

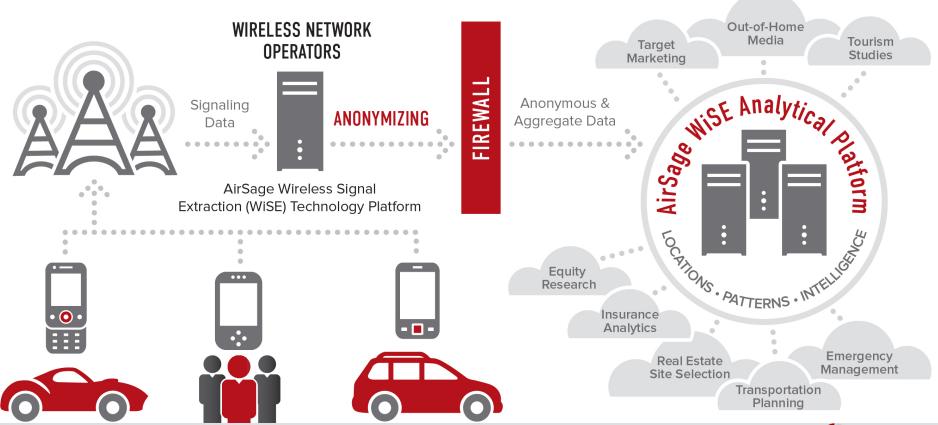
Innovations in Data Collection

The Power of Where and When

Company Overview



AirSage WiSE Platform





Practical Solutions

at my store and my competitor's

store over time to gauge changes

in equity.

I'm glad I now can get accurate counts on how many people drive I wish I knew how many people Out-of-Home by our new billboard on a typical were visiting our attractions from Media **Tourism** Target weekday between 4-7pm other counties and states Studies Marketing Vical Platform I need to better understand us & Our current methods of patterns between three potential Data gathering data describing who sites for my new restaurant to is using a roadway are make better location decisions. expensive and time consuming. How can I get more for the "I need accurate dynamic OD matrices to better simulate money?" the alternatives and ensure we are making the best "If I only knew the trip G ATTERNS . INTELLIGITA investment." generation rates and how far people were traveling to Equity similar locations. I could easily predict the impacts of this new development." Analytics It would be great to see the traffic



Transportation

Planning

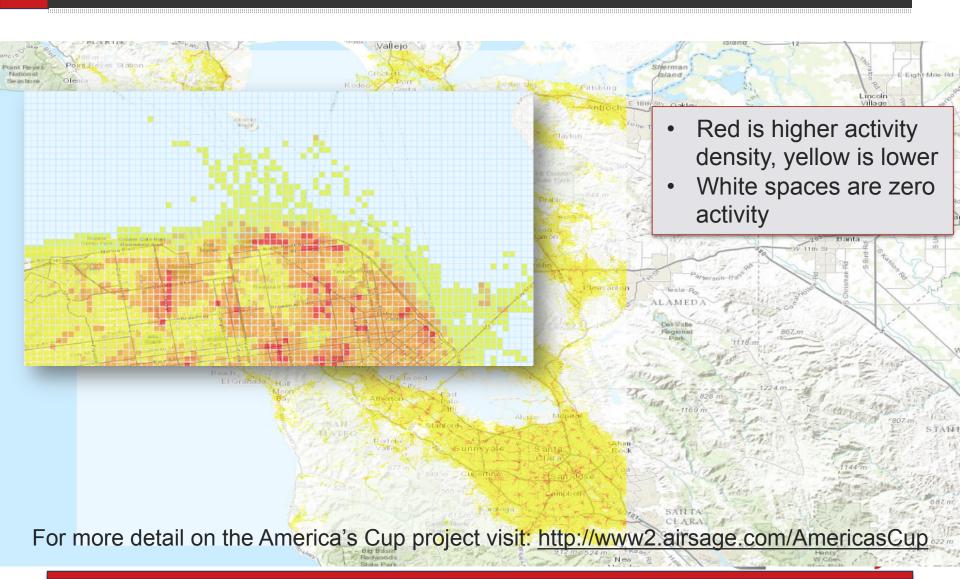
Real Estate

Site Selection

mergency

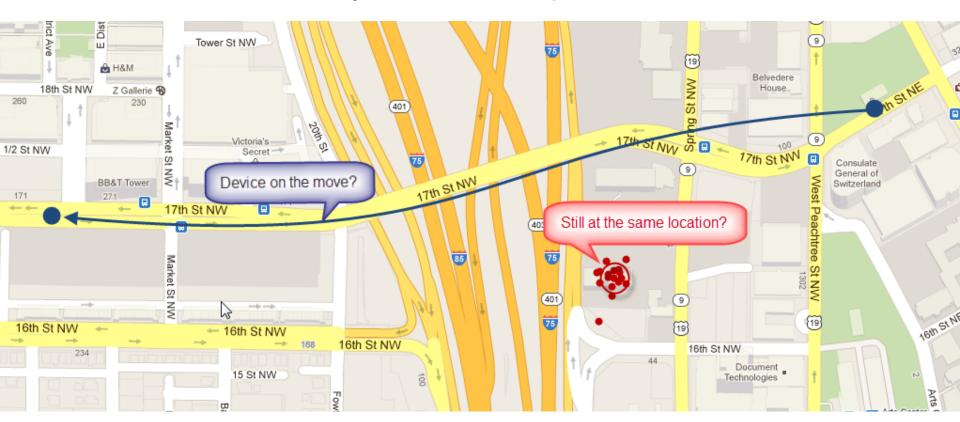
Management

What do we do?



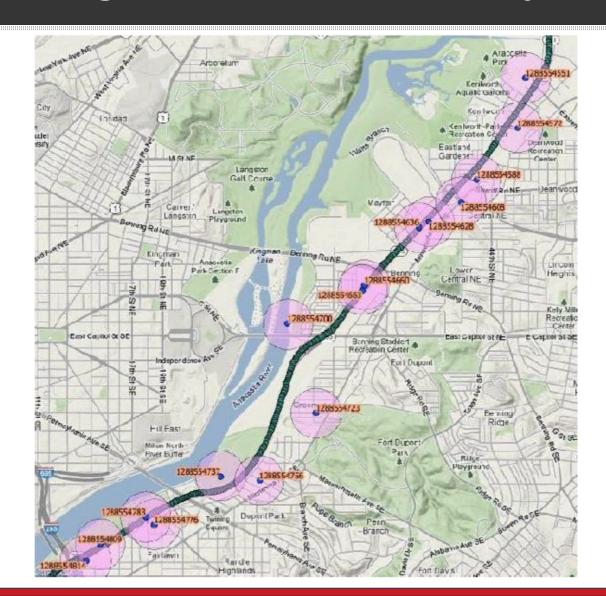
How do we do it?

Each location is analyzed and compared to other locations.



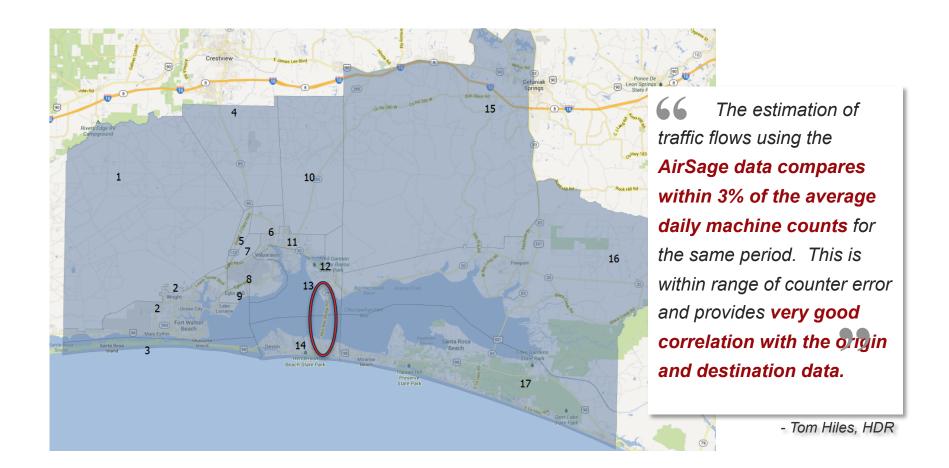


AirSage Data Accuracy





Case Study: Niceville, Florida





Niceville, FL: Trip Characteristics

Niceville, FL -- Average Weekday Travel Data (Jan 15 - Feb 15, 2012)

Census 2010 Population: 130,575

Average Daily Sampled Subscribers: 18,444
Average # of Trips per Subscriber per Day: 4.00

Proportion of Trips by Purpose (H-Home, W-Work, O-Other)

| Trip Type | HW | WH | НО | 00 | OH | WO | OW | WW | НН |
|-----------|------|------|-------|-------|-------|------|------|------|------|
| Percent | 3.9% | 2.9% | 21.2% | 32.7% | 20.4% | 6.2% | 5.2% | 2.4% | 5.2% |

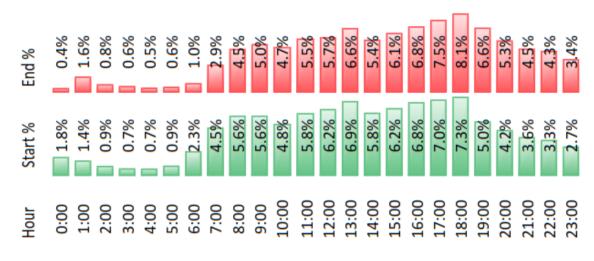
Trip Length Distribution (miles)

Less than 1 32.8% 17.1% 2 3 13.8% 8.3% 4 5 5.6% 6 3.4% 3.8% 7 8 3.0% 9 2.5% 10 2.3% 11 1.9% 12 1.3% 13 1.0%

3.2%

Greater than 13

Time of Day Distribution (Hourly)





What can the data tell us?

Activity patterns can be examined over time to determine:

- Home-Based Work (HBW)
- Home-Based Other (HBO)
- Non-Home Based (NHB)
- Combinations of Home, Work, Other
 - HO, HW, HH, WH...
- Trips per Day by Type
- Trip Lengths by Type
- Time of Day Information
- Residence Class

"Generates detailed traces that can be used to construct path histories with high fidelity across long periods of time"*

SENSEable City Laboratory



^{*} Research study by:

⁻ IBM Dublin Research Laboratory

Massachusetts Institute of Technology

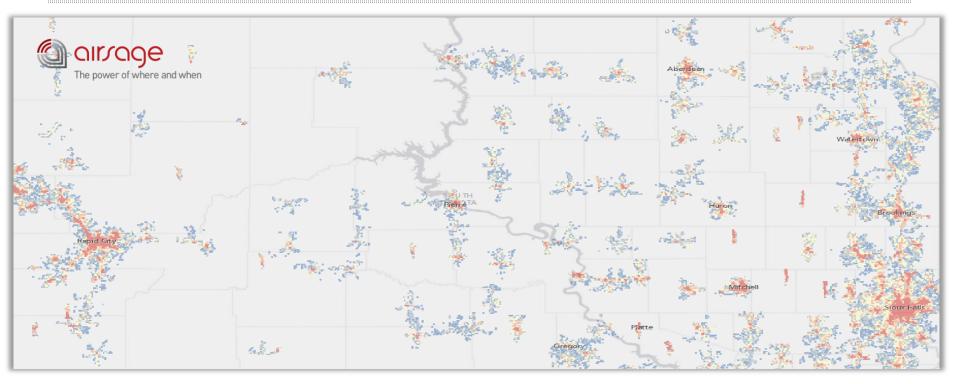
Resident Classes

| 2-Category Option | The mobile device | | | | |
|-------------------|---|--|--|--|--|
| Resident | Lives or works in the study area | | | | |
| Visitor | Does not live or work in the study area | | | | |

| 6-Category Option | The mobile device |
|--------------------|---|
| Resident Worker | Lives and works are in the study area |
| Home Worker | Lives and works at the same location in the study area |
| Inbound Commuter | Works in the study area but lives outside of the study area |
| Outbound Commuter | Lives in the study area but works outside of the study area |
| Long-term Visitor | Is a non-resident present in the study area between 2 & 14 days |
| Short-term Visitor | Is a non-resident present in the study area less than 2 days |



Wide Coverage



24th October 2012 -- South Dakota 30 million sightings generated by more than 400,000 devices Synthesized Resident Population of 808,000 people (Census 2011 Est. 824,000)



Synthesizing a Full Population

Leveraging Census data to expand our sample of 100 million devices accurately simulates the full population of the U.S. 311 million residents...

Sample Bias?

- Inclusion of all mobile devices regardless of age, income
- Attach Census data to show HHI

Cell Phone and Smartphone Ownership

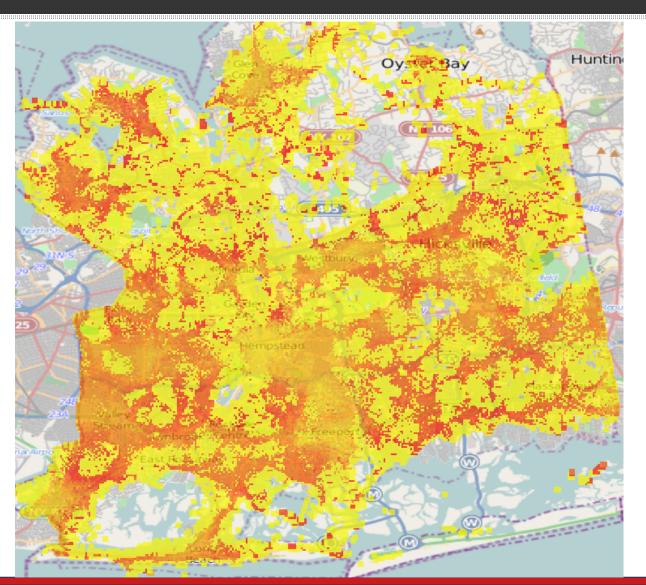
% of American adults within each group who own a cell phone and the % who own a smartphone

| | | Cell Phone | Smartphone |
|-------|-------------------------------|------------------|-------------------|
| All a | adults (n=2,252) | 91% | 56% |
| а | Men (n=1,029) | 93 ^b | 59 ^b |
| b | Women (n=1,223) | 88 | 53 |
| Rac | e/ethnicity | | |
| а | White, Non-Hispanic (n=1,571) | 90 | 53 |
| b | Black, Non-Hispanic (n=252) | 93 | 64 ^a |
| С | Hispanic (n=249) | 88 | 60 |
| Age | | | |
| а | 18-29 (n=404) | 97 ^{cd} | 80 ^{bcd} |
| b | 30-49 (n=577) | 95 ^{cd} | 67 ^{cd} |
| С | 50-64 (n=641) | 89 ^d | 45 ^d |
| d | 65+ (n=570) | 76 | 18 |

| Edu | cation attainment | | | |
|-----|--------------------------------|-------------------|--|--|
| a | No high school diploma (n=168) | 83 | 36 46 ^a 60 ^{ab} 70 ^{abc} | |
| b | High school grad (n=630) | 88 | | |
| С | Some College (n=588) | 92 ^{ab} | | |
| d | College + (n=834) | 95 ^{abc} | | |
| Hou | usehold income | | | |
| а | Less than \$30,000/yr (n=580) | 86 | 43 52 ^a | |
| b | \$30,000-\$49,999 (n=374) | 90 ^a | | |
| С | \$50,000-\$74,999 (n=298) | 96 ^{ab} | 61 ^a | |
| d | \$75,000+ (n=582) | 98 ^{ab} | 78 ^{abc} | |
| Urb | anity | | | |
| a | Urban (n=763) | 92° | 59 ^c | |
| b | Suburban (n=1,037) | 91° | 59 ^c | |
| С | Rural (n=450) | 85 | 40 | |



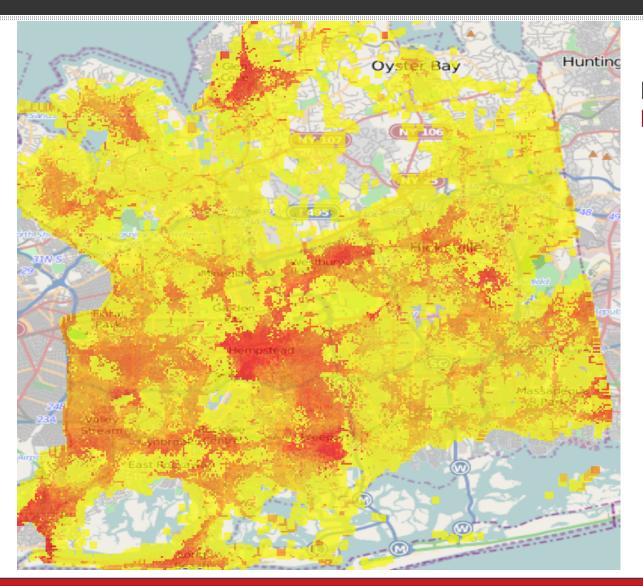
Case Study: Nassau County



Devices with Income > \$100K



Case Study: Nassau County



Devices with Income < \$35K



Population Analytics

| 4 Products | Actionable, Relevant Insights into | | | |
|-------------------------|--|--|--|--|
| Trip Matrix | The number of people and types of trips between two specific locations | | | |
| Select Zone Analysis | The number of people and types of trips that go to, or come from, a specific location and where those trips began or ended | | | |
| Arrivals and Departures | The number of people that go to, or come from, a specific location by time of day | | | |
| Activity Density | The aggregation of all Activity Points with an activity duration of 5 or more minutes in a | | | |
| Add-on Products | For Greater Analytics: | | | |
| Home-Work Matrix | The number of people who live and work in specific locations | | | |
| Home Location Report | The proportion of home locations (by zip code, county or state) for unique mobile devices included in a study | | | |

TRIP MATRIX



Case Study: Moore County, NC

1 month of data

(September 20 – October 18, 2012) consists of:

-8 weekend days*

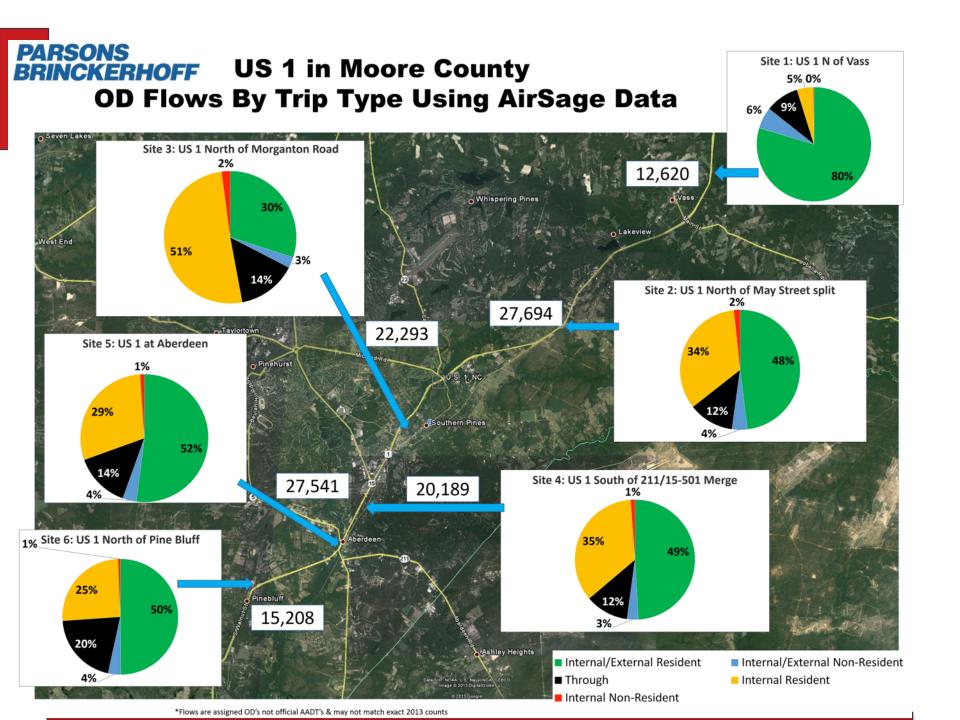
-12 weekdays



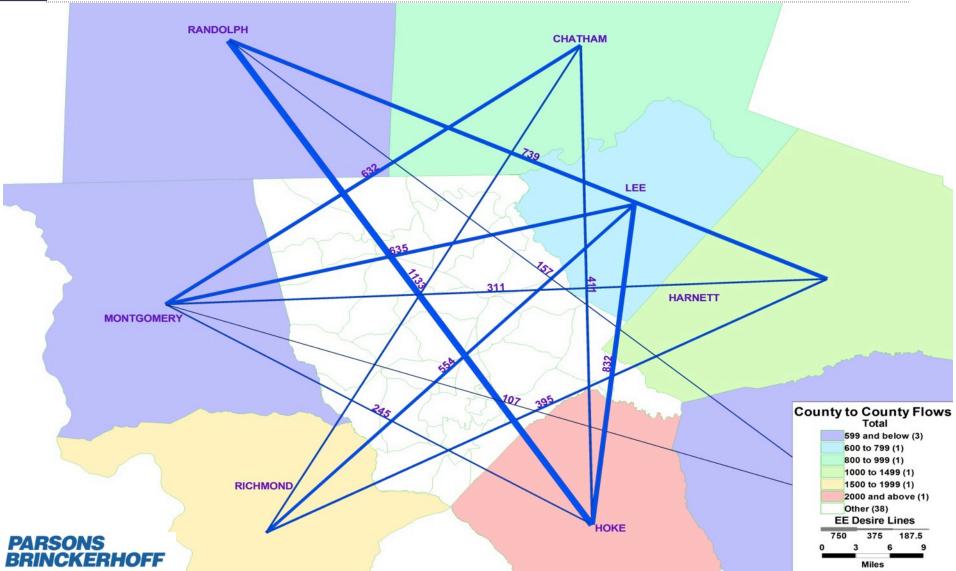
 3,017,382 unique mobile devices seen at least once in focus study area

^{*} Weekend data is only used to help determine the home and work locations of the devices. This data was not used in developing the OD matrices used in the analysis described for Moore County.





County to County Flows (Through Trips)



Assigning AirSage Matrices

"The biggest difference noted is for rural facilities...the TRM (Triangle Regional Model) and AirSage are quite different... AirSage data actually matches the traffic counts for these rural facilities better than the TRM."

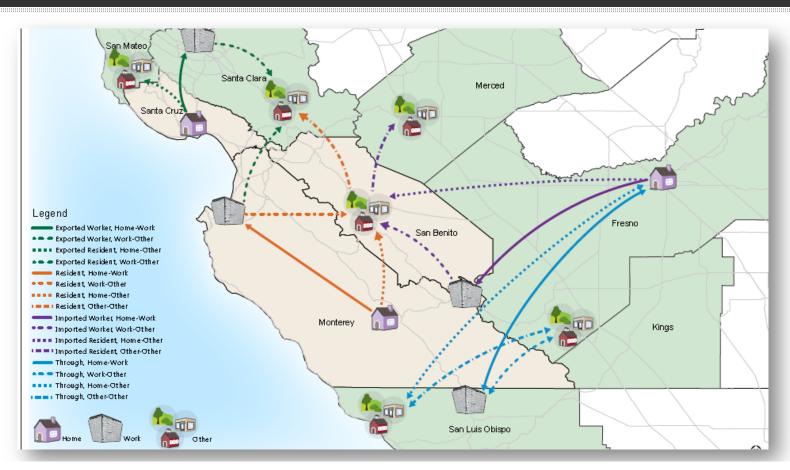
Leta Huntsinger, PhD, P.E.
Parsons Brinckerhoff Systems Analysis Group
"Reconciliation of Regional Travel Model and
Passive Device Tracking Data"



SELECT ZONE ANALYSIS



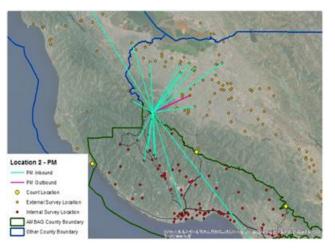
Case Study: Monterey Bay, CA

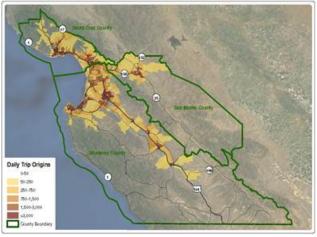


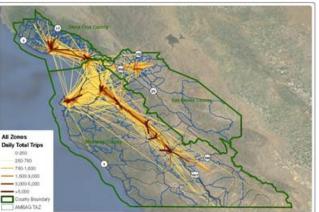
The accuracy and additional data collected with the new license plate video technology alone increases the value of the survey. The ability to also collect cell phone data over 30 days and combine with Census data to infer the household demographics has great potential but has not yet been done. Combining both approaches on the same study is very exciting. We are looking forward to the advances the project can bring to data collection, travel modeling, and transportation planning.

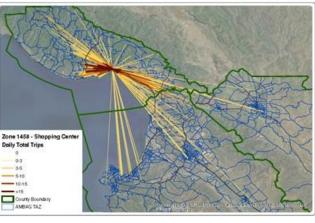
- Anais Schenk, Planner and AMBAG Project Manager

Case Study: Monterey Bay, CA









Data collected – 10 regional external gateways for control total

- Over 24-hrs, 165,000 vehicles observed
- 83,000 unique license plate numbers
- · Addresses obtained from DMV
- 5,300 license plate surveys mailed for more detailed travel characteristics
- Compared to American Community Survey for missing demographics

To supplement & complement, AirSage cellular data was used:

Provided info on all trip types occurring within the AMBAG region including:

- Local resident
- Visitor
- Inter-regional trips
 Data also stratified by 9 different trip purposes

http://asap.fehrandpeers.com/projects/monterey-bay-origin-and-destination-study/



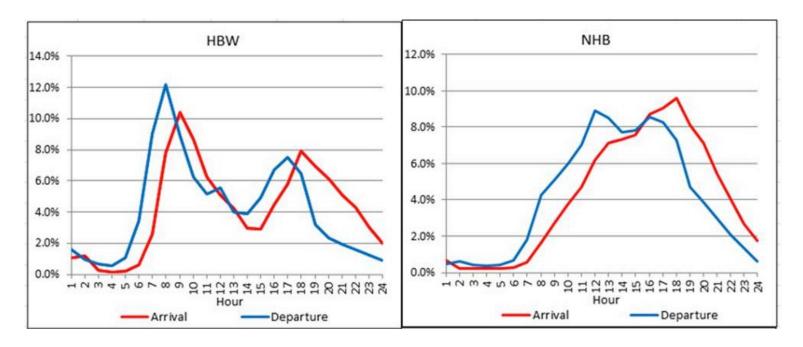


ARRIVALS & DEPARTURES



Arrivals & Departures

The data below represents the Cincinnati, OH region (March 2013). More information about the Ohio-Kentucky-Indiana Regional Council of Governments Travel Demand Model Update can be found on AirSage.com under /News/Success Stories.



"The AirSage data helps validate findings, which is very valuable. If we mess up and the model says it should take 2 lanes and it really needed 4, the cost of getting it wrong could be 10-20 years of people sitting in traffic."

- Andrew Rohne, Transportation Modeling Manager and Project Leader for OKI

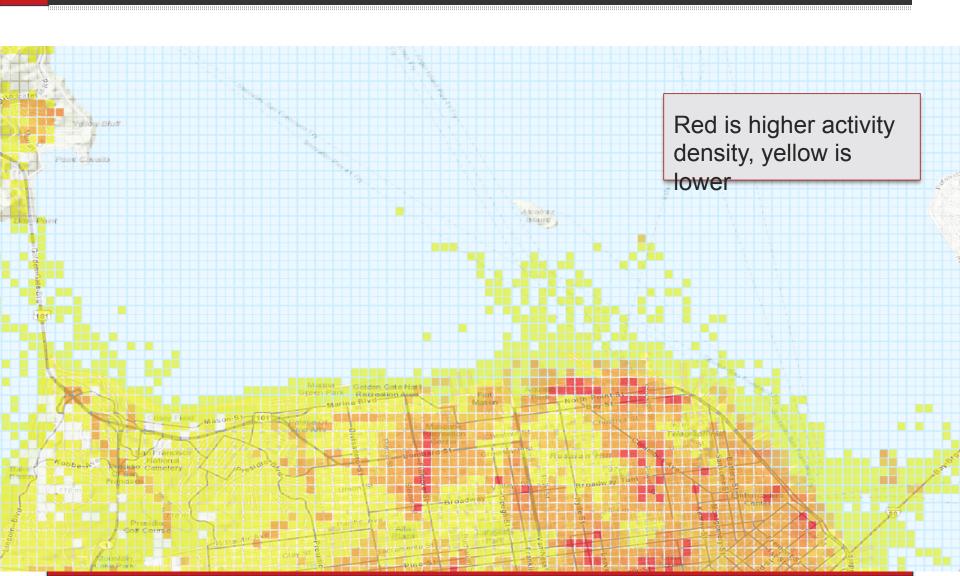
ACTIVITY DENSITY



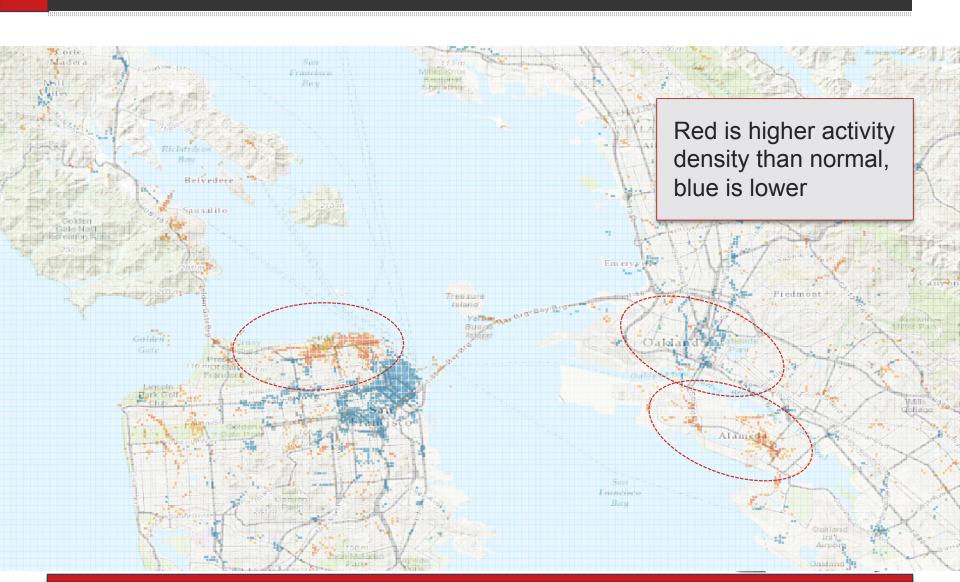
Case Study: America's Cup



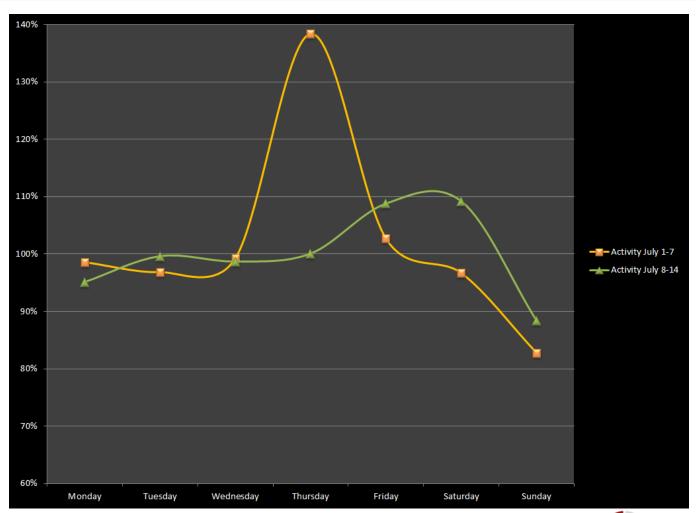
July 4 – Amer. Cup Activity



July 4 – City Difference

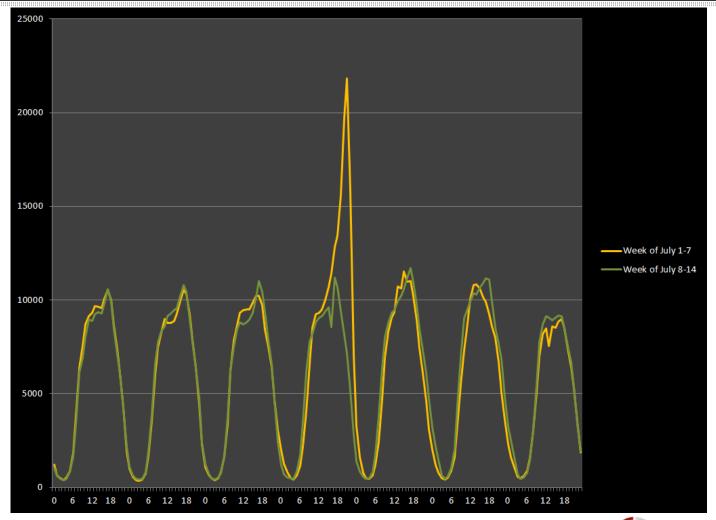


Waterfront, normal versus 4th



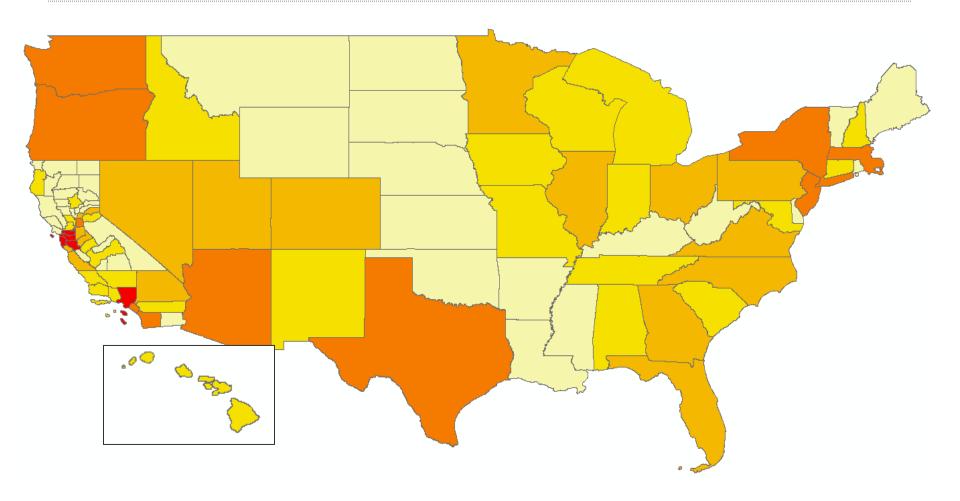


Waterfront Hourly Activity Profiles normal versus 4th



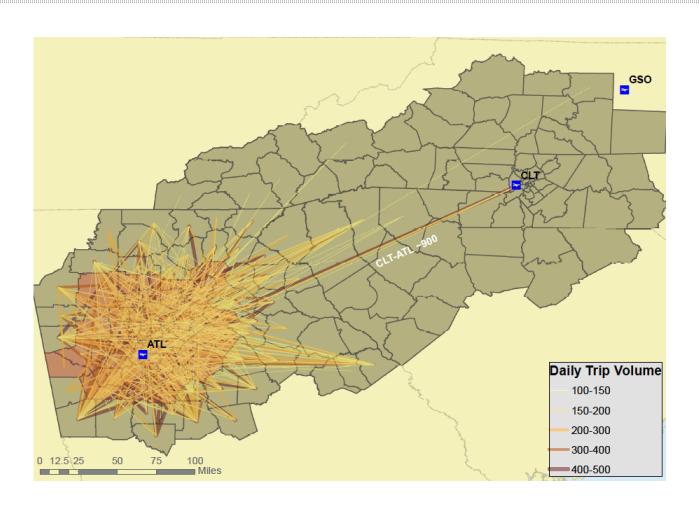


Example: America's Cup 2013 Visitor Home Locations





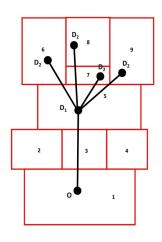
Long Distance Trips



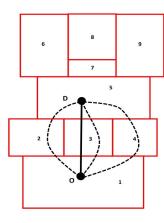


Super and Sub Matrices

Super Matrix



- Super Matrix is a breakdown of trips that were made after a completed trip leg.
- EXAMPLE: For the Zone 1 to Zone 5 trip leg, the count or % of subsequent trips.
 The subsequent trips have an origin D₁ and new destination D₂.



Sub Matrix

- Submatrix is a breakdown of trips that were made through any new zones between the origin and destination.
- EXAMPLE: For the Zone 1 to Zone 5 trip leg, the count or % of trips that that pass through intermediate Zones 2, 3, and 4.



Sample Deliverable

| Origin_Zone | Destination_Zone | Start_Date | End_Date | Aggregation | Time of Day | Counts | Zone_1 | Zone_2 | | | Zone_15 | Zone_16 |
|-------------|------------------|------------|----------|-------------|-------------|--------|--------|--------|----|----|---------|---------|
| 5 | 15 | 20121015 | 20121127 | TotWD | H0:H24 | 200 | 6% | 6% | 6% | 6% | 6% | 6% |
| 7 | 8 | 20121015 | 20121127 | TotWD | H0:H24 | 800 | 6% | 6% | 6% | 6% | 6% | 6% |
| 6 | 3 | 20121015 | 20121127 | TotWD | H0:H24 | 900 | 6% | 6% | 6% | 6% | 6% | 6% |
| 5 | 2 | 20121015 | 20121127 | TotWD | H0:H24 | 1000 | 6% | 6% | 6% | 6% | 6% | 6% |
| 4 | 1 | 20121015 | 20121127 | TotWD | H0:H24 | 1100 | 6% | 6% | 6% | 6% | 6% | 6% |
| 5 | 4 | 20121015 | 20121127 | TotWD | H0:H24 | 300 | 6% | 6% | 6% | 6% | 6% | 6% |
| 8 | 5 | 20121015 | 20121127 | TotWD | H0:H24 | 300 | 6% | 6% | 6% | 6% | 6% | 6% |
| 9 | 6 | 20121015 | 20121127 | TotWD | H0:H24 | 500 | 6% | 6% | 6% | 6% | 6% | 6% |
| 9 | 7 | 20121015 | 20121127 | TotWD | H0:H24 | 700 | 6% | 6% | 6% | 6% | 6% | 6% |
| 15 | 2 | 20121015 | 20121127 | TotWD | H0:H24 | 300 | 6% | 6% | 6% | 6% | 6% | 6% |



Sample of Data Format

| | A | В | С | D | E | F | G | Н | I | |
|----|-------------|----------------------|------------|----------|-------------|--------------|---------|-------------|-------|--|
| 1 | Origin_Zone | Destination_ | Start_Date | End_Date | Aggregation | Subscriber_O | Purpose | Time_of_Day | Count | |
| 2 | 94 | 37 | 20130702 | 20130731 | WD | Visitor | 00 | H00:H24 | 5.08 | |
| 3 | 420 | 343 | 20130702 | 20130731 | WD | Resident | HW | H00:H24 | 1.49 | |
| 4 | 548 | 33 | 20130702 | 20130731 | WD | Resident | WO | H00:H24 | 5.01 | |
| 5 | 68 | 164 | 20130702 | 20130731 | WD | Resident | 00 | H00:H24 | 4.96 | |
| 6 | 256 | 400 | 20130702 | 20130731 | .WD | Resident | НО | H00:H24 | 5.97 | |
| 7 | 498 | 62 | 20130702 | 20130731 | .WD | Resident | HW | H00:H24 | 4.37 | |
| 8 | 1 | 176 | 20130702 | 20130731 | .WD | Resident | OH | H00:H24 | 4.71 | |
| 9 | 54 | 33 | 20130702 | 20130731 | WD | Resident | 00 | H00:H24 | 18.54 | |
| 10 | 255 | 311 | 20130702 | 20130731 | WD | Resident | НО | H00:H24 | 1.65 | |
| 11 | 543 | 85 | 20130702 | 20130731 | WD | Resident | НО | H00:H24 | 1.19 | |
| 12 | 268 | 62 | 20130702 | 20130731 | WD | Resident | OH | H00:H24 | 25 | |
| 13 | 110 | 425 | 20130702 | 20130731 | WD | Resident | WH | H00:H24 | 0.41 | |
| 14 | 1005 | 249 | 20130702 | 20130731 | WD | Resident | WH | H00:H24 | 2.63 | |
| 15 | 1003 | 3 | 20130702 | 20130731 | WD | Visitor | НО | H00:H24 | 4.36 | |
| 16 | 221 | 102 | 20130702 | 20130731 | WD | Resident | НО | H00:H24 | 1.82 | |
| 17 | 12 | 45 | 20130702 | 20130731 | .WD | Resident | HH | H00:H24 | 2.97 | |
| 18 | 290 | 286 | 20130702 | 20130731 | WD | Visitor | 00 | H00:H24 | 2.59 | |
| 19 | 400 | 593 | 20130702 | 20130731 | WD | Visitor | 00 | H00:H24 | 0.46 | |
| 20 | 122 | 160 | 20130702 | 20130731 | WD | Resident | НО | H00:H24 | 1.26 | |



Summary

- AirSage data is...
 - Faster: Validate traditional methods, or replace altogether
 - Easier: Massive sample size using population synthesis
 - Better: Reduced sample bias; extremely wide coverage area
 - Cheaper: Reduced study resources



AirSage data reduced a typical NuStats data collection team of 20-60 analysts to a handful of data analysts and GIS specialists.



Data Validation

OKI Success Story

Surveys, which are done the same as they have been for the last 40-50 years, we don't always know if the data is right. But if we can tap into another source, like mobility data, then we agree that the results are more than likely accurate.

- Andrew Rohne

AIRSAGE CUSTOMER SUCCESS BACKGROUND OHIO-KENTUCKY-INDIANA REGIONAL COUNCIL OF Since 1964, the Ohio-Kentucky-Indiana Regional Council of GOVERNMENTS (OKI) Governments (OKI)—a consortium of local governments, business organizations and community groups-has been developing "The thing I like about the collaborative strategies to improve the quality of life and the AirSage data is that we can economic vitality of the tri-state region. With a primary focus on transportation, OKI cultivates partnerships that help solve set the data to test again and the data set can synthesize the interstate dilemmas, create far-reaching development plans, overcome political bureaucracy, serve the public model. This makes our work and advocate for federal funding. easier and more accurate." - Andrew Rohne OKI was tasked with a daunting deliverable: create an updated ÓKÏ travel model for the tri-state area that would accurately document current travel patterns and forecast transportation needs for the next 30 years. With travel models like this, updated only once every 10-15 years, the last similar survey was done in 1995 and relied upon a combination of data sources, including ABOUT AIRSAGE household studies (HHS), GPS tracking and freeway surveys. A pioneer in population analytics is the largest provider of consumer The new study had two primary goals: 1) measure where people locations and population movement are coming from and where they are going and 2) differentiate intelligence in the U.S. Each day, travelers who are part of the region from travelers who are AirSage uses patented technology to capture and analyze more than 15 simply going through the region. The OKI team, though, had billion anonymous, real-time, cellular to accomplish this-they thought-without access to a key data signal data points to identify travel source: freeway surveys. patterns and transportation trends Partnerships with the nation's largest Like many states across the U.S., Kentucky and Indiana have wireless carriers give AirSage exclusive enacted laws that forbid stopping drivers on the freeway to access to data from more than 100 million mobile devices, Analyzed and conduct surveys. With 98% of all traffic in the area traveling on aggregated, AirSage data provides the freeway, an accurate model would be impossible without actionable insights into where and freeway data. A solution was needed. when people travel and is transforming "The AirSage data helps validate findings, which is very enterprises and a diverse range of industries, www.airsage.com valuable. If we mess up and the model says it should take 2 lanes and it really needed 4, the cost of getting it wrong could be 10-20 years of people sitting in traffic." airsage - Andrew Rohne, Transportation Modeling Manager and Project Leader for OKI

IBM and MIT Research

- Research Concludes AirSage Data
 - Allows for lower collection cost, a larger sample size, higher update frequency, and a broader spatial and temporal coverage
 - Generates detailed traces that can be used to construct path histories with high fidelity across long periods of time
 - Produces audience measurements that are more credible than current static measurements, thus providing rich information to support transportation planning and operation



Notable Awards / Mentions

Gartner

 Gartner, Inc., the world's leading information technology research and advisory company, has named AirSage as one of four Solution Vendors to Watch

Inc.

Inc. magazine named AirSage as one of the nation's fastest-growing
 private companies on its seventh annual Inc. 500|5000 list, ranking AirSage
 No. 2039

BUSINESS INSIDER





Fierce Wireless designated AirSage as one of the top fifteen privatelyheld wireless companies to keep an eye on in 2013



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Twitter - @airsage LinkedIn - http://www.linkedin.com/company/airsage