

# Minimum Lot Size and Segregation in Connecticut

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Camp Resources XX

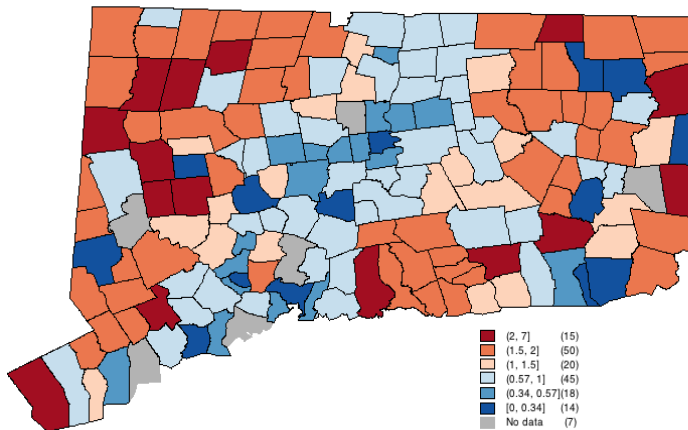
## History of zoning policies

- Rapidly developed in the 1910-1930
- Development of the interstate highway system in the 1960s made people more mobile
- Zoning policies were then widely adopted in order to prevent undesired development and became more exclusionary in the 1970s

## Policy of minimum lot size

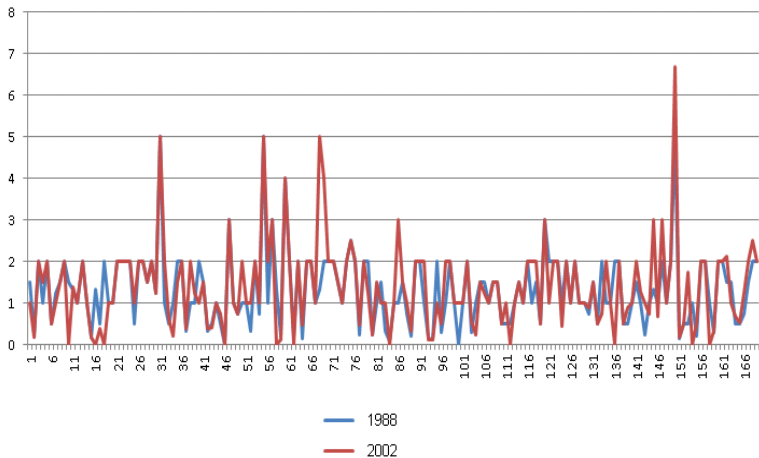
- Help minimize the negative externalities of groundwater use. In the United States, over 15 million households depend on private wells for drinking water supply.
- Improve amenities, and believed to preserve home values
- Reduce negative tax externalities
- Distort free market
- Sorting out 'less desirable' people including the poor or ethnic minorities

## Minimum Lot Size



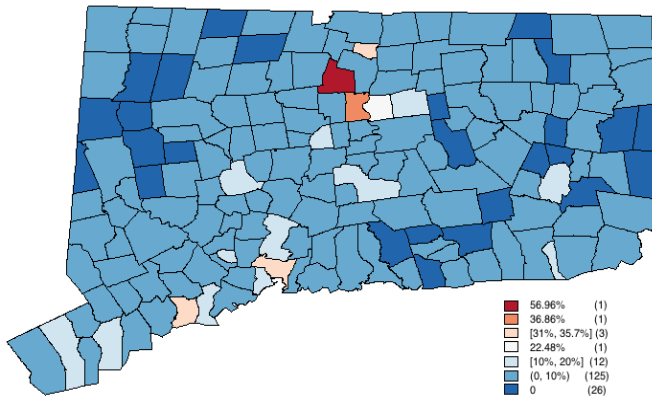
Washington in Litchfield county has minimum lot size 6.67 acres.

## Minimum Lot Size



## Status of Racial Segregation in Connecticut

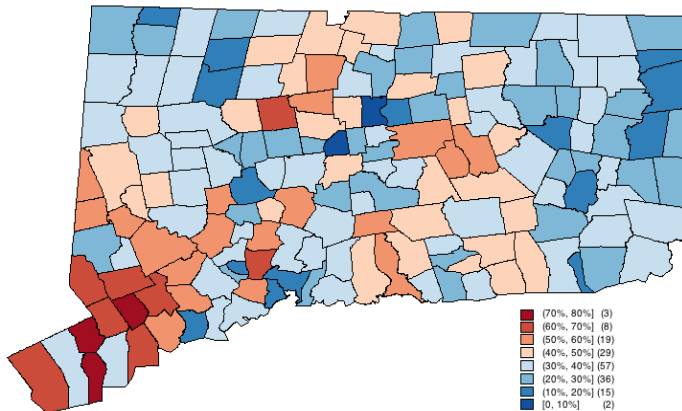
Percentage of Black or African American alone



Note: first 10 towns have 64% population of black or African American alone. The highest town is Bloomfield in Hartford county, with 57% black population

## Status of Income Segregation in Connecticut

Percentage of household income > \$100,000



Note: the first three towns are Western, Darien and New Canaan, all in Fairfield county, with more than 70% of household income great than \$100,000.

## Literature background

- Zoning could cause higher housing prices (Glaeser and Gyourko, 2003)
- Housing Demand and Racial Segregation (Bajari and Kahn, 2005)
- Racial Segregation with BLP (Bayer et al., 2004)
- Impact of Minimum Lot Size on housing price (Dalton and Zabel, 2011)



## Research outline

- Objective: to find out if minimum lot size policy contributes to racial or income segregation
- Methodology: to isolate impact of policies in a housing demand model

# Model

$$V_i = V_i(h, c) \quad (1)$$

s.t.

$$price_h + c \leq Y_i \quad (2)$$

$$V_{khl} = \sum_{m=0}^M \alpha_m^k X_{hm} + \sum_{n=0}^N \beta_n^k Z_{ln} - price_h + \eta_{kl} + \epsilon_{hkl} \quad (3)$$

- k: type of buyer by race and income
- h: house
- X: house characteristics including number of room, bathroom, bedroom, built year, interior size, lotsize
- Z: neighborhood characteristics including minimum lot size, percentage of type k, travel time to work, town average expenditure

- More specific model

$$V_{khl} = \alpha_1^k intersize_h + \alpha_2^k lotsize_h + \alpha_3^k builtyears_h + \delta_{kl} - price_h + \epsilon_{hkl} \quad (4)$$

in which:

$$\delta_{kl} = \sum_{n=0}^N \beta_n^k Z_{ln} + \eta_{kl} \quad (5)$$

- Two-step estimator
  - First-step: Generalized logit model with fixed effect and contraction mapping
  - Second-step: Decompose  $\delta_{kl}$

# First-step estimation

- Equilibrium condition

$$V_{khl} > V_{whl} \quad \forall w \neq k \quad (6)$$

- MLE

$$\max L = \sum_l \sum_h \sum_k \ln(\text{Prob}_{khl}) I_{khl} \quad (7)$$

- Market clear

$$\frac{\partial L}{\partial \delta_{kl}} = N_{kl} - \sum_h \text{Prob}_{khl} = 0 \quad \forall l \quad (8)$$

- Contraction mapping

$$\delta_{kl}^{t+1} = \delta_{kl}^t - \ln\left(\frac{\sum_h \text{Prob}_{khl}}{N_{kl}}\right) \quad (9)$$

## Second-step estimation

$$\delta_{kl} = \sum_{n=0}^N \beta_n^k Z_{ln} + \eta_{kl} \quad (10)$$

- Similar to minimum wage problem?
- Problem of truncated data or censored data?
- Problem of bias selection and need Heckman correction?
- Isolation of policy impact of minimum lot size

$$X_h = MLS_l + \xi_h \quad (11)$$

# Main Dataset

- Housing transaction data in Connecticut collected by DataQuick
- Loan application registry information under the Home Mortgage Disclosure Act (HMDA)
- Survey data of minimum lot size regulations for 162 Connecticut towns
- Fiscal Indicators by Office of Policy and Management of Connecticut
- American Community Survey by U.S. Census Bureau

## Summary statistics

Variable	Obs.	mean	Std. Dev	Min	Max
nroom	11269	6.77	1.70	2	18
nbath	11269	1.90	0.81	0.5	7
nbed	11269	3.18	0.88	0	9
intersize	11269	1620.94	1132.25	0	67410
lotsize	11269	36928.95	57888.48	1	1383466
age	11254	47.83	33.29	0	309
income (thousand \$)	10857	114.64	170.88	5	7500
min lot size	162	1.46	1.01	0	6.67
expenditure (thousand \$)	169	2949.69	600.71	1524.34	6209.00
travel	169	12.57	2.24	7.59	18.53

Table: Type definition of buyers

Type	Race	Household income	Obs.
1	White	[0, 59,999)	2545
2	White	(60,000, 99,999]	2839
3	White	[100,000, $\infty$ )	2854
4	Black	[0, 59,999)	243
5	Black	(60,000, 99,999]	150
6	Black	[100,000, $\infty$ )	47
7	Others	[0, 59,999)	813
8	Others	(60,000, 99,999]	812
9	Others	[100,000, $\infty$ )	966



# First-Step results

Table: Model 1

Type	Var 1	Estimate	ProbChiSq	Var 2	Estimate	ProbChiSq	Var 3	Estimate	ProbChiSq
1	interior size	0.000173	<.0001	lot size	2.80E-06	<.0001	built	0.00363	<.0001
2	interior size	0.000436	<.0001	lot size	5.39E-06	<.0001	built	-0.0021	<.0001
3	interior size	0.000955	<.0001	lot size	8.86E-06	<.0001	built	-0.00387	<.0001
5	interior size	0.000467	<.0001	lot size	4.71E-06	<.0001	built	-0.00322	<.0001
6	interior size	0.000286	<.0001	lot size	3.37E-06	<.0001	built	-0.00663	<.0001
7	interior size	0.000305	<.0001	lot size	4.02E-06	<.0001	built	-0.0008	<.0001
8	interior size	0.000549	<.0001	lot size	4.65E-06	<.0001	built	-0.00602	<.0001
9	interior size	0.00109	<.0001	lot size	5.71E-06	<.0001	built	-0.0122	<.0001
other									
price		1							
$\delta_{kl}$		Yes							

## Second-step results: preliminary

Variable	Type	Model 1	<i>Pr</i> >   <i>t</i>	model 2	<i>Pr</i> >   <i>t</i>	Model 3	<i>Pr</i> >   <i>t</i>	Model 4	<i>Pr</i> >   <i>t</i>
MLS	1	0.18	0.5199	1.86***	<.0001	2.37***	<.0001	4.83***	<.0001
MLS	2	0.67***	0.0072	2.84***	<.0001	2.95***	<.0001	10.37***	<.0001
MLS	3	2.26***	<.0001	6.97***	<.0001	3.97***	<.0001	18.80***	<.0001
MLS	4	0	1	0	1	0	1	0	1
MLS	5	0.73	0.6872	3.51	0.2379	6.13*	0.028	11.96***	<.0001
MLS	6	-0.14	0.9701	3.03	0.6136	1.25	0.8248	10.74*	0.0376
MLS	7	0.77	0.1954	3.28***	0.0007	3.72***	<.0001	6.48***	<.0001
MLS	8	1.34***	0.0089	4.42***	<.0001	5.38***	<.0001	11.99***	<.0001
MLS	9	-2.49***	<.0001	0.38	0.6942	0.55	0.551	12.62***	<.0001
travel res		Yes		Yes		Yes		Yes	
expenditure		Yes		Yes		Yes		Yes	
type%		Yes		Yes		Yes		Yes	
obs.		10304		10304		10045		10045	
R-square		0.67		0.73		0.71		0.93	

## Second-step results: preliminary

Type	Percentage of each type								
	1	2	3	4	5	6	7	8	9
1	-7.15***	-25.31***	4.72***	-26.51***	-26.62***	12.65**	-11.09***	-35.22***	69.58***
2	-13.49***	-39.05***	10.4***	-36.17***	-9.21***	-3.48 ***	-17.79***	-8.79***	82.29***
3	-19.11***	-94.81***	16.76***	-78.56***	-29.06***	97.37	-5.96 ***	154.43**	111.90***
4	0	0	0	0	0	0	0	0	0
5	-12.74	-61.51	12.16	-14.26	-131.66	184.84**	-29.52	100.76	-32.80
6	9.57	-70.48	6.3	-68.24	-176.83	582.83**	3.28	-76.13	-5.27
7	-5.2***	-49.99***	2.06***	-38.86***	-41.84***	76.78	-15.81***	3.48	59.83***
8	-9.48	-47.61***	8.32***	-27.61***	-45.99***	87.14	-28.08***	6.78***	94.01***
9	5.17	-98.36***	9.94	-39.18*	-236.49***	274.41***	-16.35	77.89**	-2.91

Note: \*\*\* $p < 0.01$  \*\*  $p < 0.05$  \*  $p < 0.1$

## Conclusion

- The highest demand for policy of minimum lot size is white with high income
- Higher income group attracts people; Higher income black shows very strong signal of attracting his own type; Low income group distracts people and Black low income group distracts more
- **Warning: very preliminary** The minimum lot size contributes 19% of segregation.

## Further work

- More model specifications
- Sensitivity analysis
- Collect more detailed spatial data for minimum lot size
- Collect more detailed time series data for minimum lot size
- Collect more detailed spatial data for neighborhood