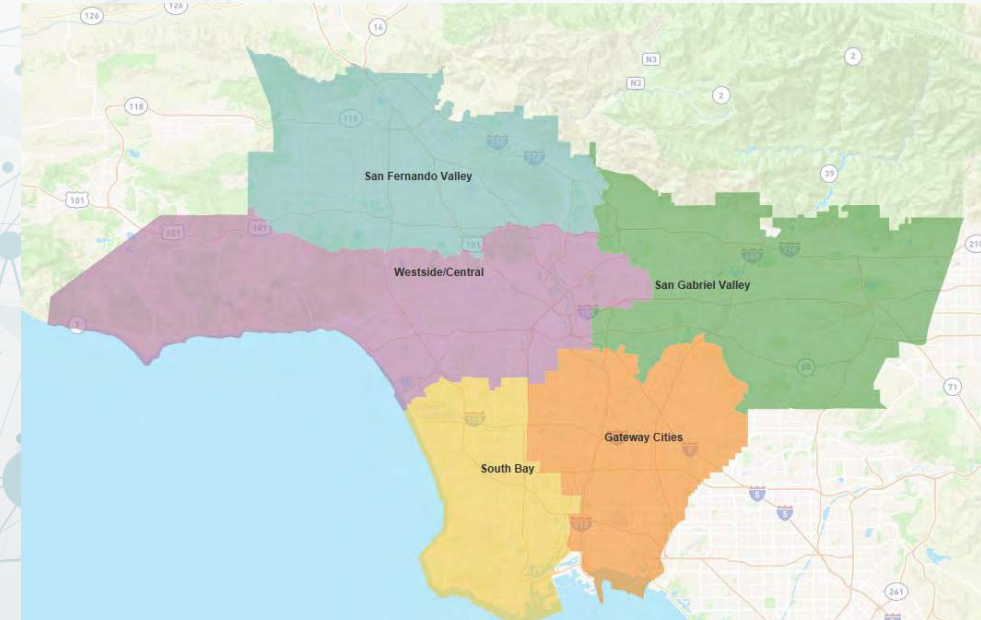
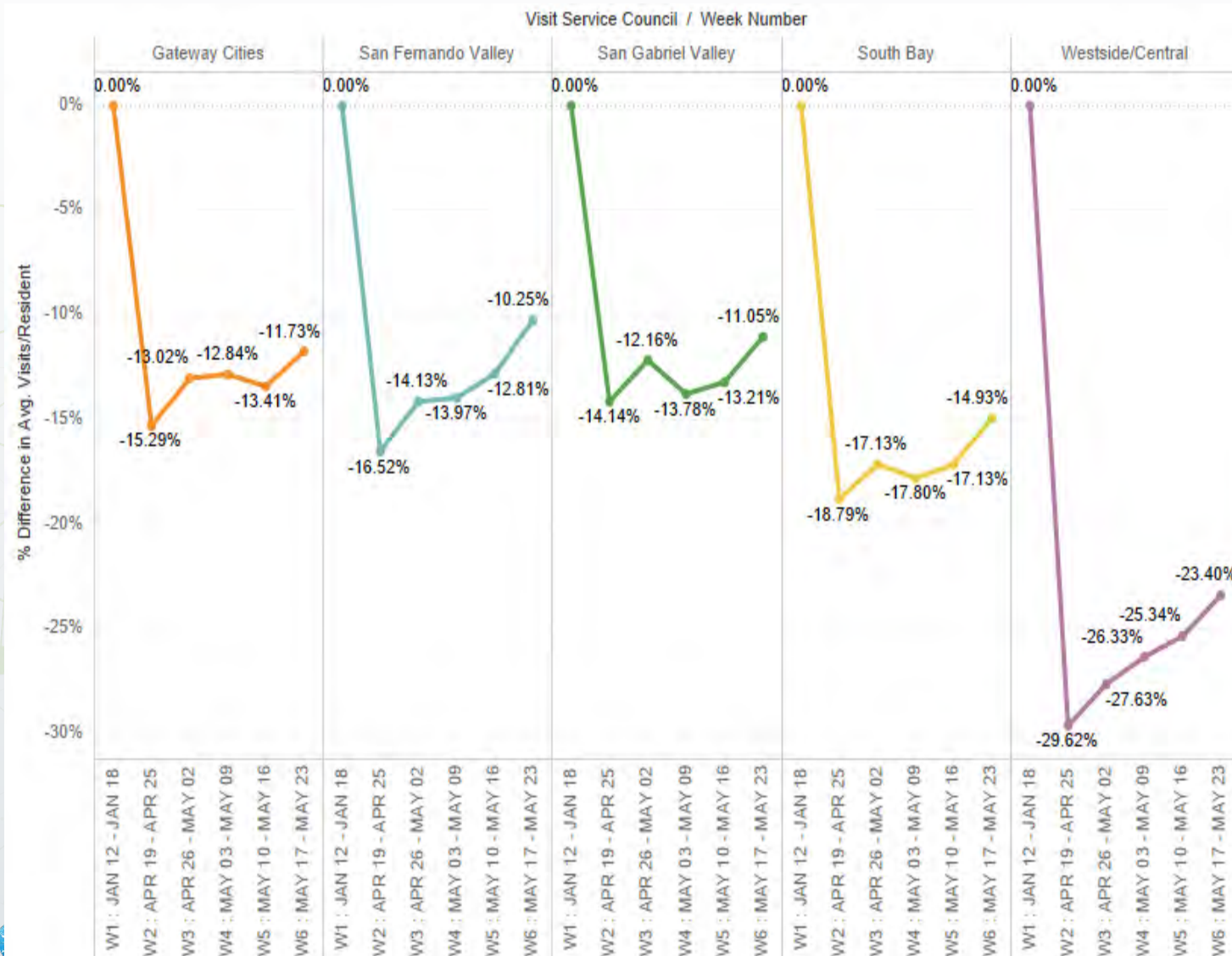
Home Locations in Recent Week



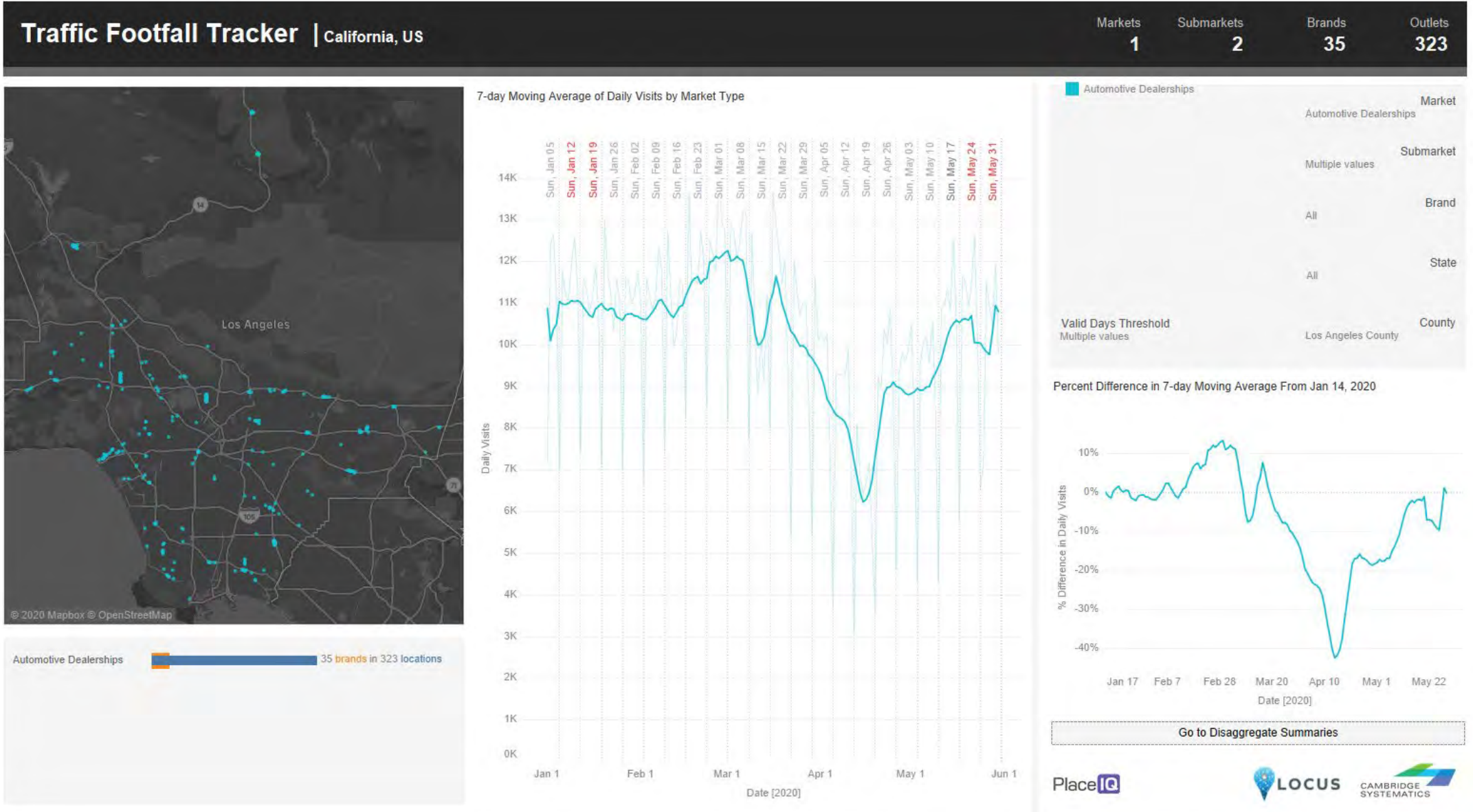
2. Travel Pattern Stabilization - Dallas and Houston



2. Travel Pattern Stabilization - Los Angeles



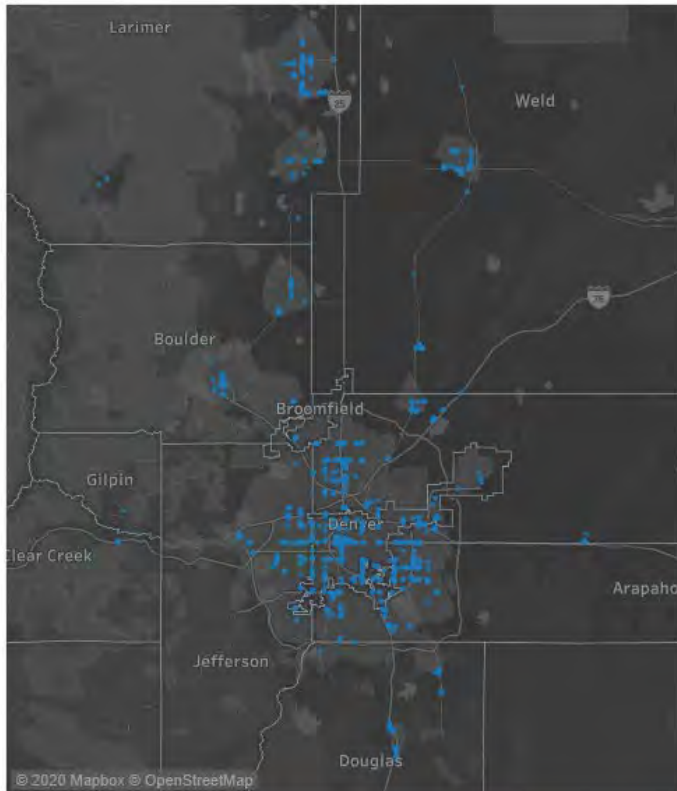
3. Automobile Dealership Traffic Trending up in LA County



4. Dining Trends in Equity Communities in Denver

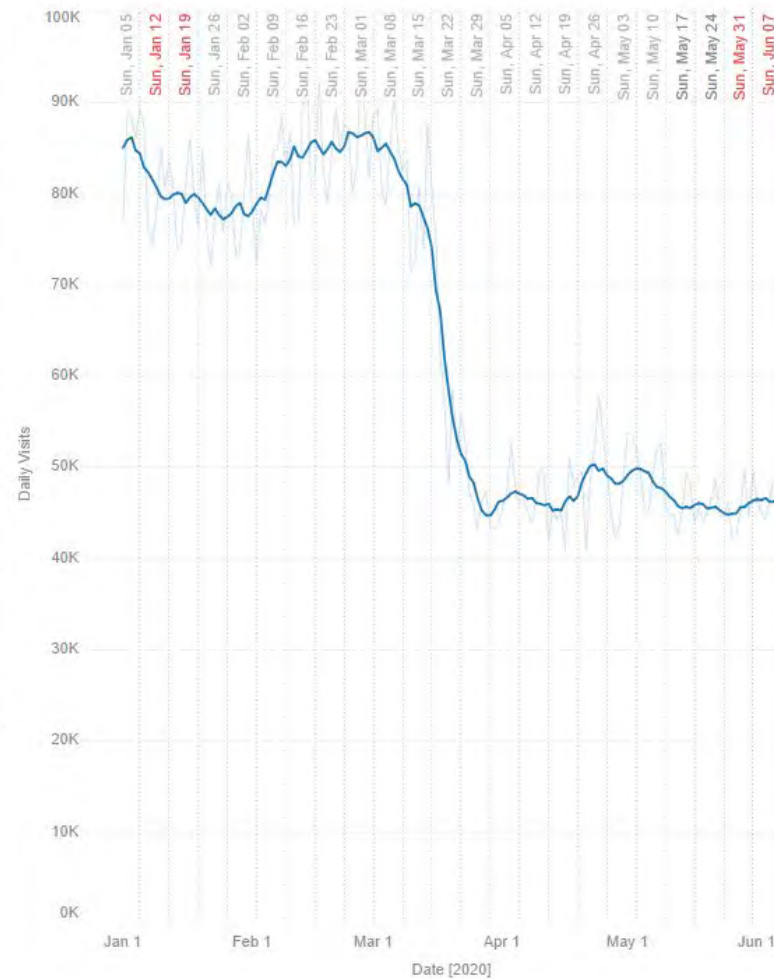
Traffic Footfall Tracker | Colorado

Markets **1** Submarkets **5** Brands **78** Outlets **976**



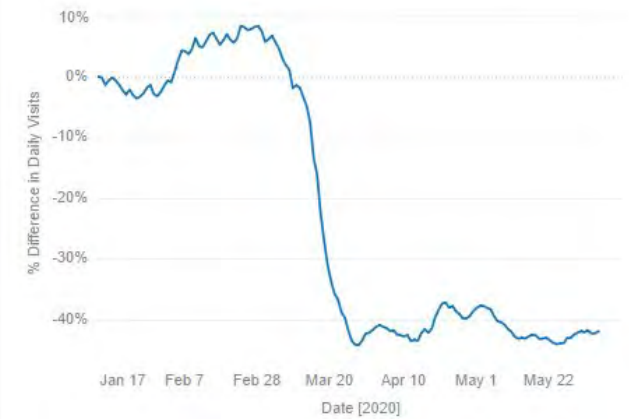
Dining ■ 78 brands in 976 locations

7-day Moving Average of Daily Visits by Market Type



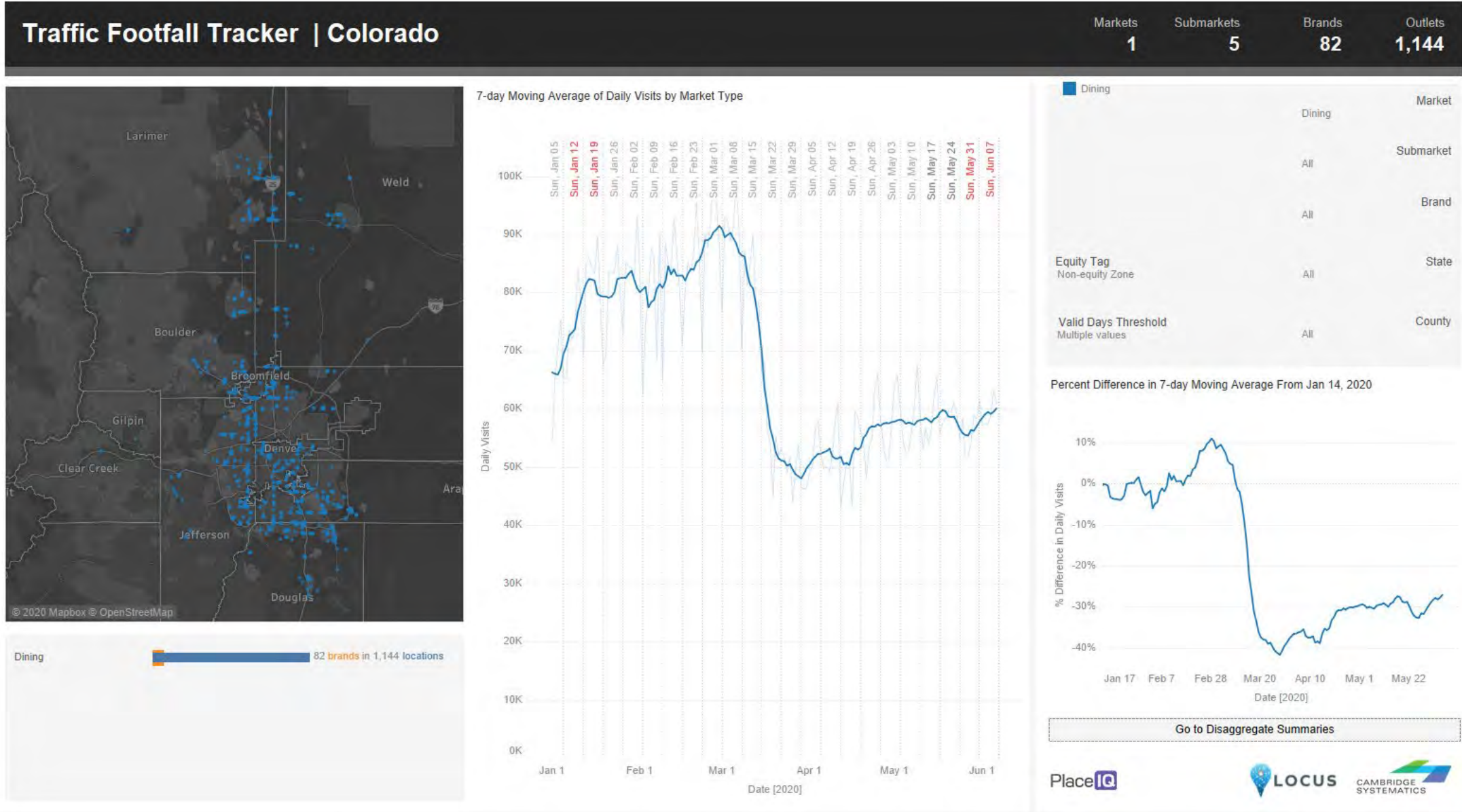
Equity Tag	Equity Zone	Valid Days Threshold
Dining	All	Multiple values

Percent Difference in 7-day Moving Average From Jan 14, 2020



[Go to Disaggregate Summaries](#)

4. Dining Trends in Other Communities in Denver



Forecasting Travel Behavior in Uncertain Times

We are developing and implementing an integrated method of forecasting and planning that is driven by near real-time LOCUS data. Our approaches are refined based on what we learn from these data. Given the uncertainty, we recommend supplementing traditional modeling with best practice risk analysis tools



Scenario Planning: Define “what if” scenarios that you want to test



LBS Data: Develop a clear picture of what is happening in real time to study impacts



Modeling Tools: Assess scenarios by ingesting LBS data to calibrate models



Risk Analysis: Bound the future given the uncertainty of key input variables

Anurag Komanduri

akomanduri@camsys.com

312-659-2000