

# Combining household survey data, key informant interviews, and hedonic modelling to understand housing demand in a dynamically shifting market

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**SCHOOL OF PLANNING**



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[environment.uwaterloo.ca](http://environment.uwaterloo.ca)

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# Why are land markets important?

- On the investment side, huge role in the financial system—look no farther than the 2009 housing crisis and Great Recession
- On the individual side:
  - Often the most significant asset/investment for home owners
  - “Lifestyle” contributions from house/neighbourhood strongly linked to identity
- On the market side, they influence:
  - Income and racial segregation
  - Tiebot sorting, public good provision (education), politics

# Land markets: Regional and global scale influences

- Housing market crash and global financial crisis
- Rising gas prices and fall of residential land values in low-accessibility areas
- Global trends towards urban migration and urbanization
- Biofuel initiatives, rising agricultural commodity prices, and agricultural land markets
- “Land Grabs”
- REDD (reducing emissions from deforestation and degradation) and emerging carbon markets

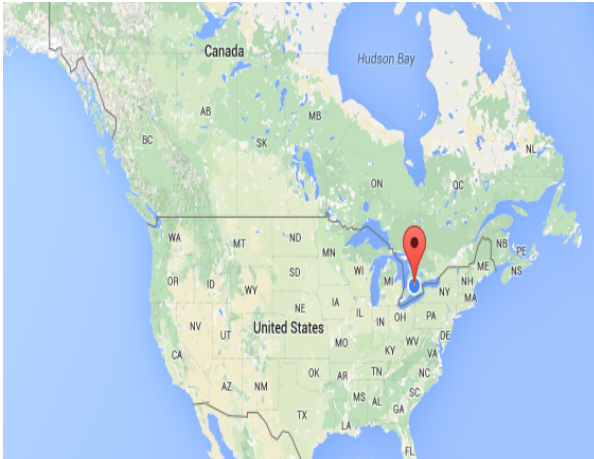
## Why are land markets different?

- Each property is a unique good—extreme case of monopolistic competition
- Few opportunities for repeat transaction -> limited opportunities for learning
- Information on prices is largely private (bids and transaction prices)
- Interaction between markets for a personal good (homeowner resident) and investment good
- Highly influenced by demographic shifts/migration
- Limited competition/innovation in supply
- Highly regulated supply process



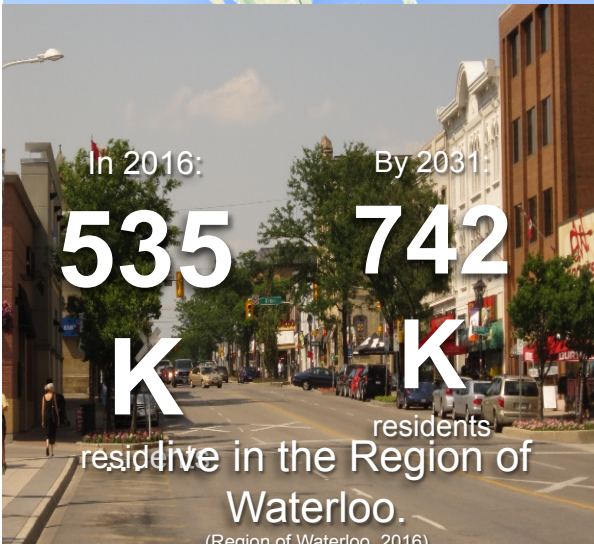
# Why are land market complex?

- **Heterogeneity:**
  - Differential resource and preferences
  - Differential risk
  - Differential knowledge and beliefs
- **Interdependencies**
  - Credit networks
  - Land markets
  - Transportation/proximity
  - Spatial spillovers (externalities)
- **Learning and adaptation**
- => Non-linearities, Analytical intractability,



# Why Kitchener-Waterloo Region?

- **Increasing people** and employment
- **High tech hub** with entrepreneurship and knowledge-intensive economy
- A new light rail transit system as a key strategy for urban revitalization and overall economic development strategy
- **Housing boom (price volatility), but why?**



*Toronto speculative buyers?*

*Foreign buyer tax?*

*Low interest rate?*

*Urban growth boundary?*

*Massive shortage?*

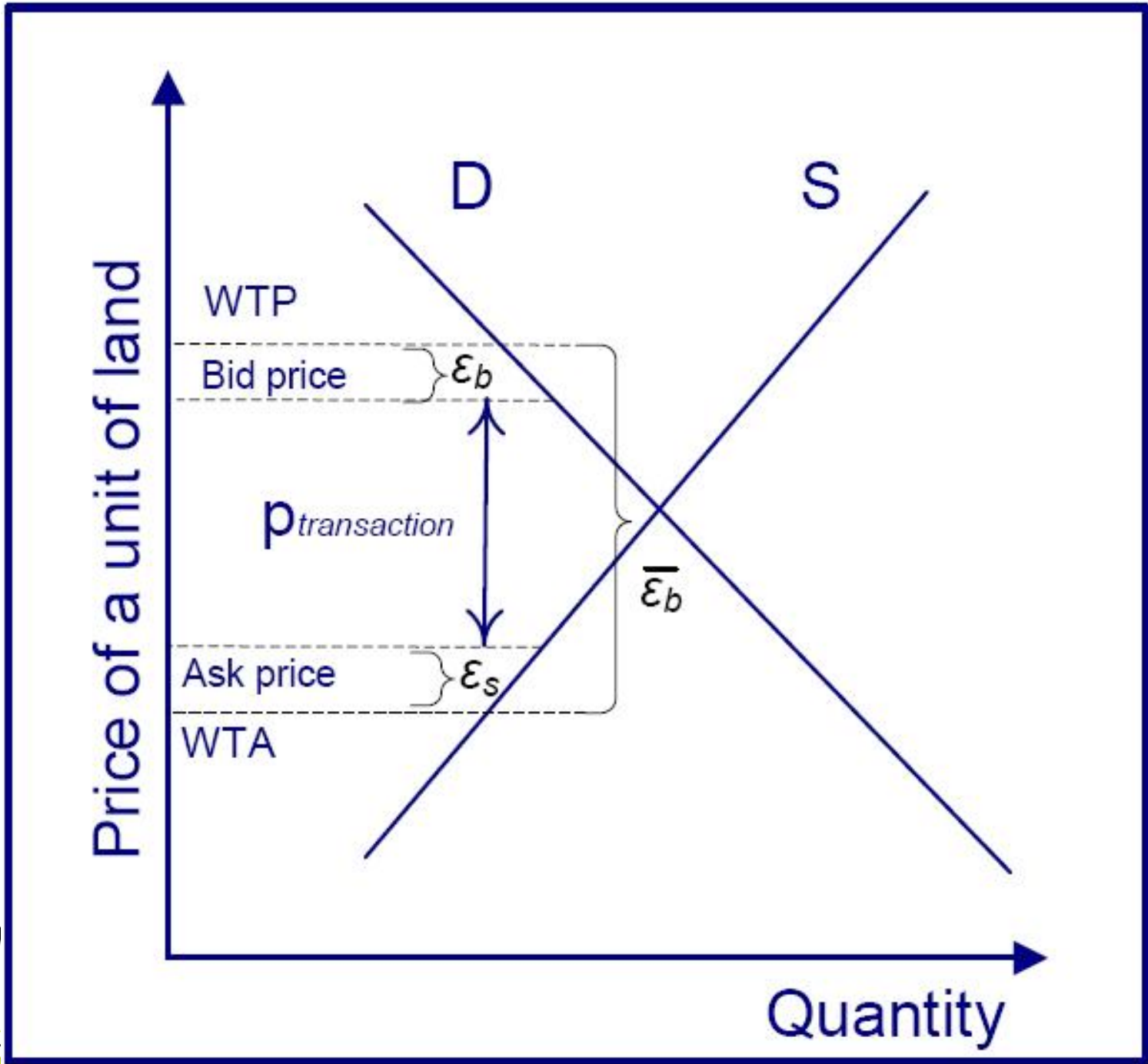


## Research Questions...

1. *How can we better interpret the housing market dynamics in Kitchener-Waterloo Region?*
2. *What are the housing demand or preferences among heterogeneous households during the boom?*
  - *How can we analyze the housing demand?*
  - *Specifically, how can we build a **theoretically-grounded, empirical model** to interpret **housing demand** in this Region?*

# Challenges in development of ABMs of land market

- Early models had endogenous prices, but no decentralized transactions
- Price formation processes mimicked Alonso/Von Thuenen models—allocation parcel to highest profit use, or using numerical price adjustment mechanisms
- Filatova and Parker (various 2008-2010) first to our knowledge to implement decentralized price formation through bilateral transactions
- Challenges for building these models outlined in Parker and Filatova (2008)
- Achievements summarized in Huang et al. (2014)



# Major challenge: Empirical WTP/WTA estimation

- Spatial econometric models estimate transaction price only
- WTA/WTP depend on house-hold level factors
- Household information difficult to obtain
- Our theoretical ABMs used budget-constrained utility maximization—but that created a challenge of how to model price expectations
- Seemed to be no ground to empirically estimate utility function parameters

# Approach one: Spatial hedonic regression models

# Model Overview

- Waterloo Regional Model (WARM)
- Vector-based parcel landscape
  - Represents individual households and parcels
  - Practical given data constraints
  - Accurately represents transportation costs and accessibility
- Land market model
- Transportation model



# Hedonic Model (Babin)

- Statistical model to deconstruct property value
- Identifying relationship between intensification related environmental amenities (like open space and transit access) and property prices
  - controls for spatial and aspatial home characteristics

$$\ln(Y_i) = \beta_0 + \beta_1 S_i + \beta_2 E_i + \beta_3 N_i + \varepsilon$$

Where:

$Y_i$  = Property value (preferably recorded sales price)

$S_i$  = Structural characteristics

$\beta_0$  = Intercept

$E_i$  = Environmental characteristics

$\beta_{1-3}$  = Estimated coefficients

$N_i$  = Neighbourhood characteristics

$\varepsilon$  = Error

# Modelling Access to Public Open Space

## Gravity-Based Open Space Access Kitchener-Waterloo, Ontario

$$A_i = \sum_{\substack{j=1 \\ i \neq j}} W_j \times S_{ij}^{-\alpha}$$

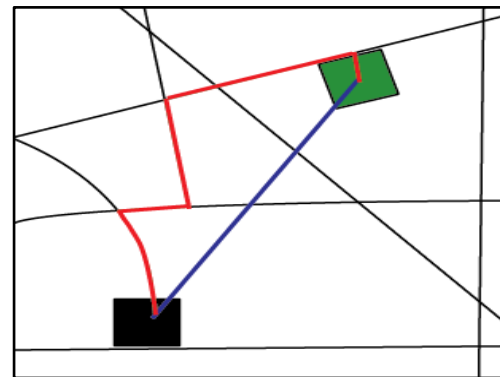
Tsou et al., 2005

Access

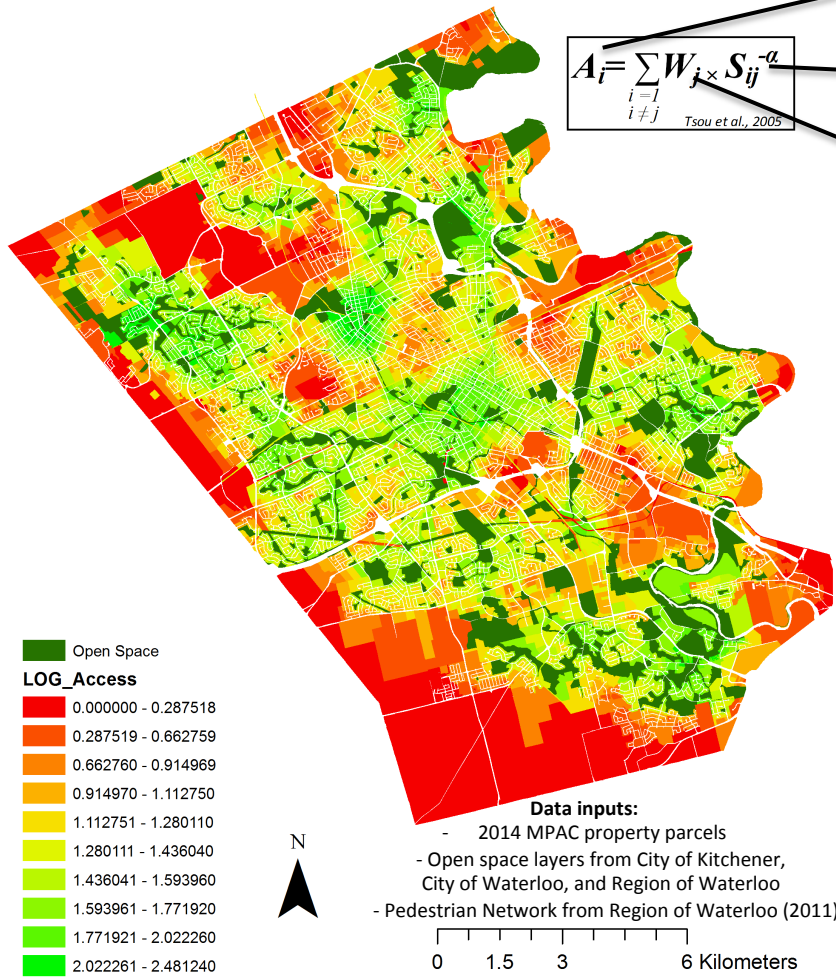
Spatial Separation (distance)

Weight/Attractiveness  
(size of open space)

### Network (walking) Distance



	Open Space		Euclidean Distance
	Property		Network Distance
	Pedestrian network		

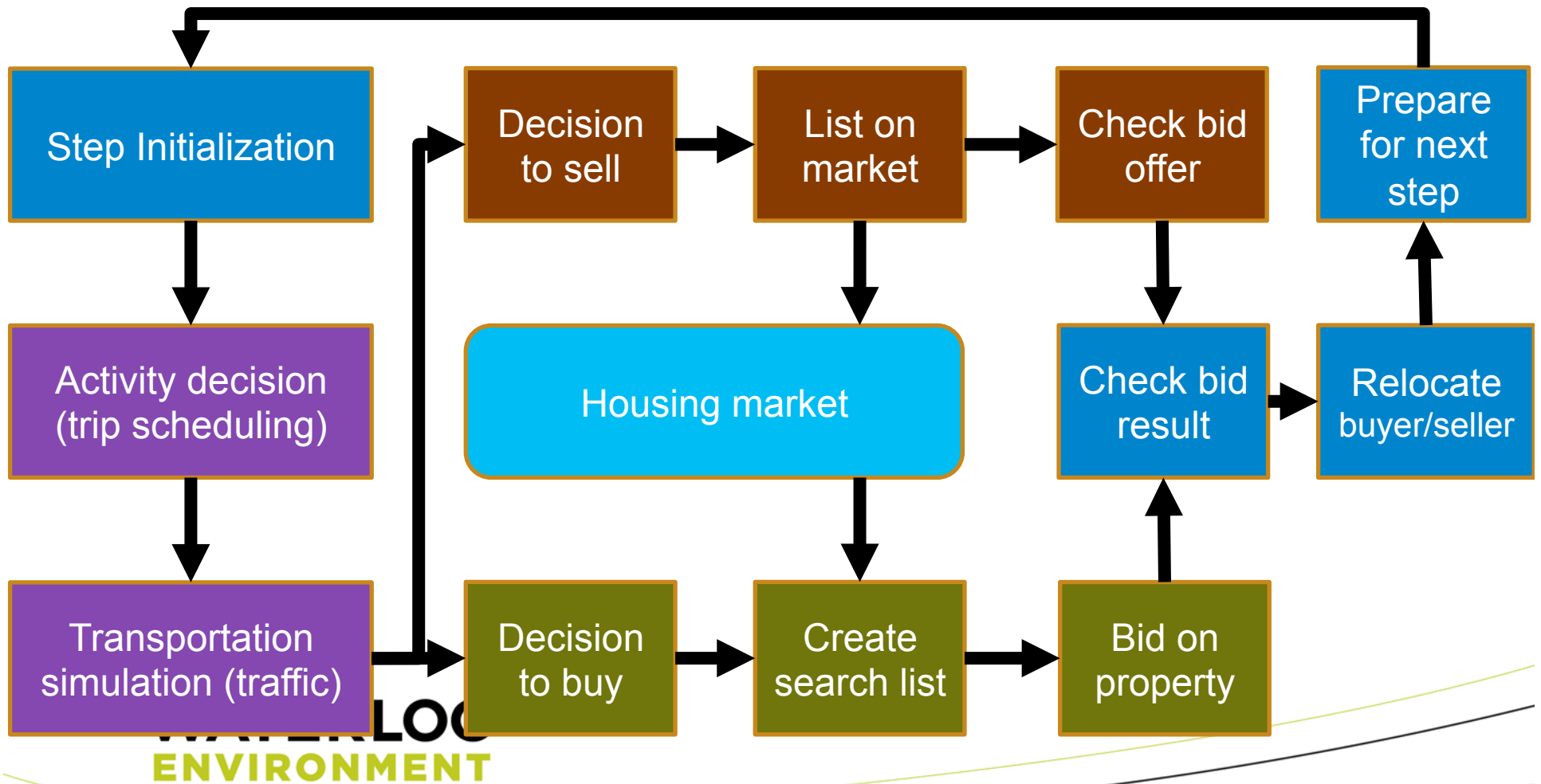


Figures by Robert Babin

## Hedonic model highlights

- Model run using data from 2005-2015, to establish pre-LRT baseline
- House characteristics (size, age, parcel size) strongly correlated with values, as expected.
- Neighbourhoods with higher appreciation rates showed higher values
- After 2011, houses inside the CTC sold for around 4.5% more than houses outside
- Walkability showed a premium; more so inside the CTC

# Agent-based Model process



# Land Market: WTA/WTP

- Seller's willingness to accept (WTA) assessment values
- Buyer's willingness to pay (WTP) – transaction values
- Both estimated via spatial econometric regression

$$\ln(Y_i) = \rho W y_i + \beta_0 + \beta_1 \times S_i + \beta_2 \times N_i + \beta_3 \times E_i + u_i$$

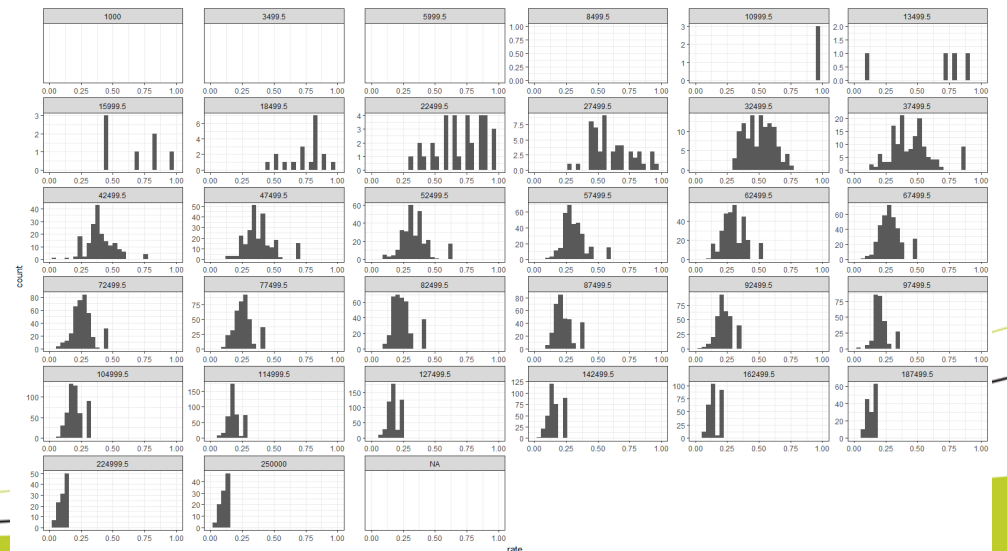
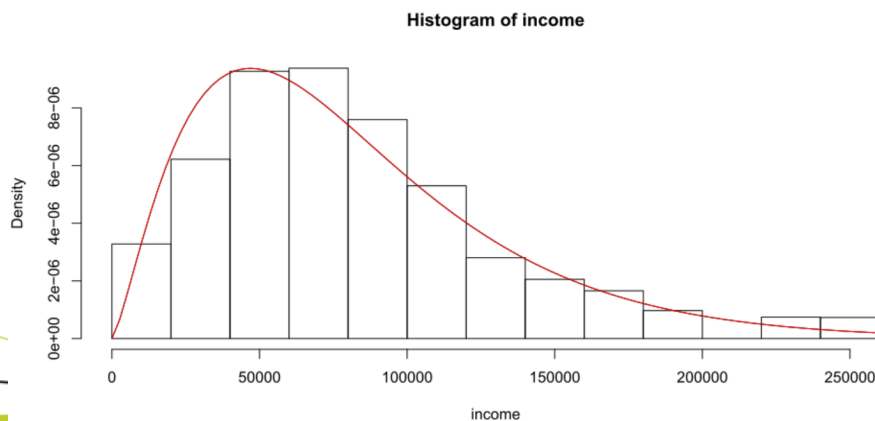
$$S_i = \begin{bmatrix} \text{Living Area}_i \\ \text{Yard Size}_i \\ \text{Building Age}_i \end{bmatrix}$$

$$N_i = \begin{bmatrix} \ln CTC_i \\ \text{Rate of Appreciation}_i \\ \text{Education Rate}_i \\ \text{Population Density}_i \\ \text{Time Period}_i \\ \ln CTC \times \text{Time Period}_i \end{bmatrix}$$

$$E_i = \begin{bmatrix} \text{Open Space Access}_i \\ \text{Transit Access}_i \\ \text{Walkability}_i \\ \text{Open Space Adjacent}_i \\ \text{Regional Road Adjacent}_i \\ \text{Open Space Access} \times \text{Yard Size}_i \\ \ln CTC \times \text{Open Space Access}_i \\ \ln CTC \times \text{Transit Access}_i \\ \ln CTC \times \text{Walkability}_i \\ \text{Open Space Access}_i \times \text{Transit Access}_i \\ \text{Open Space Access}_i \times \text{Walkability}_i \\ \text{Walkability}_i \times \text{Transit Access}_i \end{bmatrix}$$

# Land Market: Budget Constraints

- Regulatory limit in Canada: about 32% of income
- higher-income households spend significantly less portion of their income on housing
  - Model settings: generated using normal distributions. Mean at 25% for households with income of 40k, and down to 8% for households with income of 250k.
- Income generated using a Gamma distribution based on census data



# Land Market: Competitive Bidding

- Each buyer can bid on only 1 parcel each step
- Buyers bid on the parcel that provide the highest utility value (currently using WTP as proxy)
- Seller choose the highest bid that is above WTA

# Transportation Simulation

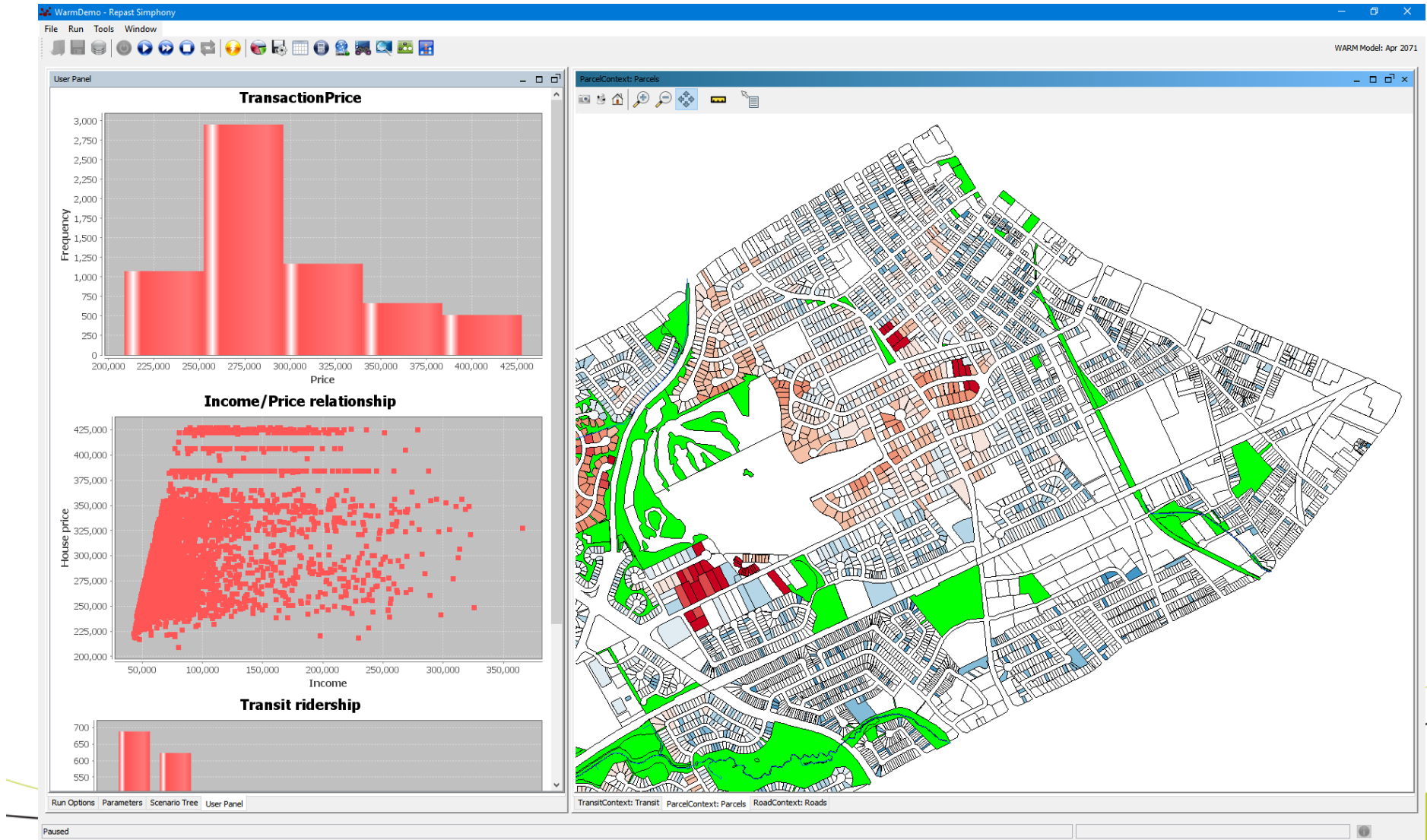
- Uses Transportation Tomorrow Survey (TTS) trip diary data
- Supplemented by Origin-Destination matrices from Region's transportation model
- Each household has its unique travel schedule, trip destinations, purposes and modes
- Internal shortest route calculation for car travel
- Utilizes OpenTripPlanner to calculate shortest transit route based on route and schedule data



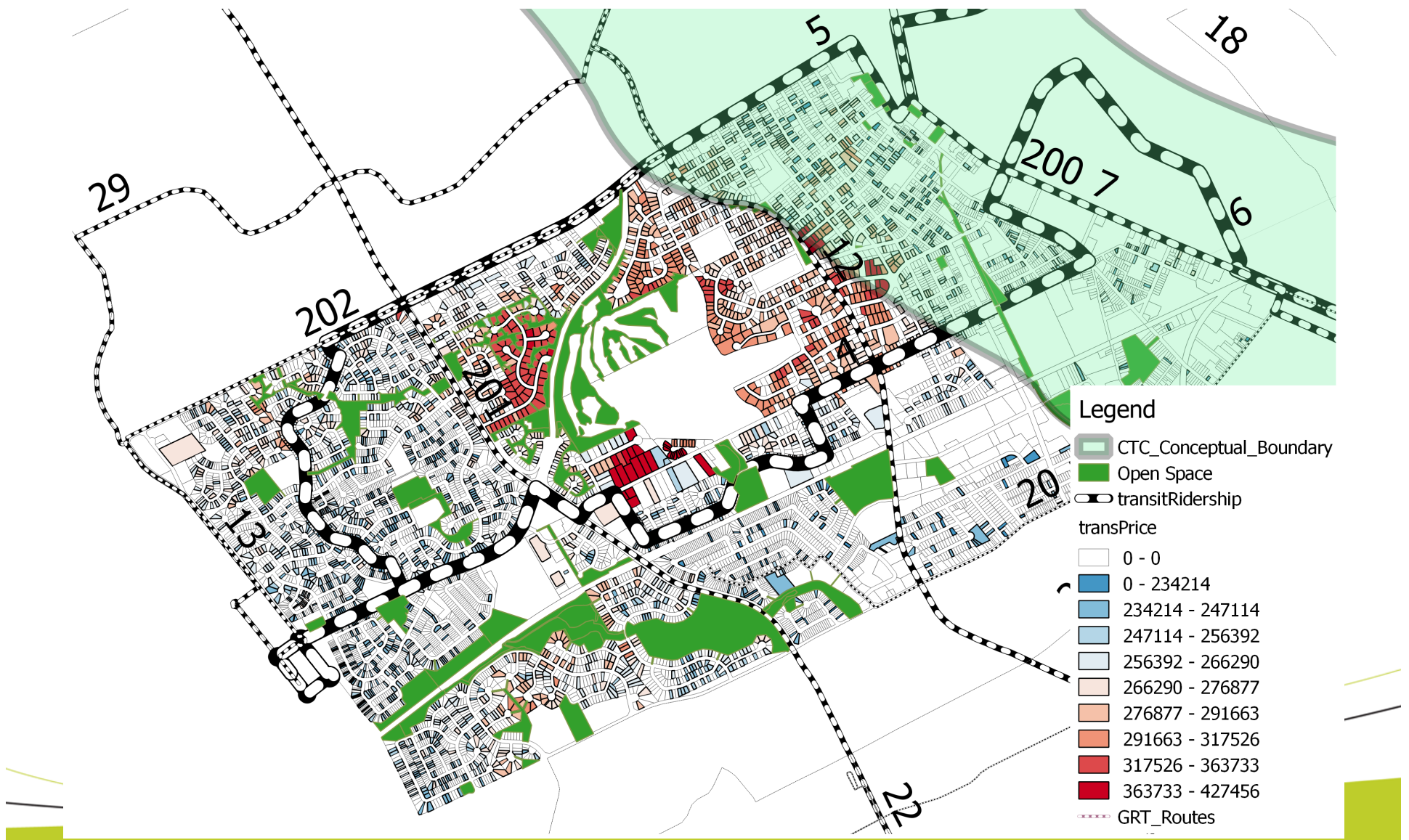
# Prototype modelling area



# Prototype screenshot (enlarged)



# Very preliminary results



## Next steps—household surveys



- ✓ *Interpreting the Housing Market Dynamics in Kitchener-Waterloo from Individual Behaviours*

# Survey Research Questions (Xinyue Pi: rental, Yu Huang: buyer/seller)

- What is the relationship between different households' resources, values, structure and transportation needs, and urban residential patterns?
- How might light rail transit affect housing and rental markets?

# Survey Structures

## Homebuyers and sellers survey

## Renters survey

- I. Residential and neighbourhood characteristics
- II. Home selling/buying experience
- III. Location choice preferences
- IV. Preferences towards LRT
- V. Household characteristics and travel behaviour

- I. Residential and neighbourhood characteristics
- II. Rental experience
- III. Location choice preferences
- IV. Preferences towards LRT
- V. Household characteristics and travel behaviour



# Approach 2: Incorporating household characteristics in regression

## Rental hedonic Model Result (n=~150) Household characteristics as ind. vars.

Category	Significant variables	Effect per unit increase	Level of significance
<b>Household variables</b>	Student household	10.34%	**
	Household with children	-9.12%	*
	One-person household	-8.53%	**
	Household income (per \$1,000)	0.12%	***
<b>Structural variables</b>	Number of bathrooms	18.02%	***
	Number of bedrooms	15.02%	***
	High-rise apartment	7.83%	*
	Low-rise apartment	-8.39%	*
<b>Neighbourhood variable</b>	In CTC	7.48%	**
<b>Behavioural variable</b>	Renting a room	12.04%	*
<b>R-squared</b>	<b>0.85</b>		



# Approach 3: Demand analysis (Two-stage regressions)

## Situating demand analysis

1. **Alonso (1964)** proposed the bid-rent theory, and pointed out that housing prices and location choices are simultaneously determined by a **bidding process**
2. **Rosen (1974)**'s first-stage hedonic regression tells nothing about demand heterogeneity; **Second-stage** hedonic (basically demand analysis) has endogeneity problem
3. **Demand analysis matters for assessing policy/environmental changes, say the LRT implementation**

# Housing Survey Summary

- ✓ **Survey target:** Home Buyers and Sellers from 06/2015 - 04/2017
- ✓ **Survey mails out:** 5000 addresses obtained from Canada Post
- ✓ **Survey responses:**

Responses		Total	
Buyers only	269	Total buyers	357
Sellers only	61	Total sellers	149
Both buyers and sellers	88	Response rate	10%

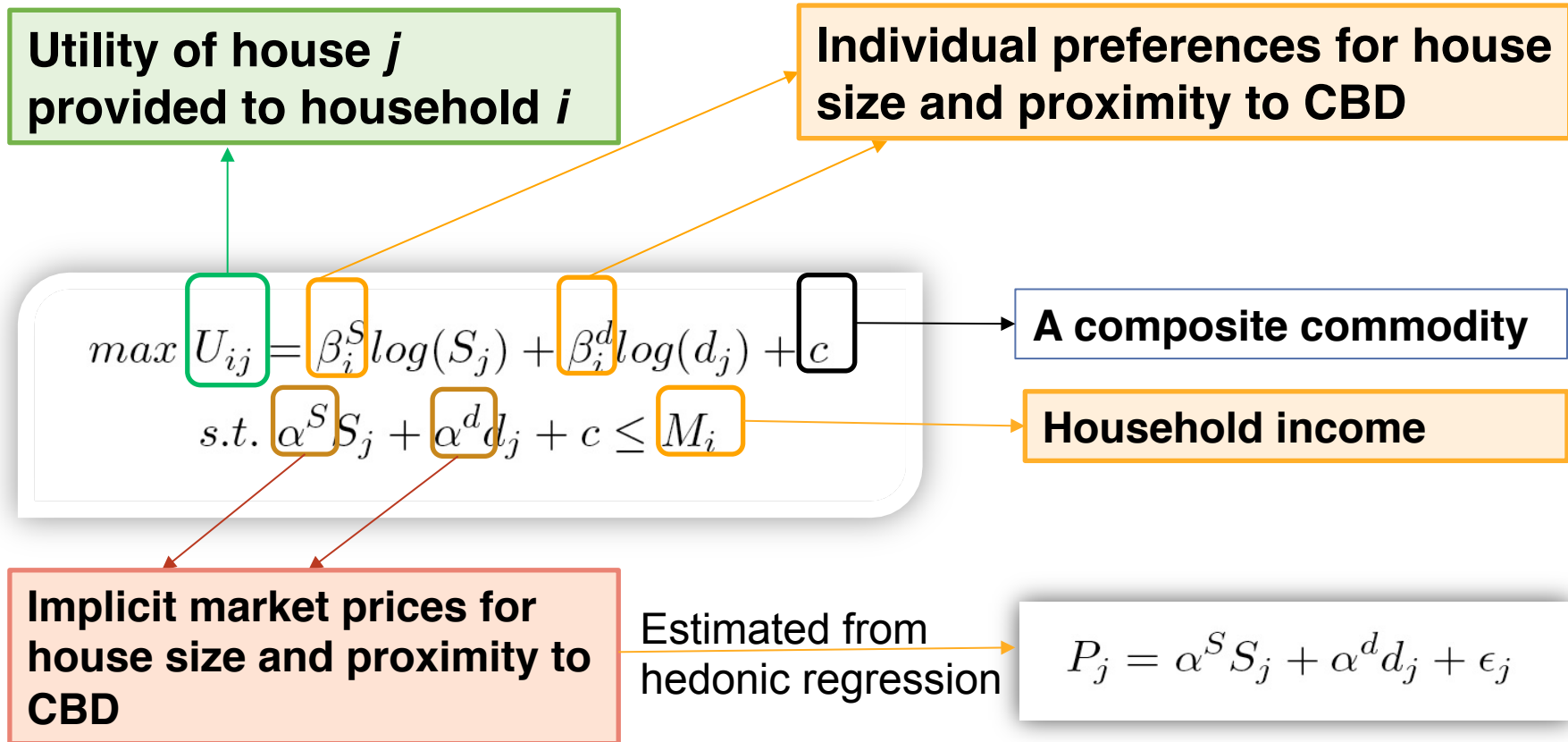
## 2. Housing Demand Analysis - theoretical foundations

- Traditional location choice problem – budget constraint, utility maximization (Alonso, 1964)
- Suppose only two characteristics - **house size ( $S_j$ )** and **proximity to CBD ( $d_j$ )** compose the house  $j$ , the optimization problem can be formulated based on the theory.

## Briefly, three estimation steps:

<b>(Bajari &amp; Kahn, 2005)</b>	<b>Step 1</b>	Estimate implicit prices by hedonic ( $\alpha$ )
	<b>Step 2</b>	Calculate expenditures on each characteristic ( $\beta$ )
	<b>Step 3</b>	Regress the expenditures ( $\beta$ ) on demographics

# The optimization problem based on Alonso bid-rent theory:



**Solving the optimization problem, we derive,**

$$\frac{\frac{\partial U_{ij}(S_j)}{\partial S_j}}{\frac{\partial U_{ij}(c)}{\partial c}} = \alpha^S$$

$$\beta_i^S = S_{j^*} \alpha^S$$

$$\frac{\frac{\partial U_{ij}(d_j)}{\partial d_j}}{\frac{\partial U_{ij}(c)}{\partial c}} = \alpha^d$$

$$\beta_i^d = d_{j^*} \alpha^d$$

- A way to “recover” household-level preference parameters in the utility function with a strong theoretical foundation

# Regress the expenditure on demographics to recover heterogeneous housing demand...

- Assume that households with similar demographic characteristics have similar preferences.

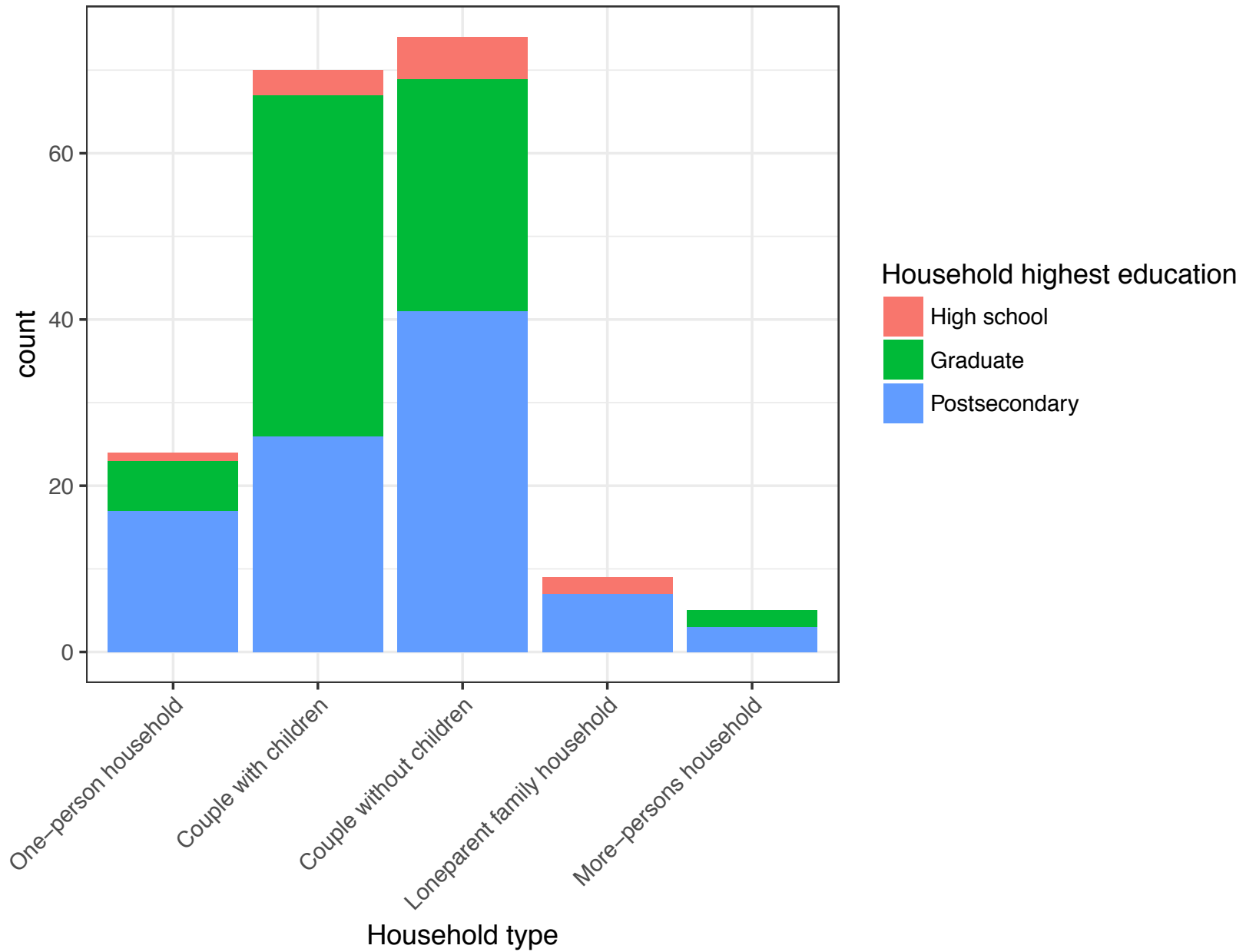
$$\beta_i^S = f(\mathbf{Dem}_i)^S + \eta_i^S$$
$$\beta_i^d = f(\mathbf{Dem}_i)^d + \eta_i^d$$

**A vector of demographic characteristics collected from housing survey:**

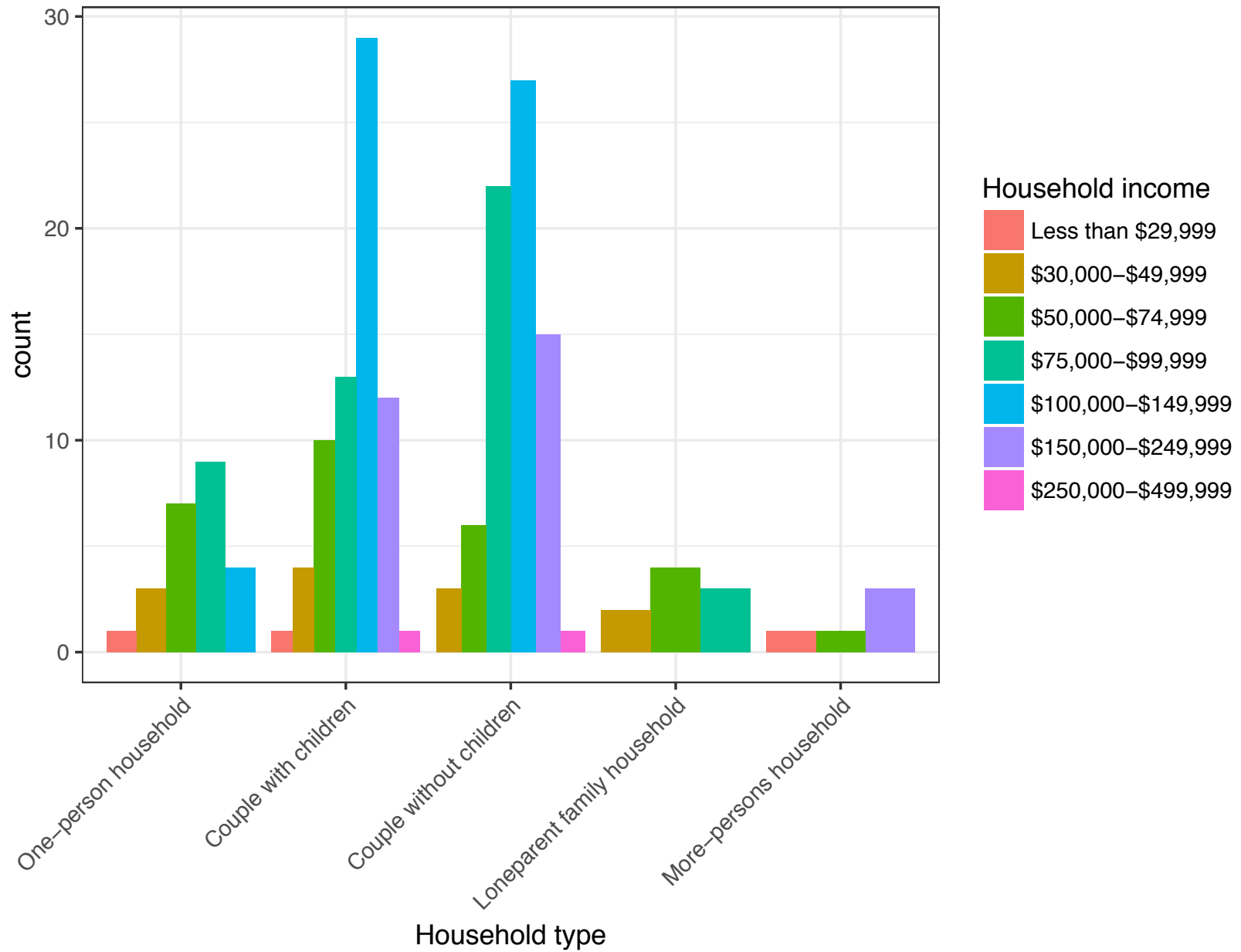
- Highest employment status
- Highest education level
- Household income
- Age of head
- Household type:
  - Couple with children
  - Couple without children
  - Lone parent
  - More persons
  - One person



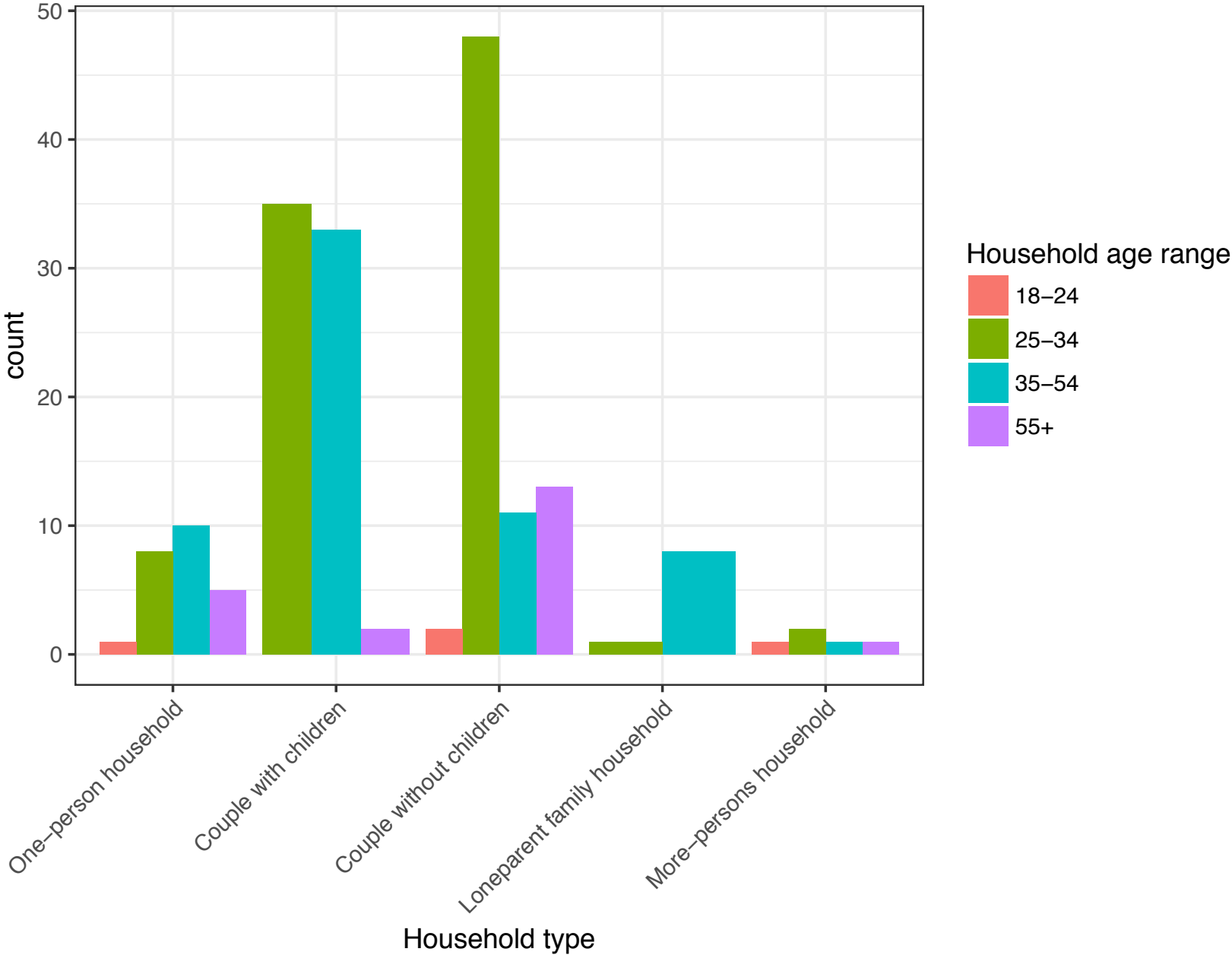
# Household types by household highest education



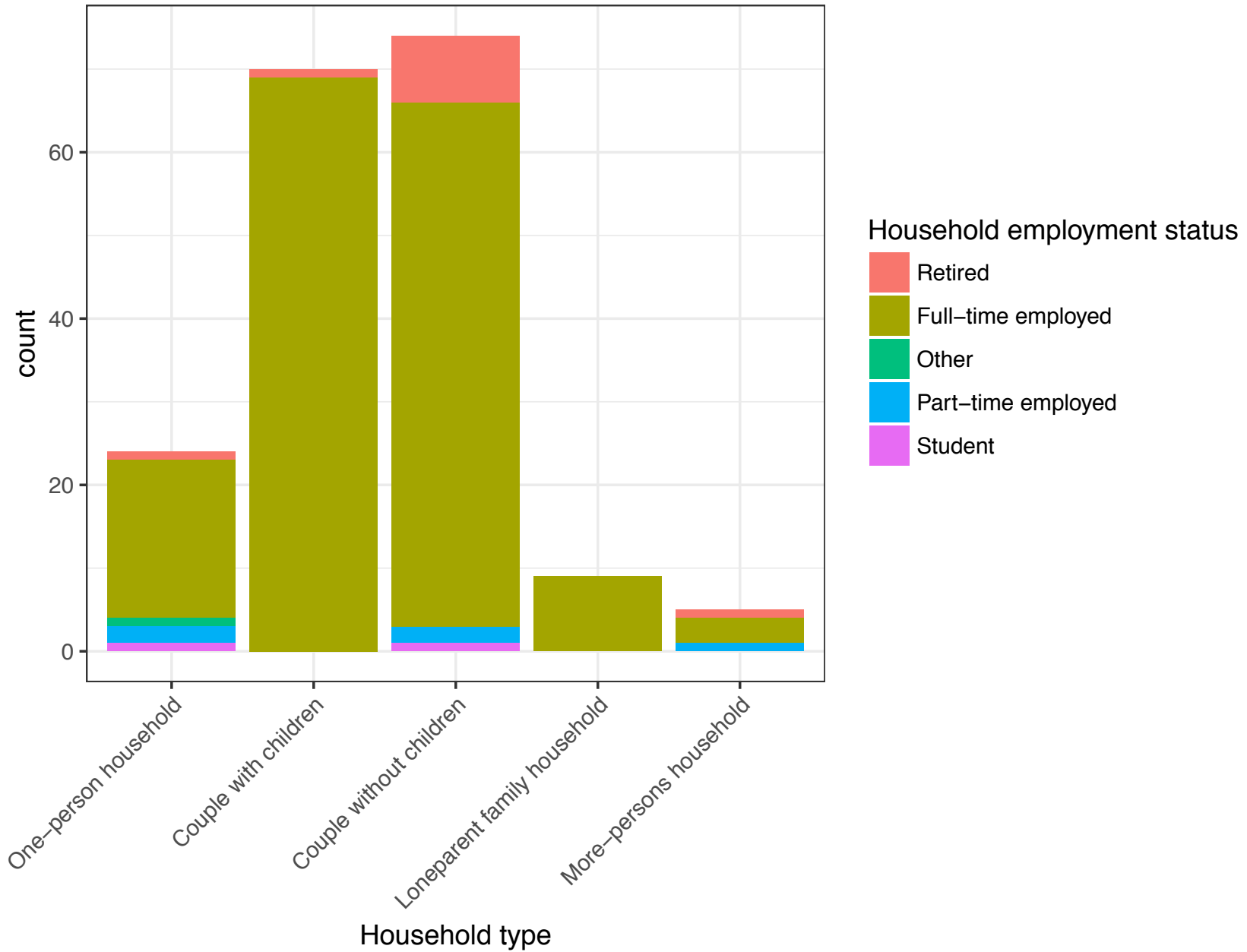
### Household types by household income



Household types by household age range



# Household types by household highest employment status



### First-stage hedonic regression results

<i>Dependent variable:</i>	
log(P)	
NOT-SINGLE HOUSE	0.094 (0.073)
SINGLE-DETACHED HOUSE	0.282*** (0.077)
BDMS	0.077*** (0.019)
FBTH	0.123*** (0.021)
HBTH	0.057** (0.023)
GRAG	0.095*** (0.023)
OPARK	0.009 (0.008)
BUL_AGE	-0.001 (0.001)
POP_DENS	-0.00000 (0.00001)
OS_ACES	0.0001 (0.001)
OS_ADJ	0.004 (0.034)
REG_RD_ADJ	-0.092** (0.046)
DIS_LRT	-0.00000 (0.00001)
DIS_BUS	0.0001** (0.00005)
POST_EDU	0.006*** (0.001)
EMPL_RATE	-0.004** (0.002)
Constant	11.864*** (0.164)
Observations	
	277
R <sup>2</sup>	
	0.666
Adjusted R <sup>2</sup>	
	0.646
Residual Std. Error	
	0.188 (df = 260)
F Statistic	
	32.473*** (df = 16; 260)

*Note:*

\*p<0.1 \*\*p<0.05 \*\*\*p<0.01

## Key points from First-stage hedonic

- Proximity to LRT – amenity (not statistically significant)
- Proximity to bus stops – disamenity
- Adjacency to regional roads – disamenity
- Open space – amenity (not statistically significant)
  
- Neighbourhood post-secondary education rate – amenity
- Neighbourhood employment rate - disamenity

### Preference regression results

	<i>Dependent variable:</i>	
	Preference for bedroom	Preference for full bathroom
	<i>OLS</i> (1)	<i>OLS</i> (2)
Couple with children	25,368.060***	17,754.490**
Lone-parent family	1,252.245	-28,910.870
More-persons household	-6,030.757	-26,173.180
One-person household	-9,142.029	-15,225.960
Less than 29,999	-32,895.700	-27,346.910
30,000-49,999	-24,198.010**	-22,281.470
50,000-74,999	-21,814.490**	-25,907.350**
75,000-99,999	-4,828.313	-6,993.440
150,000-249,999	18,065.060**	17,090.880*
250,000-499,999	105,039.100***	143,993.700***
Age	284.995	1,736.466***
Other employment	20,928.760	36,185.340*
Part-time employed	10,609.170	-2,084.516
Retired	-6,003.822	-35,255.460*
Graduate	7,102.596	7,067.105
High school	-17,961.080	-25,241.590*
Constant	81,174.910***	30,781.080**
Observations	279	279
R <sup>2</sup>	0.325	0.392
Adjusted R <sup>2</sup>	0.284	0.355
Residual Std. Error (df = 262)	46,221.790	53,386.050
F Statistic (df = 16; 262)	7.885***	10.573***

*Note:*

\*p<0.1 \*\*p<0.05 \*\*\*p<0.01

# Key points from preference regression (1)

- Couple with children households prefer most for bedrooms, and full-baths
- Older households prefer more for full-bathrooms, but not for bedrooms
- Retired households prefer less for full-baths, compared to the full-time employed households
- Preferences for bedrooms and full-baths increase with household income



**Preference regression results**

	<i>Dependent variable:</i>	
	Preference for half-bathroom	Preference for garage
	<i>OLS</i> (1)	<i>OLS</i> (2)
Couple with children	7,657.316***	12,698.630***
Lone-parent family	10,449.630**	2,919.234
More-persons household	-3,283.114	-16,294.420
One-person household	-1,529.399	-6,109.171
Less than 29,999	2,402.076	-8,364.114
30,000-49,999	-6,939.626*	-20,382.580**
50,000-74,999	-7,320.011**	-18,635.680***
75,000-99,999	-1,580.410	-6,665.410
150,000-249,999	8,509.374***	15,892.420***
250,000-499,999	20,991.050***	65,047.020***
Age	-34.313	581.163**
Other employment	522.853	17,882.950
Part-time employed	3,793.232	-4,748.108
Retired	-2,476.225	-7,735.275
Graduate	827.396	6,542.320
High school	-4,143.095	-8,506.262
Constant	17,056.950***	20,814.630**
Observations	279	279
R <sup>2</sup>	0.241	0.323
Adjusted R <sup>2</sup>	0.194	0.282
Residual Std. Error (df = 262)	14,891.370	32,089.010
F Statistic (df = 16; 262)	5.194***	7.825***

*Note:*

\*p<0.1 \*\*p<0.05 \*\*\*p<0.01|

## Key points from preference regression (2)

- Households with children prefer more for half-baths
- Couple with children households prefer most for garage
- Older households prefer more for garages
- Preferences for half-baths and garages both increase with household income

### Preference regression results

	<i>Dependent variable:</i>	
	Preference for proximity to LRT stops	Preference for distance to bus stops
	<i>OLS</i>	<i>OLS</i>
	(1)	(2)
Couple with children	-258.236***	10,088.520***
Lone-parent family	-204.499	5,432.627
More-persons household	-21.677	134.948
One-person household	-57.885	3,697.882
Less than 29,999	219.750	-8,568.055
30,000-49,999	245.999*	-9,005.569**
50,000-74,999	270.978***	-8,500.903***
75,000-99,999	19.865	-4,486.113
150,000-249,999	-105.442	41.343
250,000-499,999	-1,037.744***	48,293.320***
Age	0.456	-102.882
Other employment	-136.897	-2,042.904
Part-time employed	53.786	3,372.380
Retired	61.092	5,224.278
Graduate	-165.495**	27.230
High school	120.344	-4,938.272
Constant	-701.146***	17,023.810***
Observations	279	279
R <sup>2</sup>	0.287	0.317
Adjusted R <sup>2</sup>	0.243	0.275
Residual Std. Error (df = 262)	484.955	16,895.200
F Statistic (df = 16; 262)	6.589***	7.606***

*Note:*

\*p<0.1 \*\*p<0.05 \*\*\*p<0.01

## Key points from preference regression (3)

- Couple without children households prefer most for proximity to LRT stops, i.e., prefer living close to LRT stops
- Couple with children households prefer most for distance to bus stops, i.e., prefer living far from bus stops
- High income households prefer living far from LRT stops and bus stops
- Households with graduate degree prefer to live far from LRT stops

**Preference regression results**

	<i>Dependent variable:</i>	
	Preference for neighbourhood post- secondary education rate	Preference for neighbourhood employment rate
	<i>OLS</i> (1)	<i>OLS</i> (2)
Couple with children	27,007.750***	-17,348.310***
Lone-parent family	10,938.270	-9,950.384
More-persons household	-38,490.620	12,387.870
One-person household	-11,766.720	3,890.155
Less than 29,999	-28,647.960	24,675.380
30,000-49,999	-28,568.440*	16,589.090*
50,000-74,999	-41,262.140***	23,224.040***
75,000-99,999	-7,495.329	519.078
150,000-249,999	25,289.010**	-15,569.990**
250,000-499,999	185,973.900***	-101,294.300***
Age	1,230.977***	-268.462
Other employment	43,453.420**	-25,950.280**
Part-time employed	24,695.600	-16,894.930
Retired	-9,829.160	-1,891.157
Graduate	8,035.471	-5,838.001
High school	-38,440.200**	19,561.760**
Constant	105,677.500***	-88,175.710***
Observations	267	267
R <sup>2</sup>	0.403	0.366
Adjusted R <sup>2</sup>	0.365	0.326
Residual Std. Error (df = 250)	58,846.210	33,941.500
F Statistic (df = 16; 250)	10.569***	9.024***

Note:

\*p<0.1 \*\*p<0.05 \*\*\*p<0.01

## Key points from preference regression (4)

- Couple with children households prefer most for neighbourhood education rate; Couple without children prefer more for neighbourhood employment rate
- Higher income and older households prefer more for neighbourhood education rate
- Lower income households prefer more for neighbourhood employment rate
- Households with high-school education prefer less for neighbourhood education rate, but prefer more for neighbourhood employment rate

# Contributions of qualitative research

Jinny Tran (developers)  
Justin Cook and Jennifer Dean (Realtors)

- Context
- Cross-validation
- Future directions



# Factors that affect developers' decision making (Jinny Tran)

- **Physical** (e.g. land availability, environmental conditions)
- **Spatial** (e.g. proximity to transit, to employment centres, to commercial areas)
- **Socio-Economic** (e.g. market demand, growth potential)
- **Planning** (e.g. approval costs, timing of approval)
- In theory, developers work to maximize profit, while minimizing risk and uncertainty



**THE TANNERY DISTRICT**  
KITCHENER – MIXED USE CONVERSION  
(FORMER FACTORY)



Jinny Tran



# Developer Survey Highlights

- Conducting surveys with 17 residential developers
- Fairly wide distribution of specializations and built form found; shift towards intensified and mixed use forms-but segmented target markets
- Few developers consider what others are doing when making plans
- Response to LRT generally positive, but more so for infill developers than the other two—some “wait and see” expressed



**SEAGRAM  
LOFTS**  
WATERLOO –  
RESIDENTIAL  
CONVERSION



Jinny Tran

# Realtor interviews/Focus Groups

## Qualitative Approach

- Deeper understanding of why people are buying in the CTC
- Complementing quantitative research to draw stronger conclusions

## Why Realtors?

- Key informants with specialized knowledge
- Emotional/cultural interpreters

# Key Discussion Points

**Three broad themes emerged from discussions:**

1. CTC development and investment
2. Resident perception of attractiveness of CTC
3. CTC creating connections within region and beyond

# Findings: 1. CTC Development and Investment

## Encouraging Investment in Real Estate

- Understood as stimulating land value uplift
- Investors primarily from within the Region and GTA
- CTC Investment potential more desirable than long term residence

# Findings: 1. CTC Development and Investment

## “Tech Hub” Development

- Key piece of infrastructure supporting growth
- Connecting residents with emerging employment trends

# Findings: 1. CTC Development and Investment

## Regional Image

- Signifier of Region's status as “the Silicon Valley of the north”
- Symbol of the Region being “world class”
  - Allowed for comparison with many other international centers

## Findings: 1. CTC Development and Investment

*“We’re seeing investment, local people that are buying in uptown, or downtown **just for investment purposes**. I think the families, the 30 plus demographic, that are now looking for more investment opportunities, they realize [the CTC] is something they can grasp and they realize that’s **an up and coming area**.”*

# Findings: 2. Resident Perceptions

## Lifestyle Choice

- More attractive to new residents than long term
- CTC is attractive for relative affordability of services and amenities
- Reflected the services and amenities available in other cities



# Findings: 2. Resident Perceptions

## Aging Populations

- View the CTC as desirable due to amenities
- Lack of affordable/appropriately sized options preventing downsizing

# Findings: 2. Resident Perceptions

## Long Term Residents

- Viewed more favourably as construction nears completion
- Few long-term residents show interest in using it
- Compared to Conestoga Parkway (freeway controversial when built) as likely to be more appreciated/used over time

## Findings: 2. Resident Perceptions

*“Even some of the **older demographics**, I think they **are really looking forward to [the LRT]**. They are definitely buying to be close to it, not right on it but somewhat close to it, within a block or two. So it will be really good. I think it will impact [the Region] in a positive way.”*

# Findings: 3. Creating Connections

## Connecting the Region

- Bringing Kitchener and Waterloo together as a seamless urban environment
- Extension to Cambridge will bring the Region together as a unified whole

# Findings: 3. Creating Connections

## Connecting Southern Ontario

- Seen as a localized connection to Toronto and other near by municipalities
- Increased connectivity with GO/high-speed rail essential next step

## Findings: 3. Creating Connections

*“In a real estate perspective, all the **condos**, the **Google** building... the **Zehr** group building; those are only there **because of the LRT**. They're looking at it as it's not just a north and south train, **it's connection** to Barrie, Hamilton, Niagara. All these places are going to have LRT that lead **to these fast trains** that all spine into Toronto. That's what [people are] investing on.”*

# Implications for modelling

- Clear supply constraints
  - Lack of supply for families in the central transit corridor
  - Lack of strategic behaviour likely to lead again to oversupply dynamics
  - Actual demand seems poorly understood/anticipated
- Clear evidence of market segmentation
  - “Urban lifestylers” create demand for core properties
  - Locals more likely to see suburban properties
  - Future regression/modelling will respond to this new information

# Relative to other studies, this study ...

- 1) builds on richer, more detailed data through a comprehensive housing survey and realtor interview
- 2) examines the housing market dynamics from individual behaviours
- 3) allows a strong direct connection between our implemented model and Alonzo's classic bid-rent theory models by
  - parameterising the utility function for empirical housing study with strong theoretical foundations
  - recovering heterogeneous housing demand by combining survey data and theoretical methods within 3-Steps
  - explaining varying preferences among heterogeneous households and thus provides more information than a traditional first-stage hedonic model



# Future work ...

- **1) Improve current model by**
  - using the stated preferences from survey to validate our proposed model
  - building a multi-level hedonic regression with potential more data source
  - using probit models to estimate heterogeneous demand for dichotomous characteristics, such as In CTC, or Large Yard, Single detached house.
- 2) Estimate **heterogeneous household *WTP*** for each house given their demographics
- 3) Simulate housing location choices in our **Agent-Based land market model**: by adding more theoretically-grounded and empirically-validated behaviour rules (especially, utility parameterization and *WTP* estimation from this study)
- 4) Model and better interpret the housing market dynamics

		Estimation steps	Details
Our proposed demand estimation method	(Bajari & Kahn, 2005)	Step 1	Estimate implicit prices by hedonic ( $\alpha$ )
		Step 2	Calculate expenditures on each characteristic ( $\beta$ )
		Step 3	Regress the expenditures ( $\beta$ ) on demographics
	Step 4		Estimate the demand curve for each characteristic
	Step 5		Estimate <i>WTP</i> for each characteristic
	Step 6		Estimate the total <i>WTP</i> for each house

***WTP estimation***

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## Collaborating partners

- Region of Waterloo
- Cities of Waterloo and Kitchener
- Kitchener-Waterloo Association of Realtors,
- Coldwell Banker Peter Benninger Realty

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- *Estimating Homebuyer Preferences Under Intensification: Hedonic Modelling of Open Space and Multimodal Transit Amenities Preceding Light Rail in Kitchener-Waterloo* (Babin, 2016) <http://hdl.handle.net/10012/10936>
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# Current vs. Ideal Housing Types

- Most preferred rental housing type: **single-detached house**

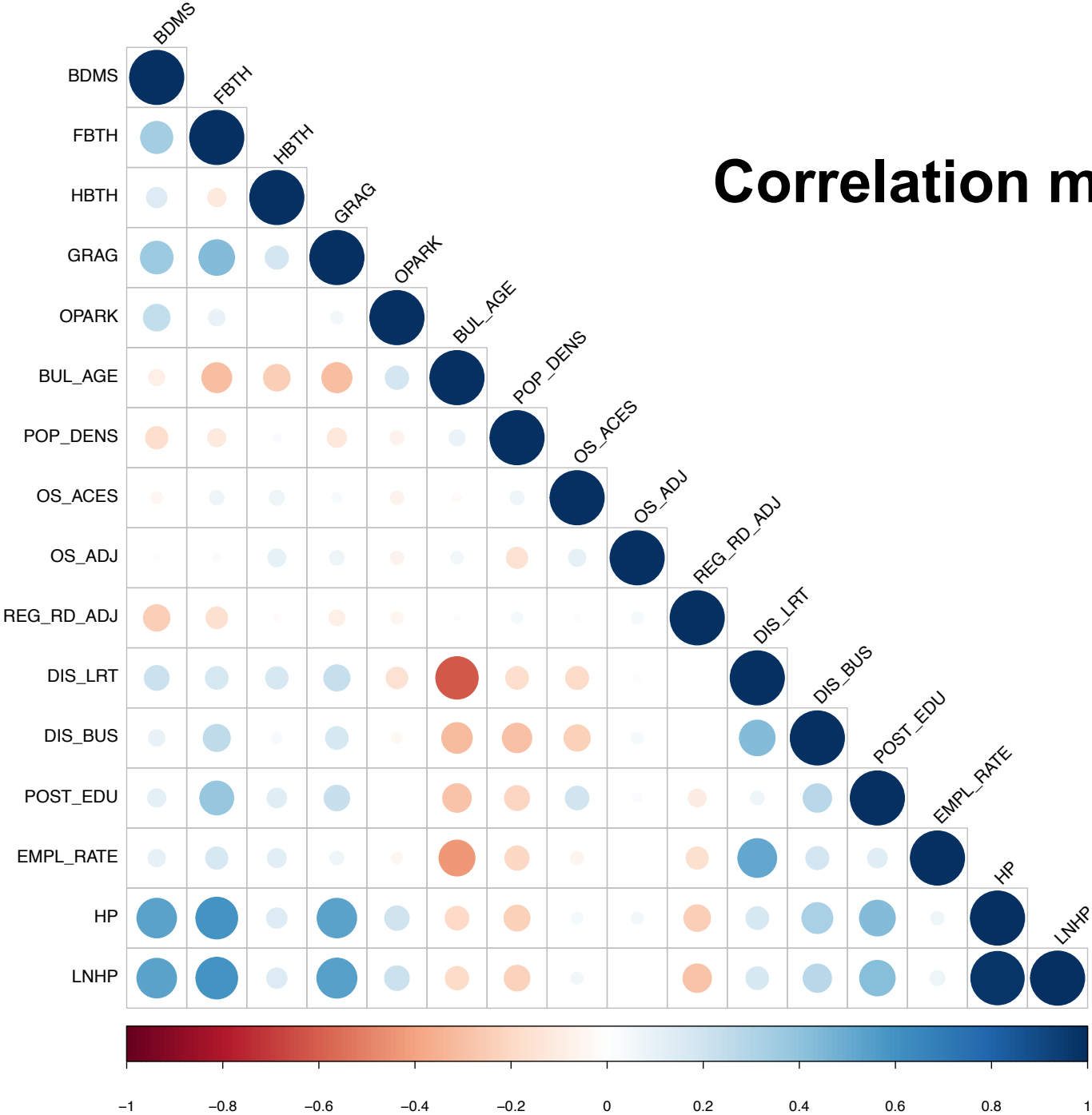
Current	Count and % of total responses	Ideal matches Current	Ideal matches most popular other than current
Single-detached house	N=35, 12%	71% still prefer single	24% prefer apartment
Semi-detached house	N=10, 3%	40% still prefer semi	40% prefer single
Row house	N=28, 10%	38% still prefer row house	41% prefer single
Apartment(<5 storeys)	N=85, 30%	36% still prefer apartment(<5 storeys)	35% prefer single
Apartment(>=5 storeys)	N=114, 40%	56% still prefer apartment(>=5	19% prefer single



# Descriptive Statistics of Selected Variables

Descriptive Statistics					
Statistic	N	Mean	St. Dev.	Min	Max
BDMS	339	<b>3.20</b>	0.80	1	8
FBTH	339	<b>1.87</b>	0.73	1	4
HBTH	340	0.78	0.57	0	3
GRAG	340	<b>1.14</b>	0.65	0	4
OPARK	340	1.80	1.59	0	10
BUL_AGE	297	<b>30.32</b>	22.05	0	118
POP_DENS	327	2,961.55	2,106.47	38.80	15,811.80
OS_ACES	340	42.76	17.84	8.66	103.77
OS_ADJ	340	0.16	0.37	0	1
REG_RD_ADJ	340	0.09	0.29	0	1
DIS_LRT	340	<b>3,605.16</b>	1,636.60	227.91	7,509.33
DIS_BUS	340	<b>347.84</b>	310.65	10.00	1,600.00
POST_EDU	327	62.35	9.52	34.75	87.70
EMPL_RATE	327	65.31	8.88	25.00	83.00
HP	327	404,046.40	143,633.20	135,000	975,000
LNHP	327	12.85	0.34	11.81	13.79

# Correlation matrix





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