

University of Denver Analytics Challenge

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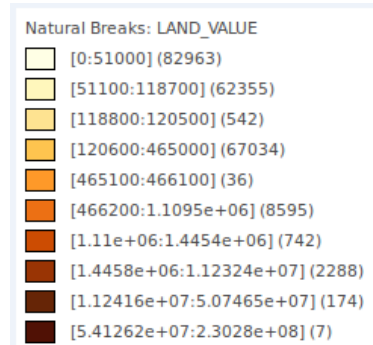
Colorado School of Mines

March 4, 2016

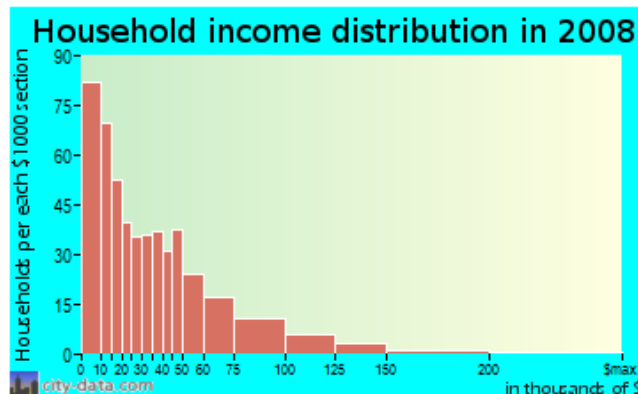


Social Control Theory: Society “Buy In”

■ Example Case: Montbello

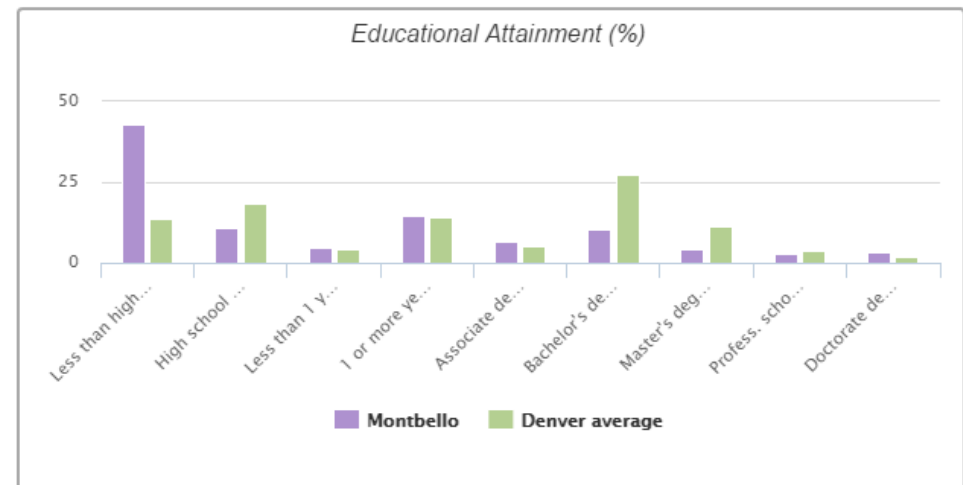


<http://data.denvergov.org/dataset/city-and-county-of-denver-crime>



Median household income in 2013:

Montbello: \$46,175
Denver: \$51,089

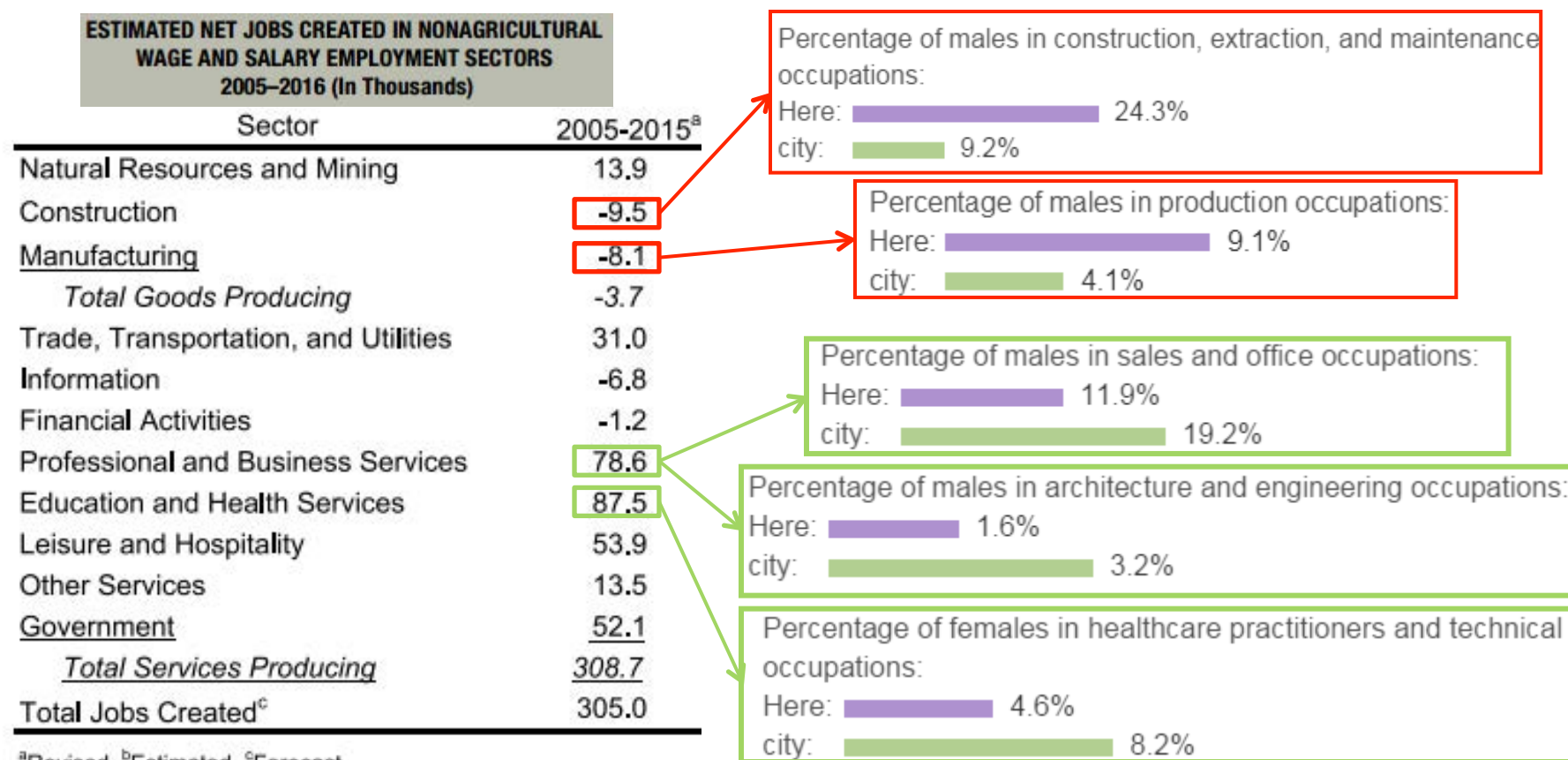


<http://www.city-data.com/neighborhood/Montbello-Denver-CO.html>



Input Data for Society “Buy In”

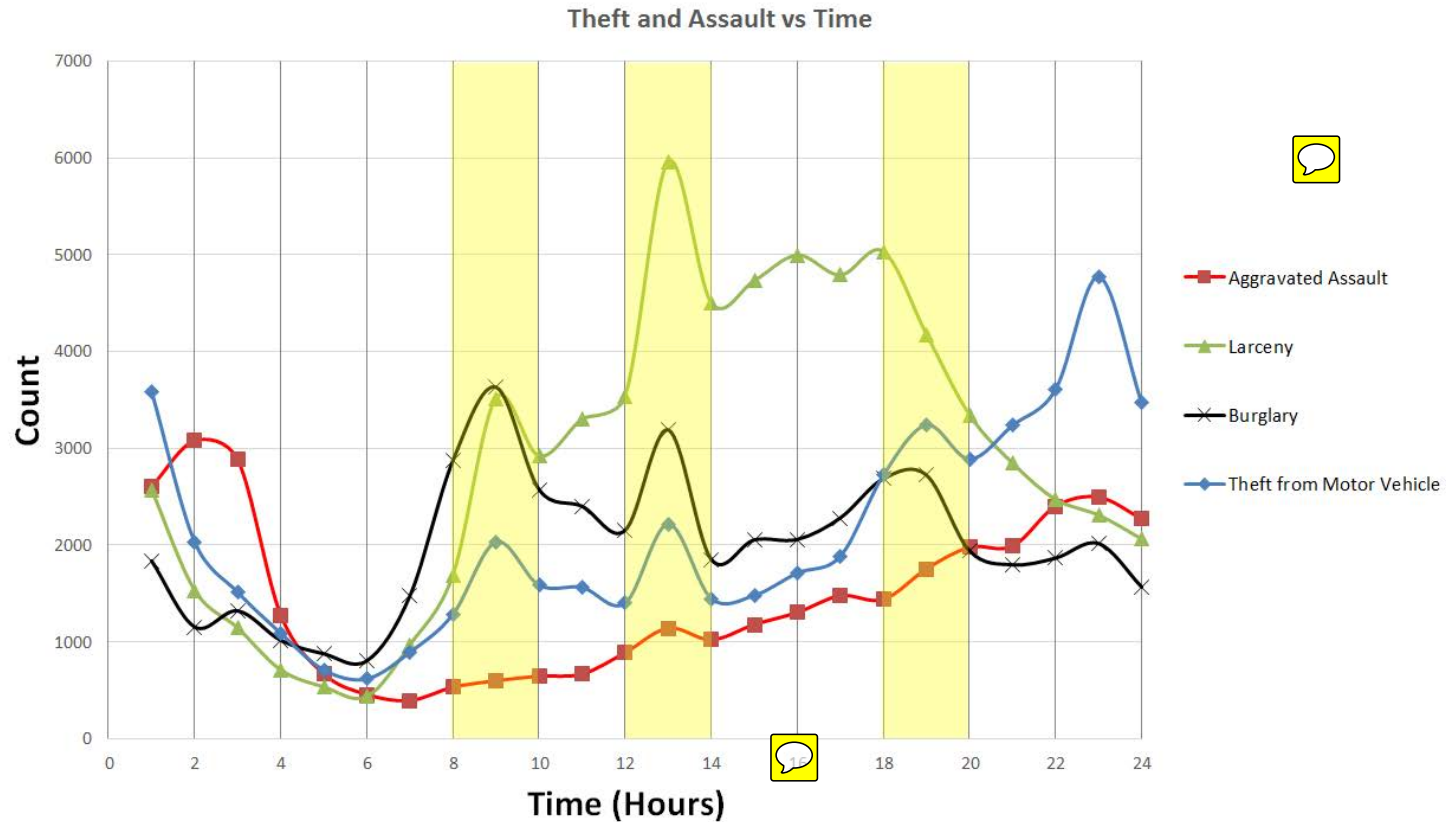
■ Montbello Economy vs. Colorado



<http://www.city-data.com/neighborhood/Montbello-Denver-CO.html>

http://www.colorado.edu/business/sites/default/files/attached-files/colo_business_econ_outlook_2016.pdf

Theft and Assault Time Analysis



Crime Category	Total	Weekday	Weekend	Weekend (%)
<i>Burglary</i>	48178	36911	11267	23.39%
<i>Larceny</i>	70126	51657	18469	26.34%
<i>Theft from Motor Vehicle</i>	51016	36593	14423	28.27%
<i>Simple Assault</i>	49266	32719	16547	33.59%
<i>Aggravated Assault</i>	35192	20590	14602	41.49%

Hotspots Time Analysis : Larceny



Time	Patterns and Location
0 – 300	-Parking lots -Fast Food Restaurants
300-600	-Decrease in theft
600-800	- Slow Increase in larceny - Denver Airport on the radar - Parks and country clubs
800-1200	-Steady rise
1200-1400	- Sudden increase around restaurants (lunch) and near departmental stores
1400-1800	- Steady rate - Denver Airport is not a hot spot after approximately 3 pm
1800-2100	-Decline in larceny -Downtown hot spots translated (<i>right top figure</i>)
2100-2300	-Steady rate
2300-2400	-Decline in larceny

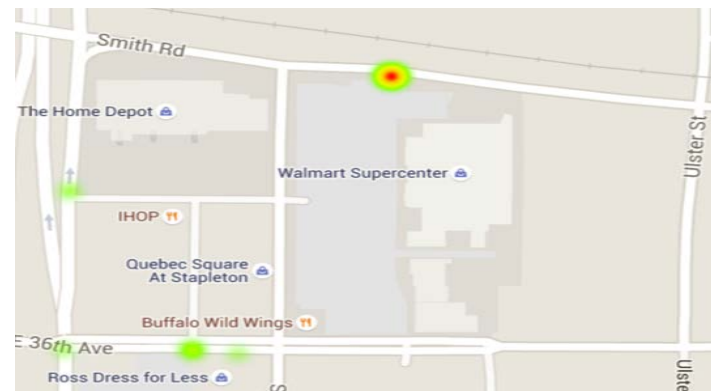
HOTSPOTS : DOWNTOWN

**14th / 19th-20th
street (10 pm
– 6 am)**



**16th – 17th
Street (6 am
– 10 pm)**

HOTSPOT : ISOLATED EVENT



7800 Smith Rd, Denver, CO 80207



Statistical Methodology

- Statistically, crimes are an example of events in a *spatial point process*, generated by $Pois(\lambda)$.



$$\lambda(x) = \lim_{|ds| \rightarrow 0} \frac{\mathbb{E}[N(ds)]}{|ds|}$$

where $N(\cdot)$ is the number of points in the distance ds away from any point s .

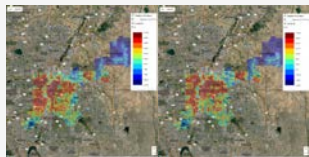


- The first assumption is *Complete Spatial Randomness*. This hypothesis was tested and the data do not display this behavior.
- Therefore, we modeled the crimes as spatially-varying a.k.a. an *Inhomogeneous Poisson Process* and estimated λ for each type of crime for each year with kernel-density smoothing:

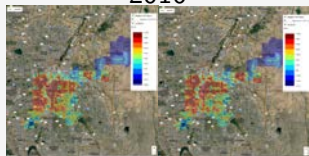


$$\hat{\lambda}(s) = \sum_{x \in N \cap D} \frac{k[(x - s)/h]}{c(s)}$$

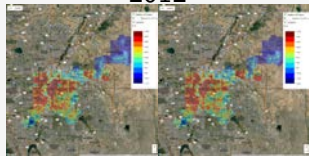
Simple and Aggravated Assault vs Street Checks



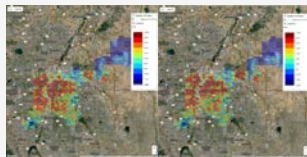
2010



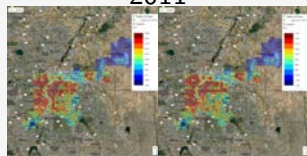
2012



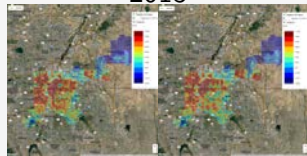
2014



2011



2013



2015

Optimization Model

Parameters:

- *Weight of crimes
- *Location of crimes

Variables:

- *Location of each patrol

Objective:



- *Minimize total distance between patrols and crimes in surrounding area

Constraints:

- *At least one patrol serves each crime
- *Crimes are evenly distributed between patrols



A scatter plot showing the spatial distribution of crimes and police patrols in San Francisco. The x-axis represents Longitude, ranging from -105.15 to -104.7. The y-axis represents Latitude, ranging from 39.6 to 39.82. Blue diamonds represent 'Crimes' and red squares represent 'Patrols'. The plot shows a high density of crimes in the central urban area, with patrols distributed across the city, often appearing in clusters or along major thoroughfares. Two yellow speech bubble icons are present: one on the left side of the plot and one at the bottom center.

