

# Bike-Sharing & the Built Environment

UT Austin School of Architecture's Urban Information Lab (UIL)

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# Big Picture: How does the built environment relate to bikeshare activity?



Vs.



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- **Bicycle Safety** – cycling infrastructure & high comfort roadways
- **Aesthetics** – Parks/open space & tree canopy
- **Community Character** – density, average structure age, median rent, & median home value
- **Active Transport Connectivity** – sidewalks, bus stops, rail stations, & transit trip frequency

# Past Research Efforts in this Area of Study:

- Noland, Smart, and Guo. 2016. “Bikeshare Trip Generation in **New York** City.” *Transportation Research Part A*.
  - **Positive correlation** with Bikeshare utilization near **subway stations** and **bike infrastructure**
- Wang, X., Lindsey, G., Schoner, J., and Harrison, A. (2016) “Modeling Bike Share Station Activity: Effects of Nearby Businesses and Jobs on Trips to and from Stations”. *Journal of Urban Planning and Development*, 142(1).
  - In Minneapolis, stations associated with neighborhood **sociodemographics**, **proximity to CBD** + water
- [under review] Alcorn, Louis & Jiao, Junfeng. 2018. “Bike Sharing Station Usage and the Surrounding Built Environments in Major Texas Cities”. *Journal of Planning Education and Research*.
  - Limited significance of predictive models except for **high comfort bike facilities in Austin**
- Ma, T., C. Liu, and S. Erdoğan. 2015. “Bicycle Sharing and Transit: Does Capital Bikeshare Affect Metrorail Ridership in Washington, D.C.?” *Compendium of TRB 94th Annual Meeting*: 1-21.
  - People **replacing short-distance bus trips with bikeshare** but still riding metro for longer trips

# Focus on the Largest 4 US Systems

- New York Citibike
- Chicago Divvy
- Washington D.C. Capital Bikeshare
- Boston Hubway

City	Start Date	End Date	Stations	Avg. Daily Station Use	Maximum	Minimum
New York	Jul-13	Oct-17	832	102	575	1
Chicago	Jun-13	Jul-17	631	38	485	1
Wash. D.C.	Sep-10	Apr-17	512	51	319	1
Boston	Jul-11	Oct-17	212	21	86	1

# Dependent Variable

Average daily bikeshare station usage

$$= \frac{(\# \text{ check outs} + \# \text{ check ins})}{(\# \text{ of days station is open})}$$

# Independent Variables

- High Comfort Roadways (People For Bikes dataset)
- Bike Network - Any Treatment above “sharrows” (City/County GIS portals)
- Open Space/Park Area (City/County GIS portals)
- Sidewalk Area (City/County GIS portals)
- Trees/Tree Canopy (City/County GIS portals)
- Population Density, Housing Unit Density, Median Gross Rent, Median Home Value, Median Structure Age (2016 ACS 5-yr. est.)
- # of Bus/Rail Stops within  $\frac{1}{4}$  mile of bikeshare stations & # of unique daily transit trips available (Historic GTFS feeds)

# Methodology

- Built environment inventory within  $\frac{1}{4}$  mile of bikeshare station:
  - Generate  $\frac{1}{4}$  mile airline buffer around station areas
  - Intersect independent variable data with these  $\frac{1}{4}$  mile access-sheds
  - Spatially join dependent variable count/area/length data with each station location buffer
  - Bivariate correlations with Pearson's R
  - Forward Stepwise Regression to account for covariance of dependent variables.

Join Data

Join lets you append additional data to this layer's attribute table so you can, for example, symbolize the layer's features using this data.

What do you want to join to this layer?

Join data from another layer based on spatial location

1. Choose the layer to join to this layer, or load spatial data from disk:

CHIEVA\_bikelane\_intersect

2. You are joining: Lines to Polygons

Select a join feature class above. You will be given different options based on geometry types of the source feature class and the join feature class.

☒ Each polygon will be given a summary of the numeric attributes of the lines that intersect it, and a count field showing how many lines intersect it.

How do you want the attributes to be summarized?

☐ Average ☐ Minimum ☐ Standard Deviation  
☒ Sum ☐ Maximum ☐ Variance

☐ Each polygon will be given all the attributes of the line that is closest to its boundary, and a distance field showing how close the line is (in the units of the target layer).

Note: A line falling inside a polygon is treated as being closest to the polygon, (i.e. a distance of 0).

3. The result of the join will be saved into a new layer.

Specify output shapefile or feature class for this new layer:

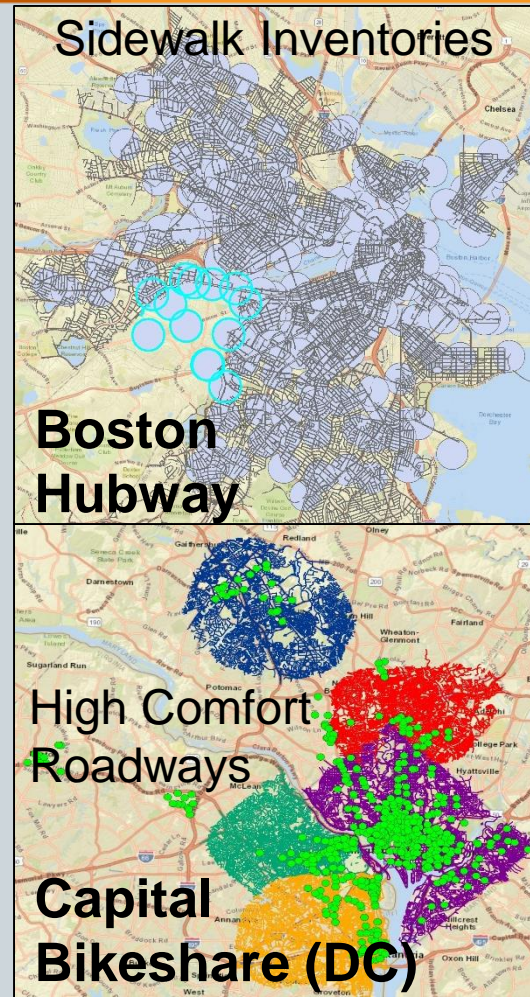
C:\Users\sweet\Documents\UILL\ChicagoBoston-SPRING2018\

[About joining data](#)



# Limitations of Data

- Weaving together multiple datasets
  - Boston = Boston + Cambridge + Brookline + Somerville
  - Chicago = Evanston + Chicago
  - NYC = NYC + Jersey City (NJ)
  - Washington D.C. = DC + Alexandria (VA) + Arlington (VA) + Bethesda (MD) + Rockville (MD) + Fairfax County (VA) + Montgomery County (MD), and more!



# Preliminary Results: Chicago

<i>R</i>	<i>R-Squared</i>	<i>Adjusted R-Squared</i>	<i>N</i>
0.7681	0.5900	<b>0.5839</b>	609
<i>VAR</i>	<i>Coefficient</i>	<i>p-value &gt; t</i>	<i>TOL</i>
<b>GrossRent</b>	0.0342	5.7086E-9	0.2348
<b>HouDens</b>	0.5171	3.4071E-9	0.4789
<b>OpenSpaceContinuousAccessSqMeters</b>	5.7387E-6	0.0001	0.4015
<b>BikeLaneIntersectMeters</b>	0.0021	1.9045E-9	0.3546
<b>SidewalkAreaSqMeters</b>	-6.4038E-5	0.0031	0.4214
<b>PCB_HC_Meters</b>	-0.0018	0.0314	0.7208
<b>OpenSpaceIntersectSqMeters</b>	2.4046E-5	0.0034	0.4763
<b>REValue</b>	3.4652E-5	0.0122	0.3471
<b>TreeCanopySqMeters</b>	-3.3078E-5	0.0141	0.6626
<b>Intercept</b>	-30.0444		

# Preliminary Results: Boston

<i>R</i>	<i>R-Squared</i>	<i>Adjusted R-Squared</i>	<i>N</i>
0.6668	0.4446	<b>0.4243</b>	171
<i>VAR</i>	<i>Coefficient</i>	<i>p-value &gt; t</i>	<i>TOL</i>
<b>REValue</b>	2.5446E-5	0.0002	0.5002
<b>DistRail_meter</b>	-0.0088	0.0024	0.6172
<b>HCBikeLength</b>	0.0004	0.0345	0.7128
<b>GrossRent</b>	0.0080	0.0031	0.6001
<b>PopDens</b>	0.1062	0.0582	0.5721
<b>RailStop</b>	-1.2046	0.0875	0.6209
<b>Intercept</b>	-12.9255		

# Preliminary Results: NYC

<i>R</i>	<i>R-Squared</i>	<i>Adjusted R-Squared</i>	<i>N</i>
0.6974	0.4863	<b>0.4802</b>	765
<i>VAR</i>	<i>Coefficient</i>	<i>p-value &gt; t</i>	<i>TOL</i>
<b>GrossRent</b>	0.0347	9.3283E-5	0.3605
<b>Bikelane_Length_M</b>	0.0024	1.2693E-6	0.3648
<b>HouDens</b>	1.6148	2.7720E-11	0.0876
<b>TreeCount</b>	-0.1181	0.0000	0.7588
<b>DistRail_meter</b>	-0.0426	0.0002	0.8366
<b>OpenSpaceArea</b>	-3.0130E-5	0.0045	0.9544
<b>PFB_HC_Length</b>	0.0007	0.0022	0.3686
<b>PopDens</b>	-0.4346	0.0061	0.0891
<b>REValue</b>	3.7006E-5	0.0105	0.6812
<b>Intercept</b>	-3.5885		

# Trends

- Always Positive Coefficients:
  - Median Gross Rent (3)
  - Median Home Value (3)
  - Bike Lane Length (2)
- Always Negative Coefficients:
  - Trees/Tree Canopy Area (2)
  - Distance to Rail Stops (2)
- Mixed Results:
  - High Comfort Cycling Roadways (3)
  - Population Density (2)
  - Open Space (2)
- Limited Significance:
  - Sidewalk area (1)
  - Median Structure Age (0)
  - Frequency of transit trips available within  $\frac{1}{4}$  mile of bikeshare station (0)
  - Proximity to Bus Stops (0)

# Work Ahead

- Add Transit-related independent variables to NYC model
- Create Model for Washington D.C.
- Combine all four models to make a fifth 4-city model (pending data continuity)
- Submit to present at TRB

# Questions?

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