

Addressing Unobserved Neighborhood Quality in Hedonic Models Using Google Street View Imagery and Urban Perception Data

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Unobserved Neighborhood Quality in Housing Hedonics

$$p(x, g) = \alpha \cdot h(x) + \beta \cdot g + \epsilon$$

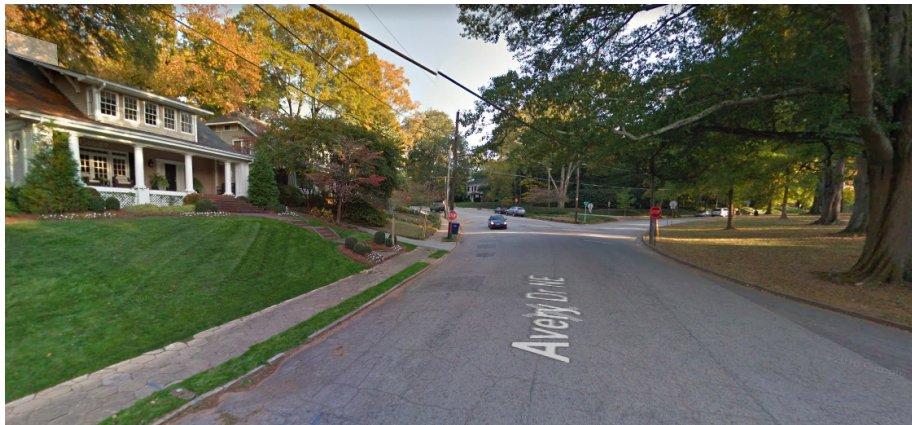
p = price

x = housing characteristics

g = neighborhood characteristics (schools, pollution, etc.)

Concern: $Cov(g, \epsilon) \neq 0$

Aesthetic Character of Neighborhoods



Aesthetic Character of Neighborhoods



- Extensive coverage
 - Every MSA in the US
 - increasing global coverage
- Data collection began in 2007, many cities now have multiple years of observation

But how do we use this information?

Place Pulse Survey

PLACE PULSE

1,553,128 clicks

[Vision](#)

[Rankings](#)

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Which place looks safer ?



For this question: **506,606** clicks collected

Goal: **500,000** clicks

Place Pulse Survey

- 56 cities from 28 countries across 6 continents
- Over 100,000 images
- Over 1.5 million comparisons

Enough data to *generalize*

Binary Comparisons \rightarrow Rankings

Treat each comparison as a 'competition' between images, where performance levels are drawn from a normal distribution centered on each contestant's rank. Iteratively update rankings with the Microsoft 'TrueSkill' algorithm.

$$\mu_{winner} \leftarrow \mu_{winner} + \frac{\sigma_{winner}^2}{c} \cdot v \left(\frac{(\mu_{winner} - \mu_{loser})}{c}, \frac{\varepsilon}{c} \right)$$

$$\mu_{loser} \leftarrow \mu_{loser} - \frac{\sigma_{loser}^2}{c} \cdot v \left(\frac{(\mu_{winner} - \mu_{loser})}{c}, \frac{\varepsilon}{c} \right)$$

$$\sigma_{winner}^2 \leftarrow \sigma_{winner}^2 \cdot \left[1 - \frac{\sigma_{winner}^2}{c^2} \cdot w \left(\frac{(\mu_{winner} - \mu_{loser})}{c}, \frac{\varepsilon}{c} \right) \right]$$

$$\sigma_{loser}^2 \leftarrow \sigma_{loser}^2 \cdot \left[1 - \frac{\sigma_{loser}^2}{c^2} \cdot w \left(\frac{(\mu_{winner} - \mu_{loser})}{c}, \frac{\varepsilon}{c} \right) \right]$$

$$c^2 = 2\beta^2 + \sigma_{winner}^2 + \sigma_{loser}^2$$

Deep Learning The City

