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# Decompositions of Spatially Varying Quantile Distribution Estimates:

## The Rise and Fall of Tokyo House Prices

# Outline

- Quantile procedures for estimating house price indices across the full distribution of prices:
  1. Hedonic
  2. Locally weighted versions to allow for smooth variation in appreciation rates over space.
  3. Tokyo & Chicago
  4. Decomposition of the change in distributions over time into the portions due to changes in the coefficients, explanatory variables, and location of the sales. Condo prices in Tokyo, 1986 – 1990 and 1991 – 1995.

# Conventional House Price Indices

- Medians or Means
- “Quality – Controlled” Indices
  - Hedonic model with controls for structural characteristics
  - Repeat Sales
  - Hybrid

All focus on a central tendency – mean or median.

# Quantile Price Indices

- $Q_{lnP}(q|X_{it}, D_{it}) = X_i\beta(q) + \sum_{t=2}^T D_{it}\delta_t(q)$
- $q = .50$  is comparable to hedonic estimation. Also directly comparable to repeat sales estimator if the sample is restricted to properties that have sold at least twice.
- Can trace out the full distribution by estimating across many quantiles.

# Price Index Estimation

- Estimating Equation:  $Q_{lnP}(q|X_{it}, D_{it}) = X_i\beta(q) + \sum_{t=2}^T D_{it}\delta_t(q)$
- Estimates in base period:  
 $X\hat{\beta}(q)$
- Estimates at time t:  
 $X\hat{\beta}(q) + \hat{\delta}_t(q)$
- n x B estimates in each period, where B is the number of quantiles.
- Kernel density function for the nB-vector of estimated prices shows the full distribution of predicted sales prices in time 0 and time t. Set all other time variables to 0 but keep X at actual values.

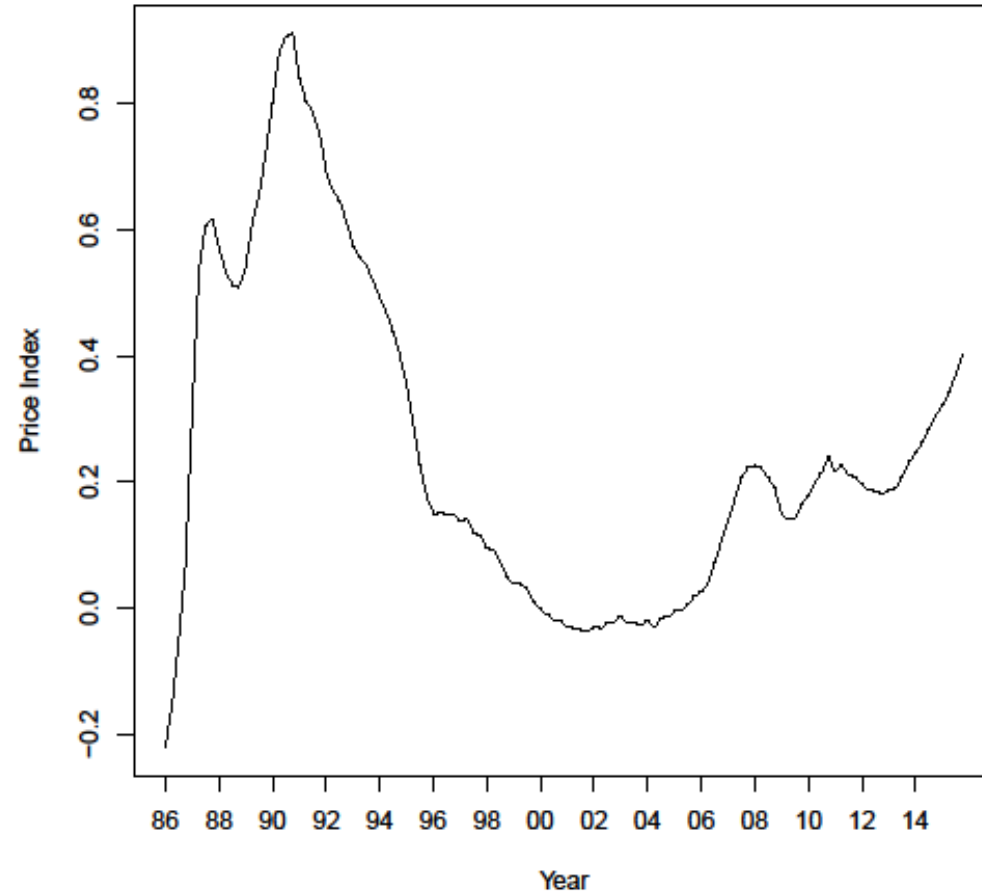
# Other Counter- Factual Distributions

- Example: house age = 25, 75
- $Q_{\ln P}(q|Age_i, X_i, D_{it}) = Age_i \lambda(q) + X_i \beta(q) + \sum_{t=2}^T D_{it} \delta_t(q)$
- Predicted lnP at time t for age = 25, 75:  
$$25 \hat{\lambda}(q) + X_i \hat{\beta}(q) + \hat{\delta}_t(q)$$
$$75 \hat{\lambda}(q) + X_i \hat{\beta}(q) + \hat{\delta}_t(q)$$
- Kernel density function for the nB-vector of estimated prices shows the full distribution of predicted sales prices for age = 25 and age = 75.

# Locally Weighted Estimation

- Coefficients vary over space, e.g.,  $z_1$  = longitude and  $z_2$  = latitude:
- $$Q_{lnP}(q|X_i, D_{it}, z_{1i}, z_{2i}) = X_i\beta(q, z_{1i}, z_{2i}) + \sum_{t=2}^T D_{it}\delta_t(q, z_{1i}, z_{2i})$$
- A weighted version of quantile regression. Weights decline with distance to target points. Can interpolate from target points to full data set.
- With  $k$  variables and  $B$  quantiles, have  $n \times k \times B$  estimated coefficients. But still have  $n \times B$  predicted values. Again can use kernel density functions to summarize the results.

Tokyo:  
226,983  
condo sales,  
1986 - 2014





# Descriptive Statistics, 1986 – 1990, 32,029 sales

Variable	1986 – 1990	1991 – 1995
Price per Square Meter	104.721	88.079
Log Price per Square Meter	4.553	4.408
Building Area (square meters)	50.466	54.249
Log Building Area	3.854	3.928
Age	9.721	12.864
South View	0.209	0.383
1 <sup>st</sup> Floor	0.098	0.097
2 <sup>nd</sup> Floor	0.158	0.157
Floor (range 1 – 25)	4.753	4.766

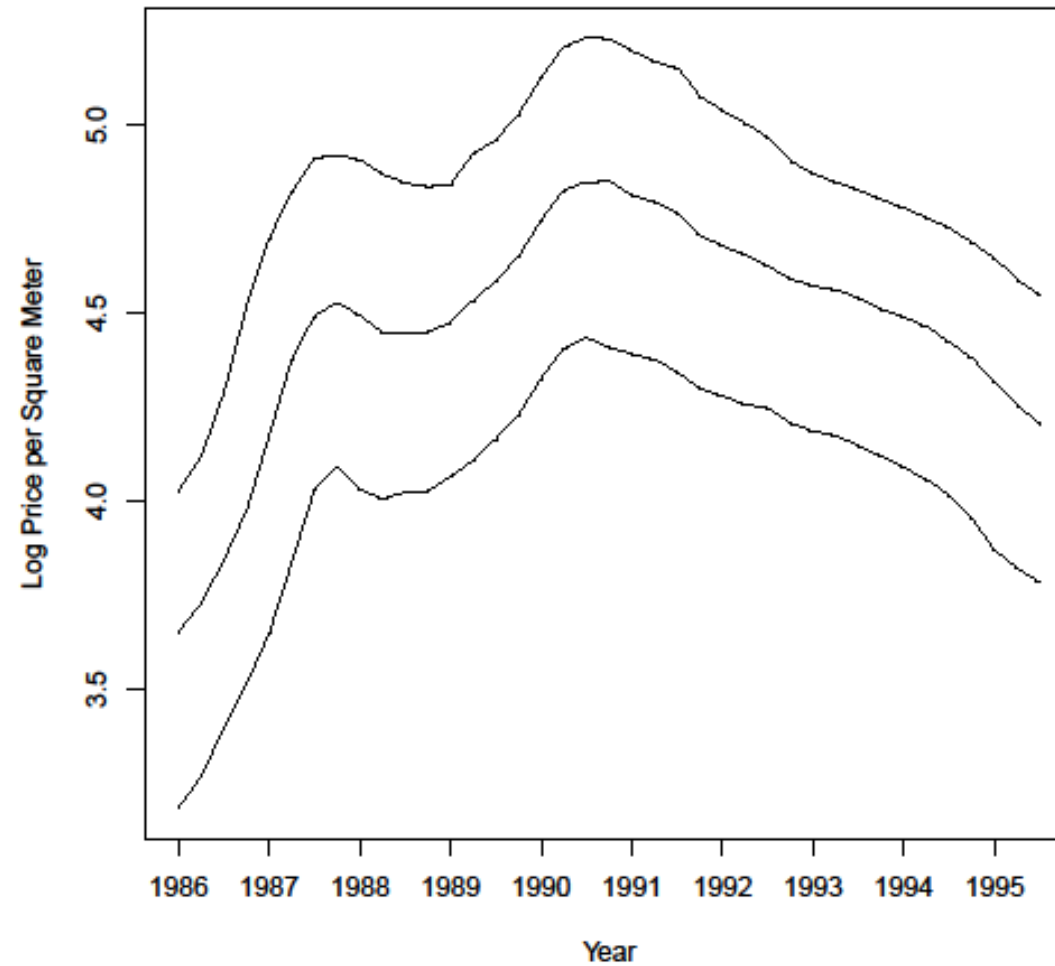
# Estimates for 1986-1990 with controls for location and time

Variable	OLS	0.1	0.5	0.9	Quantile .9 - .1
Log Building Area	-0.0431 (0.0028)	-0.0258 (0.0038)	-0.0519 (0.0022)	-0.0584 (0.0041)	-0.0326 (0.0040)
Age	-0.0223 (0.0002)	-0.0246 (0.0003)	-0.0223 (0.0002)	-0.0218 (0.0003)	0.0028 (0.0004)
South View	0.0048 (0.0025)	0.0056 (0.0040)	0.0088 (0.0023)	-0.0009 (0.0044)	-0.0064 (0.0037)
1 <sup>st</sup> Floor	-0.0140 (0.0035)	-0.0174 (0.0057)	-0.0168 (0.0033)	-0.0116 (0.0063)	0.0058 (0.0085)
2 <sup>nd</sup> Floor	-0.0061 (0.0028)	-0.0040 (0.0046)	-0.0062 (0.0027)	-0.0079 (0.0051)	-0.0039 (0.0070)
Floor	0.0062 (0.0004)	0.0085 (0.0006)	0.0061 (0.0004)	0.0044 (0.0007)	-0.0041 (0.0008)

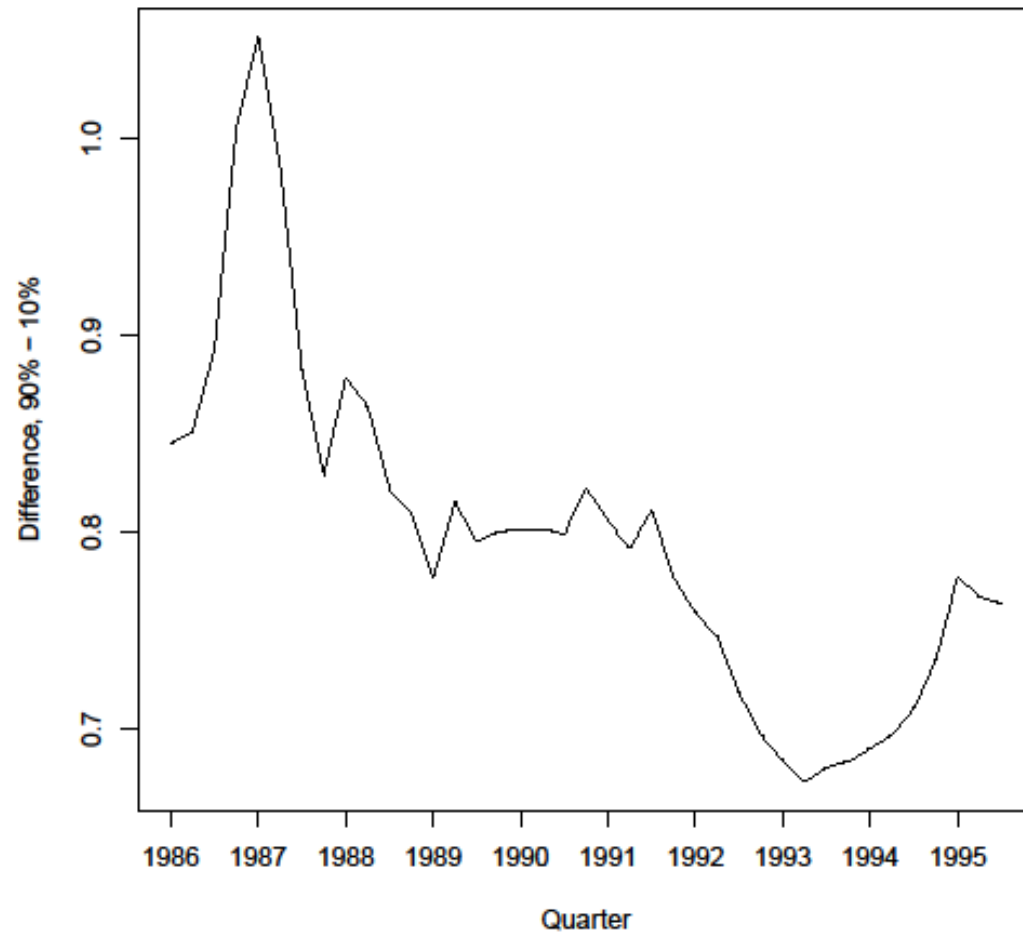
# Estimates for 1991 - 1995 with controls for location and time

Variable	OLS	0.1	0.5	0.9	.9 - .1
Log Building Area	-0.0440 (0.0020)	0.0104 (0.0026)	-0.0472 (0.0017)	-0.0686 (0.0030)	-0.0790 (0.0048)
Age	-0.0221 (0.0001)	-0.0233 (0.0002)	-0.0221 (0.0001)	-0.0213 (0.0002)	0.0020 (0.0002)
South View	0.0135 (0.0014)	0.0132 (0.0020)	0.0158 (0.0013)	0.0104 (0.0023)	-0.0028 (0.0026)
1 <sup>st</sup> Floor	-0.0117 (0.0025)	-0.0128 (0.0038)	-0.0129 (0.0025)	-0.0109 (0.0045)	0.0020 (0.0052)
2 <sup>nd</sup> Floor	-0.0063 (0.0020)	-0.0043 (0.0031)	-0.0072 (0.0020)	-0.0082 (0.0036)	-0.0039 (0.0040)
Floor	0.0057 (0.0003)	0.0065 (0.0004)	0.0055 (0.0003)	0.0049 (0.0005)	-0.0016 (0.0006)

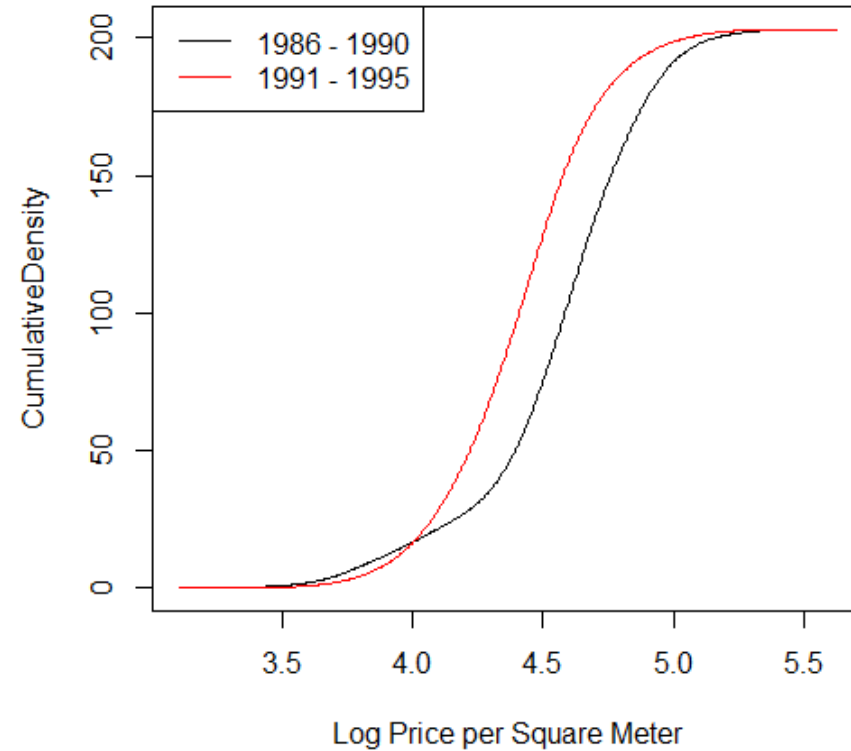
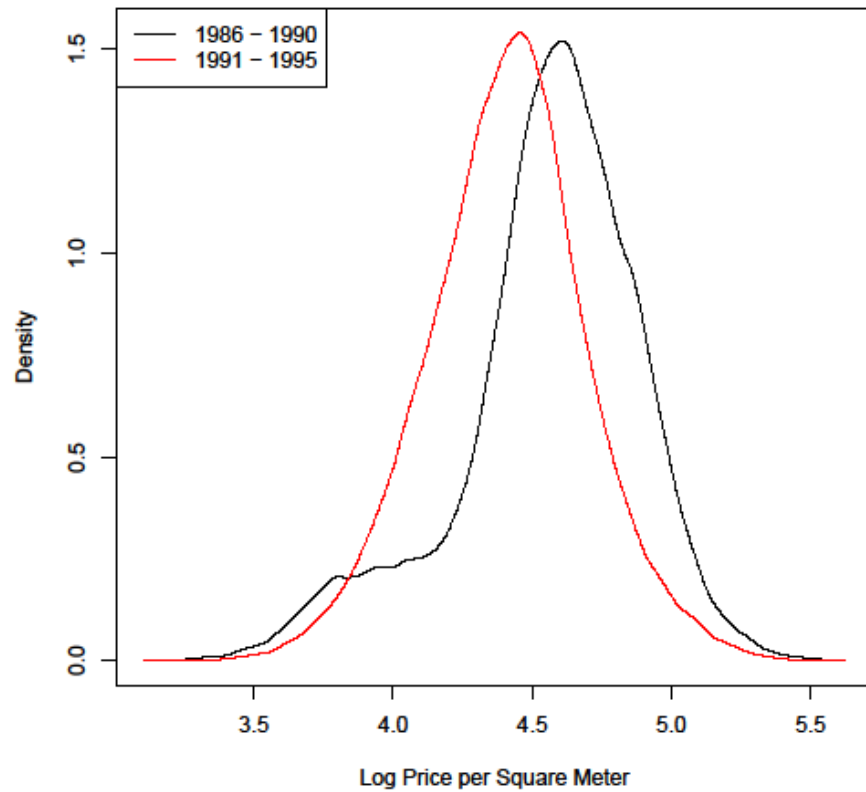
# 10%, 50%, and 90% Quantile Estimates



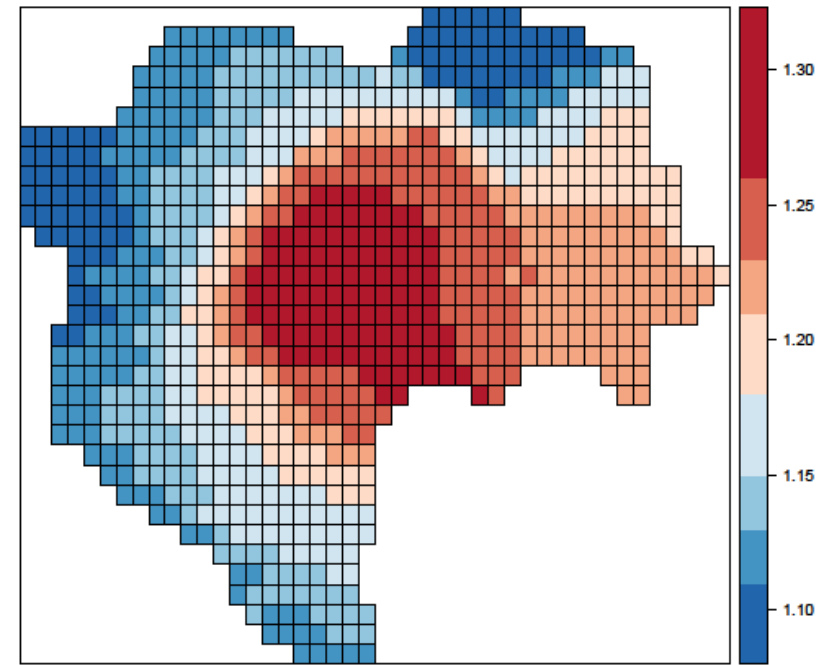
# Difference between 90% and 10% Quantile Estimates, 1986 – 1995



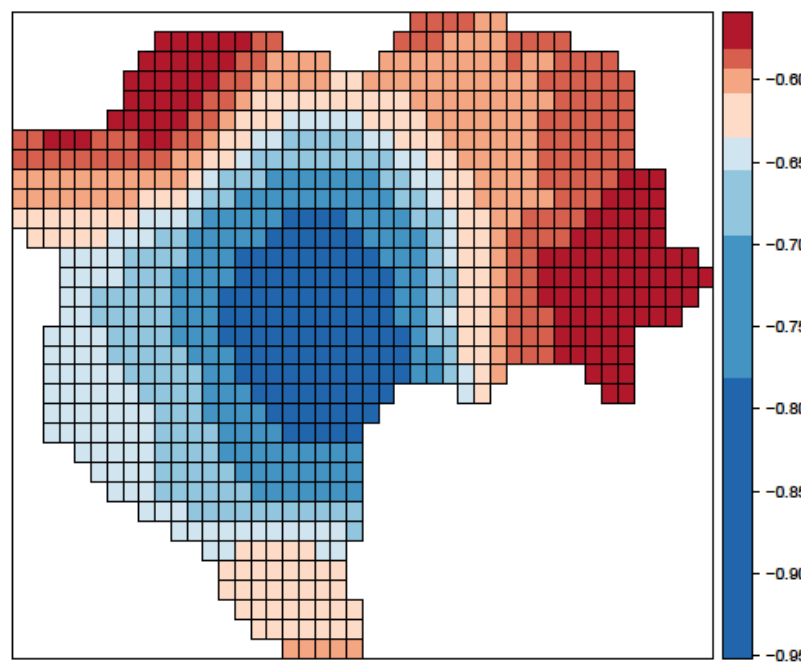
# Kernel Density Estimates for Log Price per Square Meter



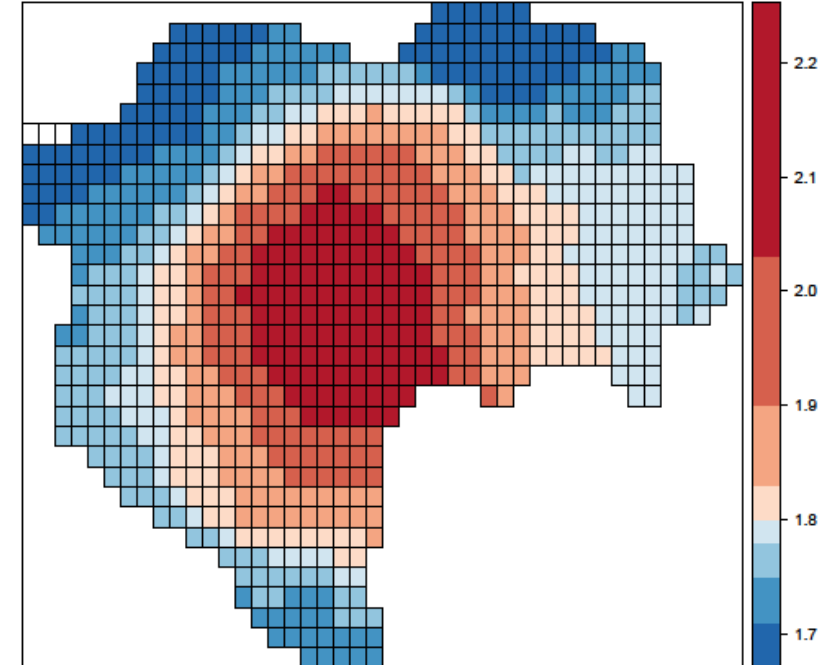
# Spatially Varying Median Appreciation Rates



1986 – 1990

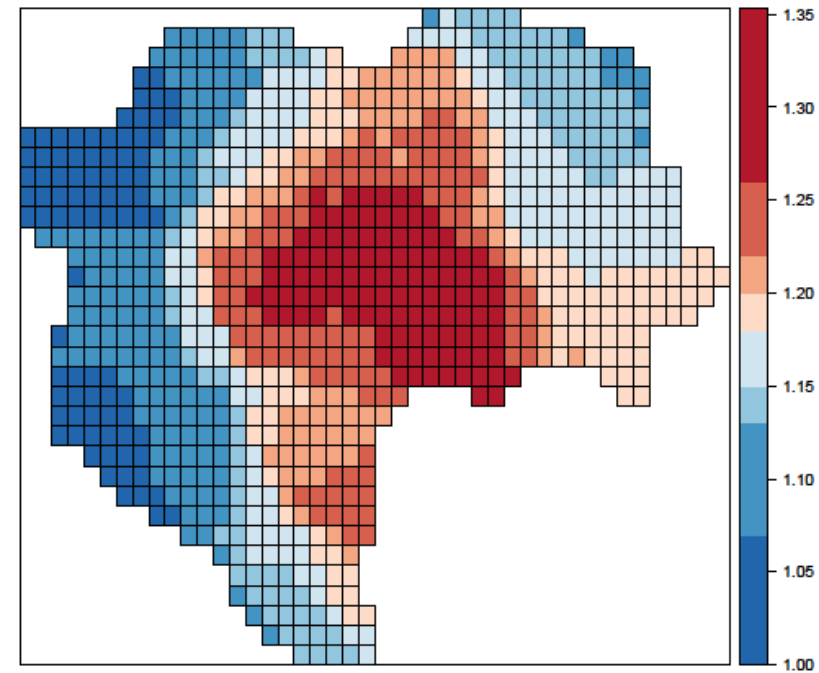


1991 – 1995

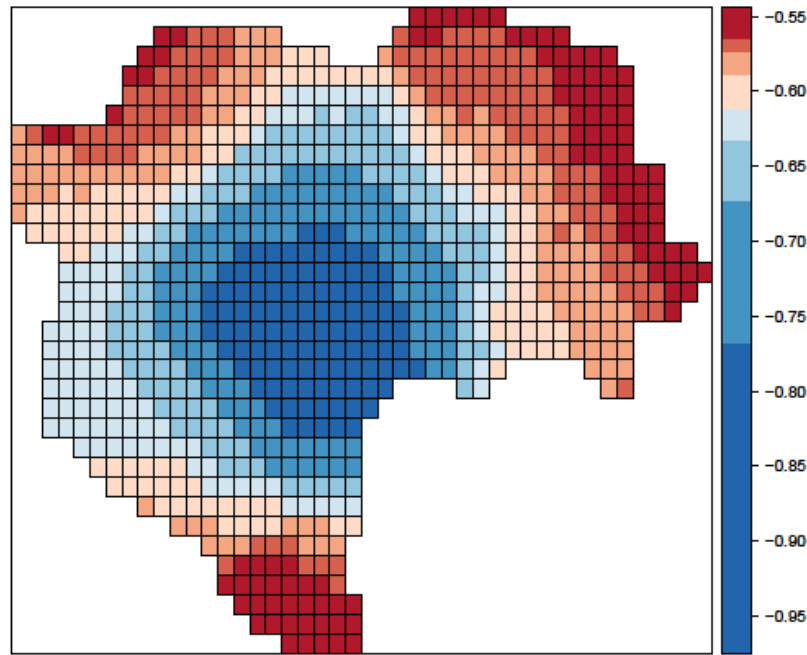


Difference, 1986-90 – 1991-95

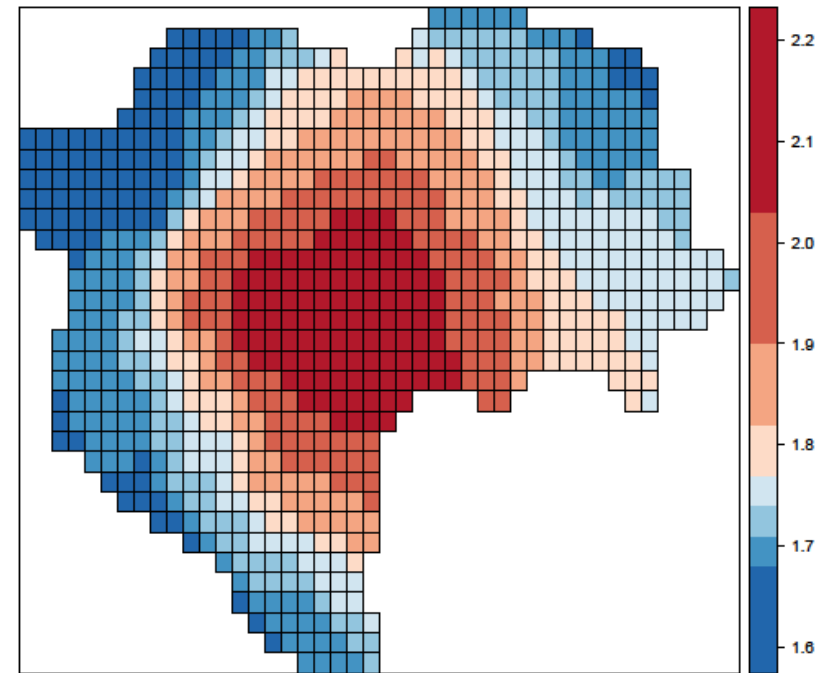
# Spatially Varying 10% Appreciation Rates



1986 – 1990



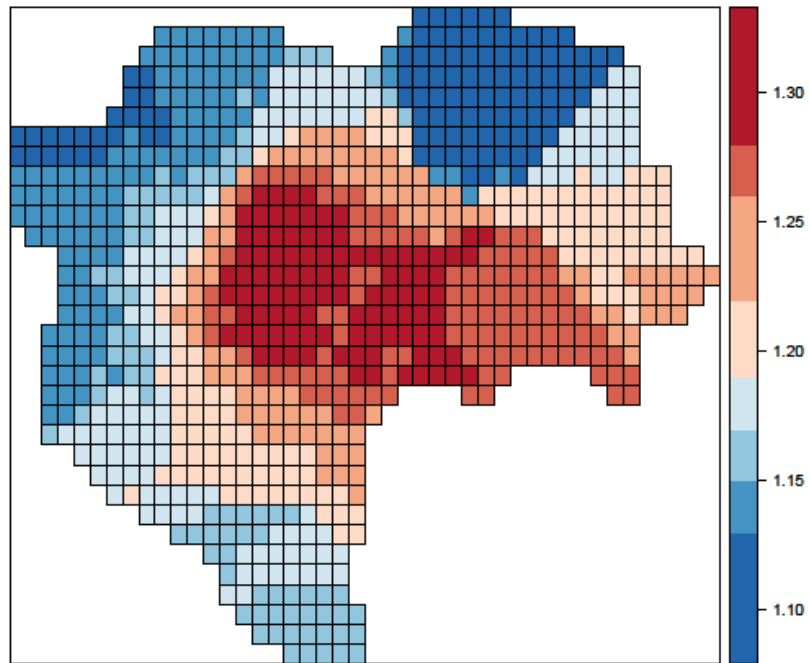
1991 – 1995



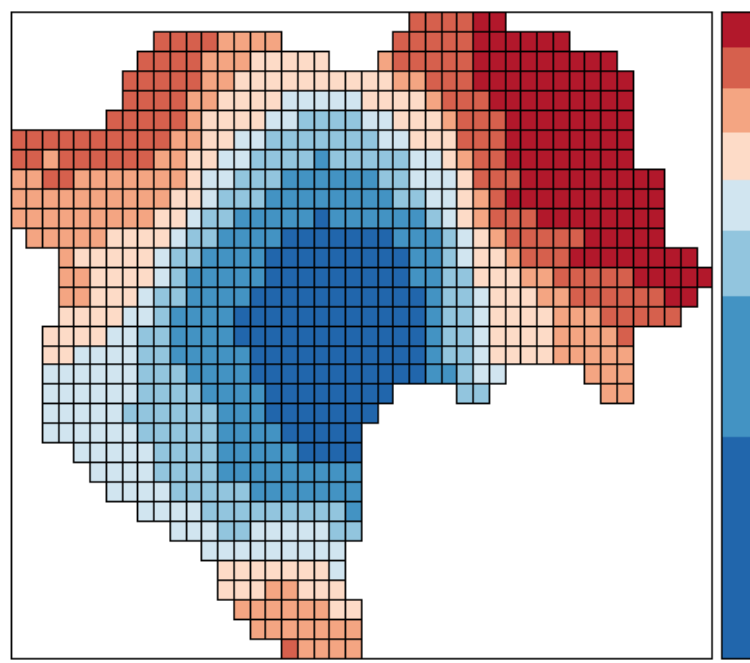
Difference, 1986-90 – 1991-95



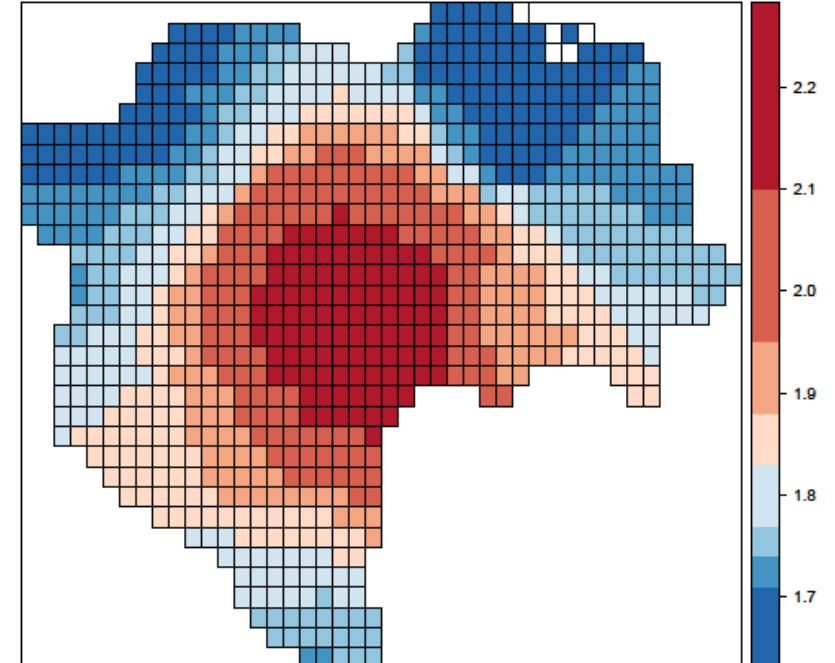
# Spatially Varying 90% Appreciation Rates



1986 – 1990



1991 – 1995



Difference, 1986-90 – 1991-95

## Decompositions: Non-Spatial Version

- Machado & Mata, Journal of Applied Econometrics (2005)

$$y_1(\tau) = x_1\beta_1(\tau) + \delta_1(\tau)$$

$$y_2(\tau) = x_2\beta_2(\tau) + \delta_2(\tau)$$

$$y_1(\tau) - y_2(\tau)$$

$$= (x_1\beta_1(\tau) - x_2\beta_1(\tau)) + (x_2\beta_1(\tau) - x_2\beta_2(\tau))$$

Variables

Coefficients

$$+(\delta_1(\tau) - \delta_2(\tau))$$

Time

# Computations

- $n$  observations,  $k$  explanatory variables,  $B$  quantiles
- $x_1\beta_1(\tau)$ :  $x_1$  is  $n_1 \times k$ ,  $\beta_1(\tau)$  is  $k \times B$
- $x_2\beta_2(\tau)$ :  $x_2$  is  $n_2 \times k$ ,  $\beta_2(\tau)$  is  $k \times B$
- $x_2\beta_1(\tau)$ :  $x_2$  is  $n_2 \times k$ ,  $\beta_1(\tau)$  is  $k \times B$

Machado & Mata: Draw  $M$  draws from the rows of  $x_1$  and the columns of  $\beta_1(\tau)$ . Also make  $M$  draws from the rows of  $x_2$  and the columns of  $\beta_2(\tau)$ .

Construct  $x_1\beta_1(\tau)$ ,  $x_2\beta_2(\tau)$ , and  $x_2\beta_1(\tau)$  from the new  $M \times k$  and  $k \times M$  matrices.

# Spatially varying coefficients

$$y_1(z_1, \tau) = x_1\beta_1(z_1, \tau) + \delta_1(z_1, \tau)$$

$$y_2(z_2, \tau) = x_2\beta_2(z_2, \tau) + \delta_2(z_2, \tau)$$

$$y_1(z_1, \tau) - y_2(z_2, \tau)$$

$$= \underbrace{(x_1\beta_1(z_1, \tau) - x_2\beta_1(z_1, \tau))}_{\text{Variables}} + \underbrace{(x_2\beta_1(z_1, \tau) - x_2\beta_2(z_1, \tau))}_{\text{X Coefficients}} + \underbrace{(\delta_1(z_1, \tau) - \delta_2(z_1, \tau))}_{\text{Time Coefficients}}$$

$$+ \underbrace{(x_2\beta_2(z_1, \tau) - x_2\beta_2(z_2, \tau))}_{\text{Location, X}} + \underbrace{(\delta_2(z_1, \tau) - \delta_2(z_2, \tau))}_{\text{Location, Time}}$$

Need to estimate  $\beta_2(z_1, \tau)$  and  $\delta_2(z_1, \tau)$ : the coefficients in time 2 at the time 1 sale locations.

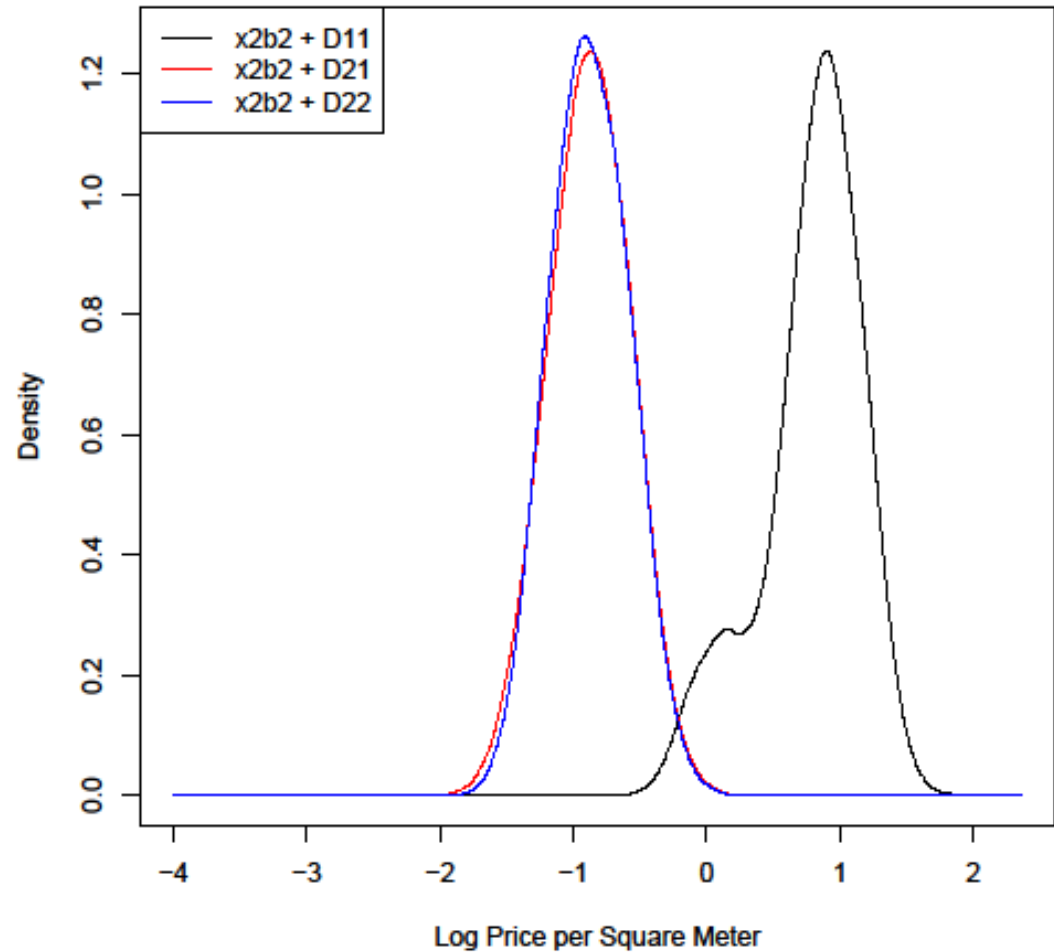
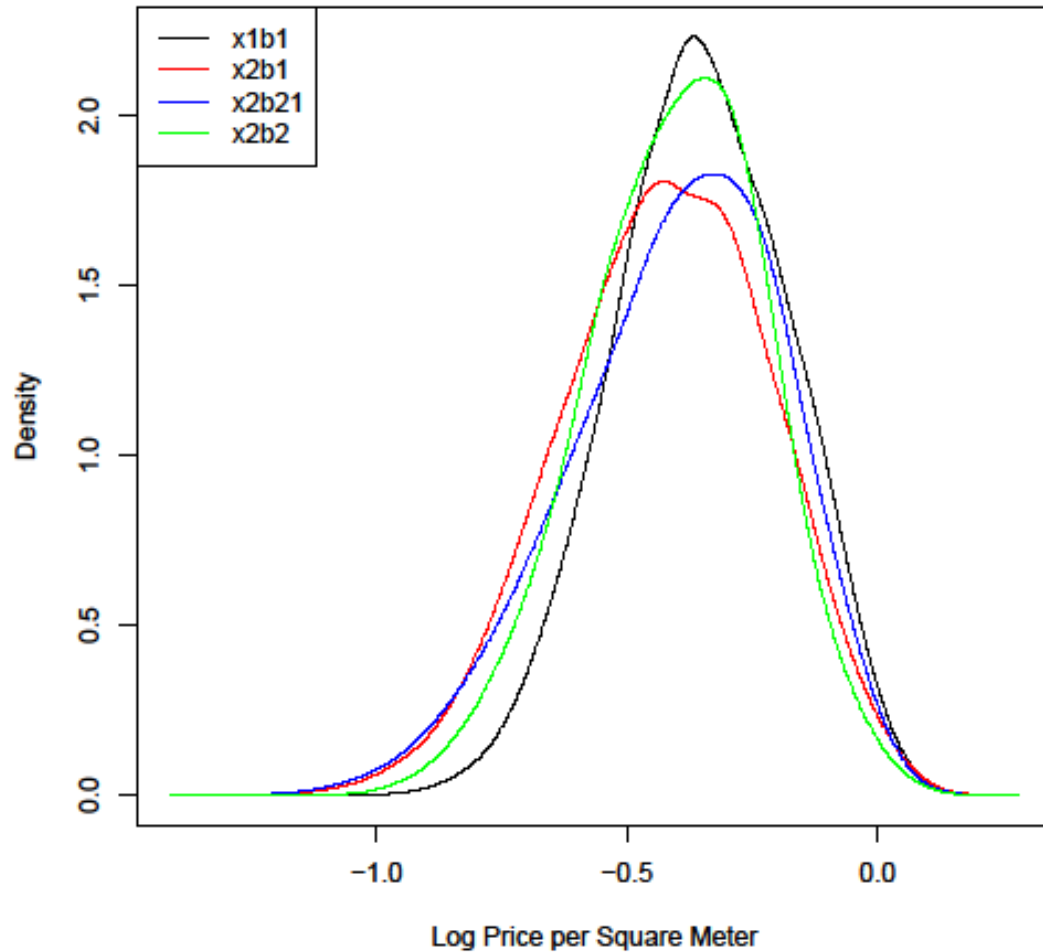
# Spatially varying coefficients

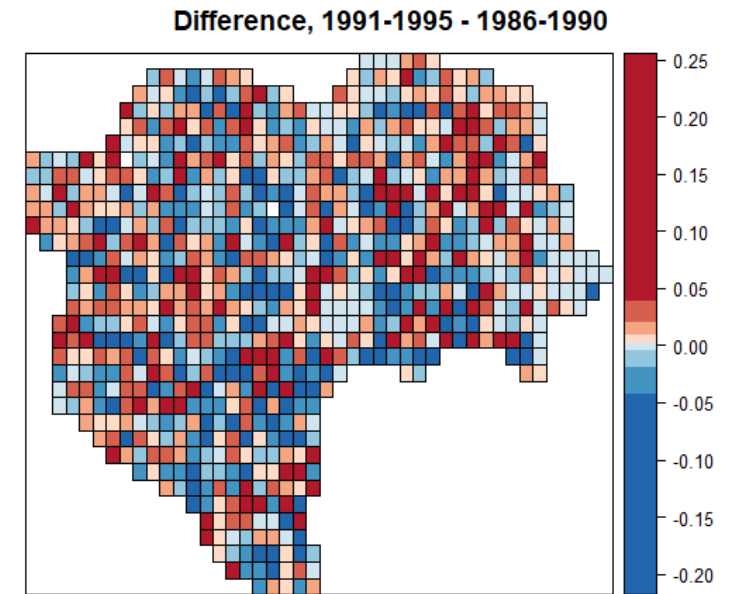
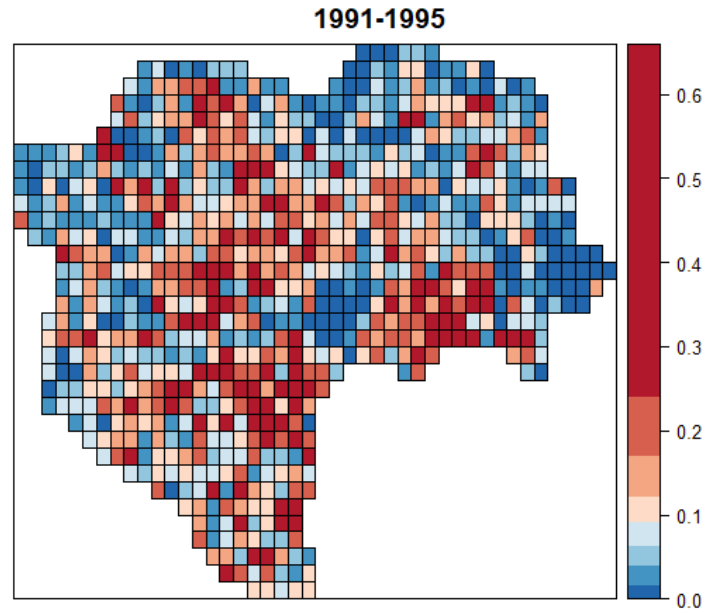
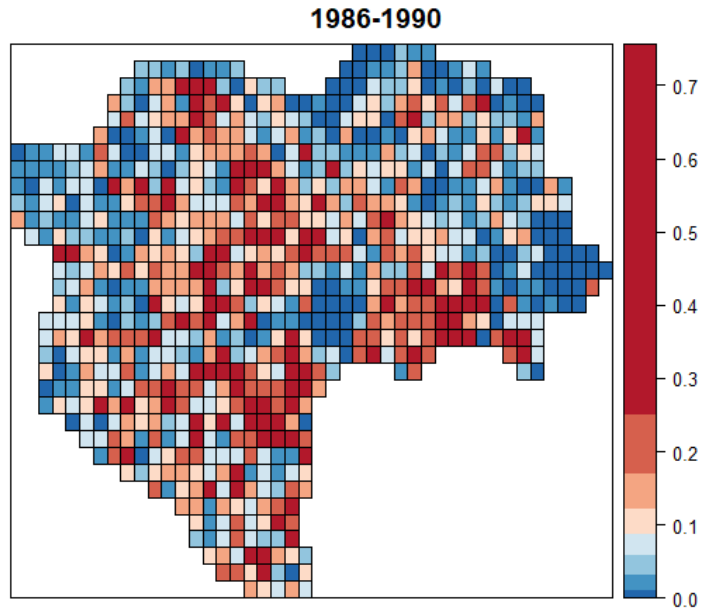
Estimate 3 locally weighted models:

- $y_1(z_1, \tau) = x_1\beta_1(z_1, \tau) + \delta_1(z_1, \tau)$ , time 1 data and locations, B quantiles
- $y_2(z_2, \tau) = x_2\beta_2(z_2, \tau) + \delta_2(z_2, \tau)$ , time 2 locations, time 2 data, B quantiles
- $y_2(z_1, \tau) = x_2\beta_2(z_1, \tau) + \delta_2(z_1, \tau)$ , time 1 locations, time 2 data, B quantiles

Results:  $\hat{\beta}_1, \hat{\delta}_1, \hat{\beta}_2, \hat{\delta}_2, \hat{\beta}_{21}, \hat{\delta}_{21}$

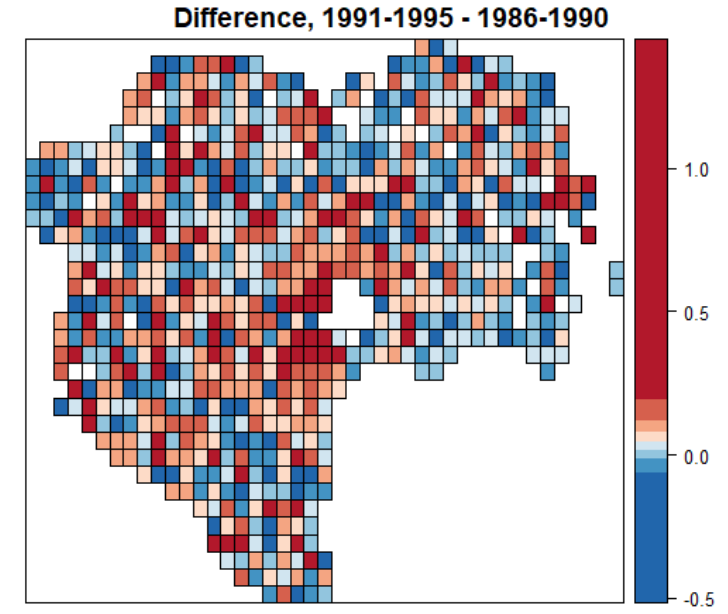
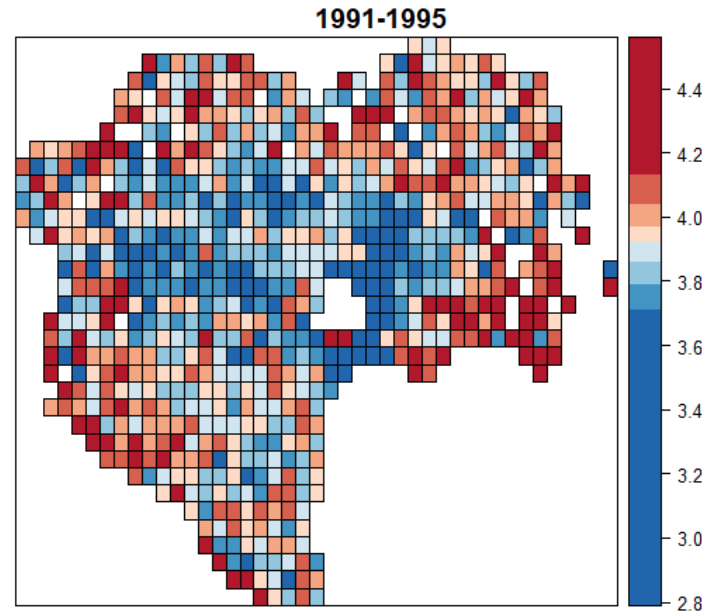
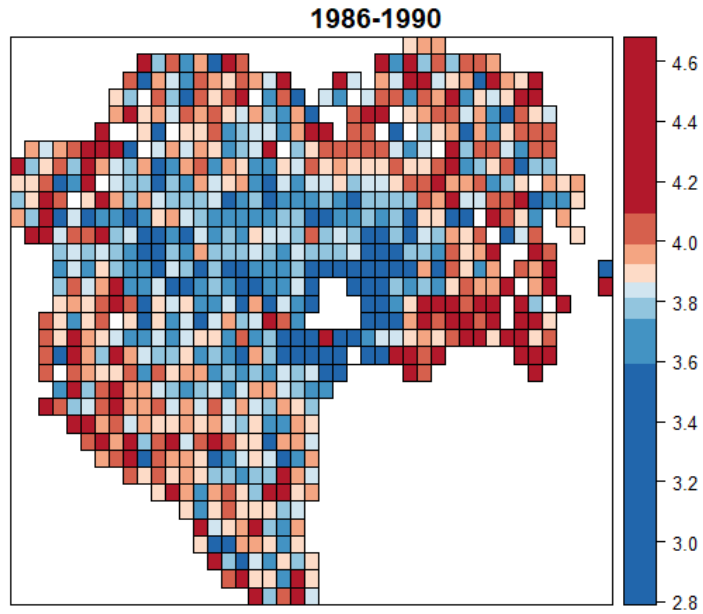
# Decomposition, Tokyo: In building area, age, south view, floor variables





Percent of Observations by Tract

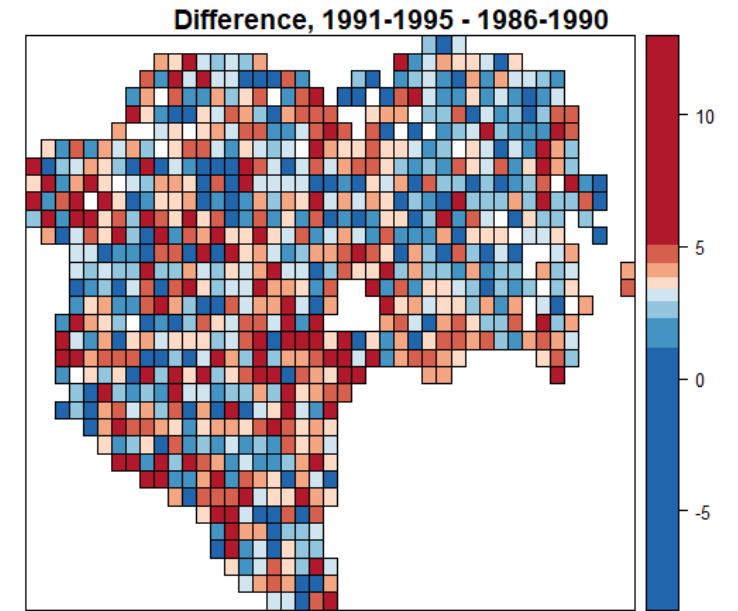
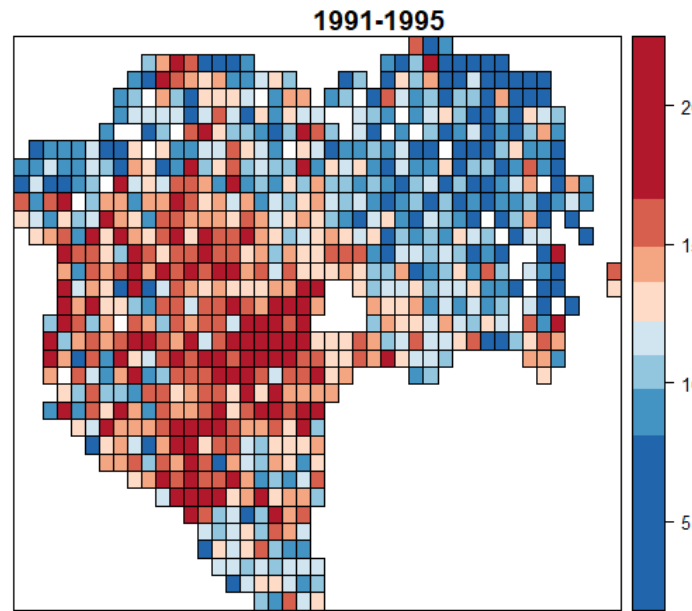
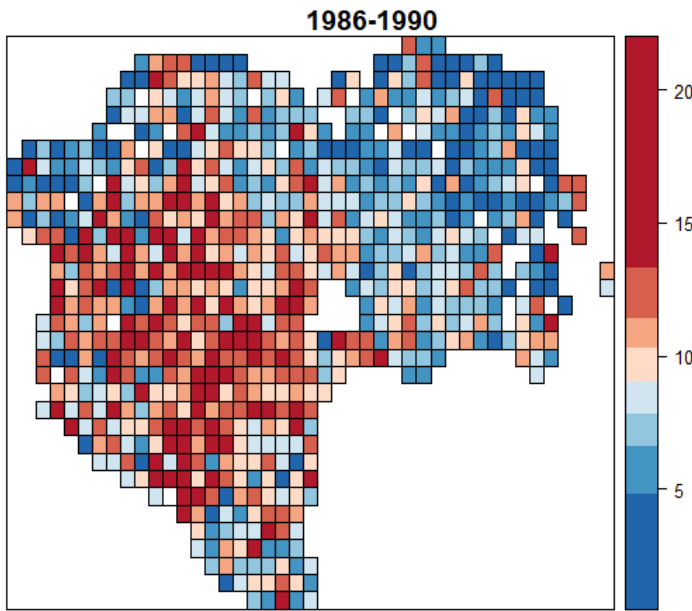
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Mean Log Building Area

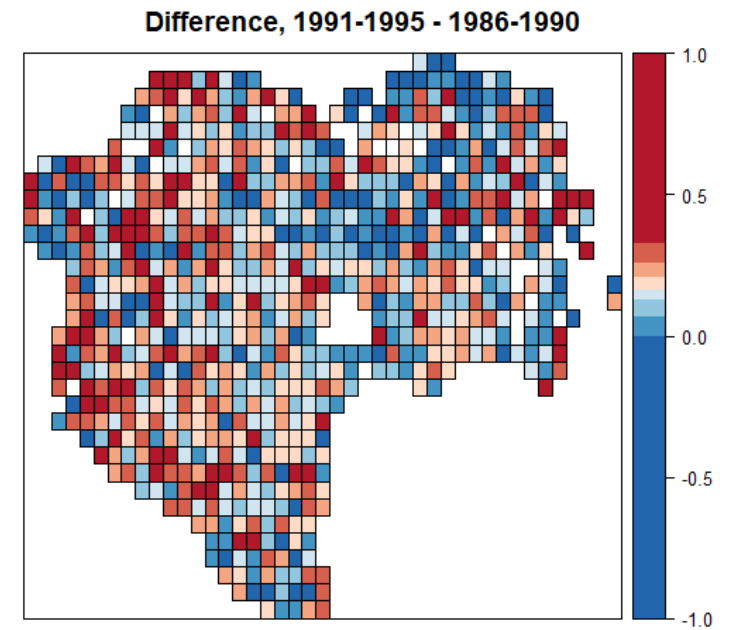
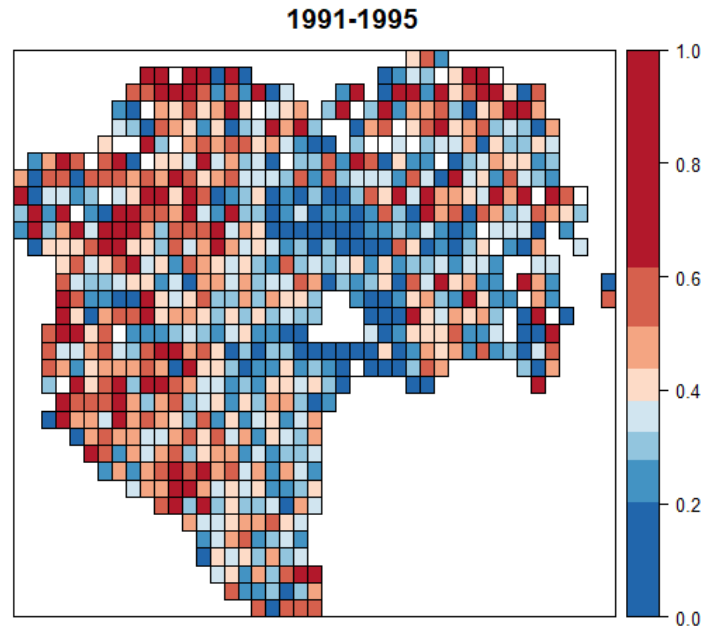
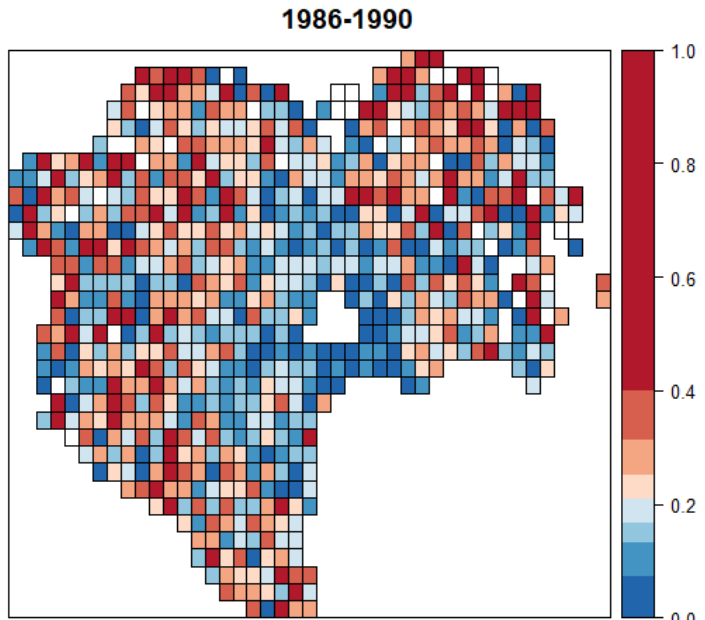
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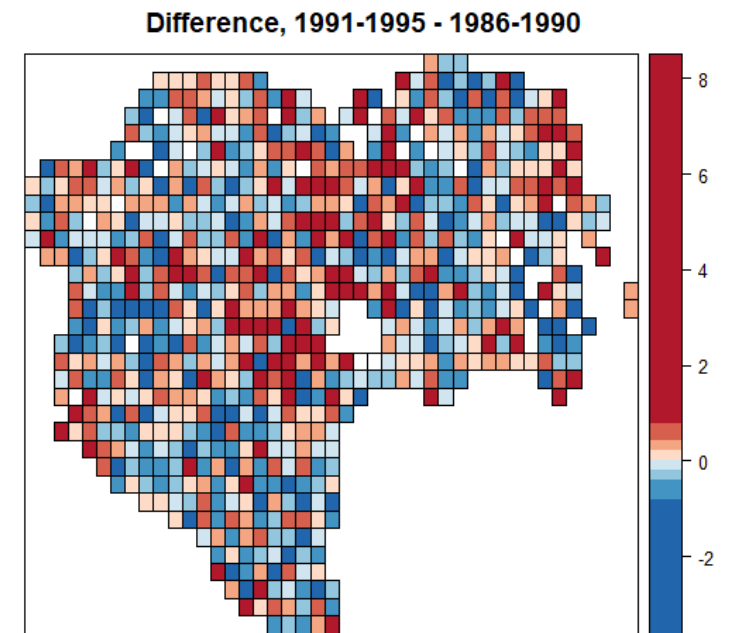
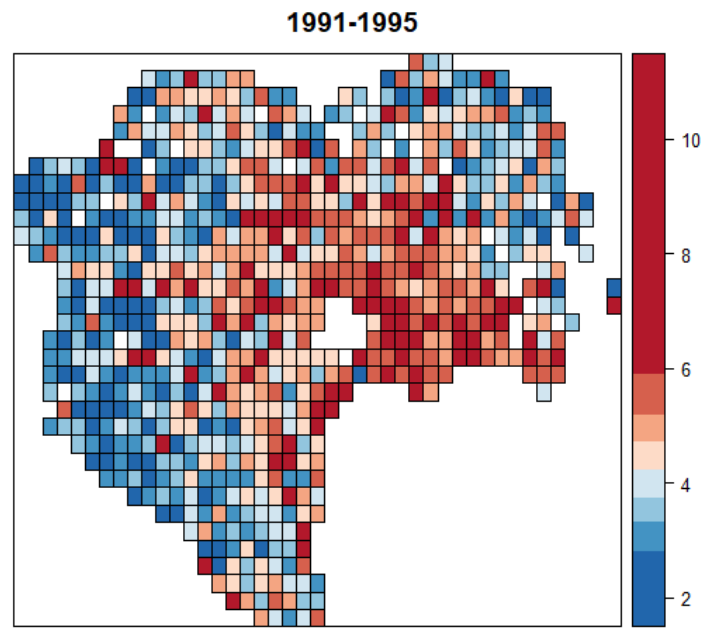
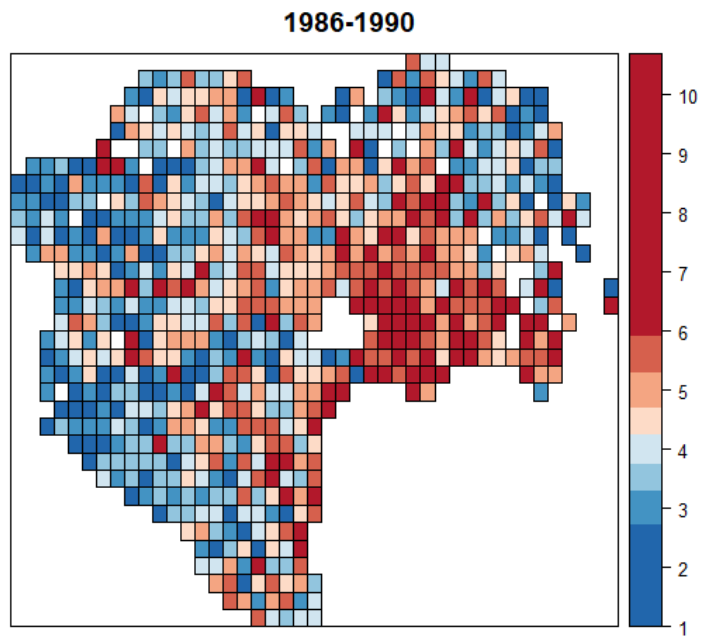


Mean Age

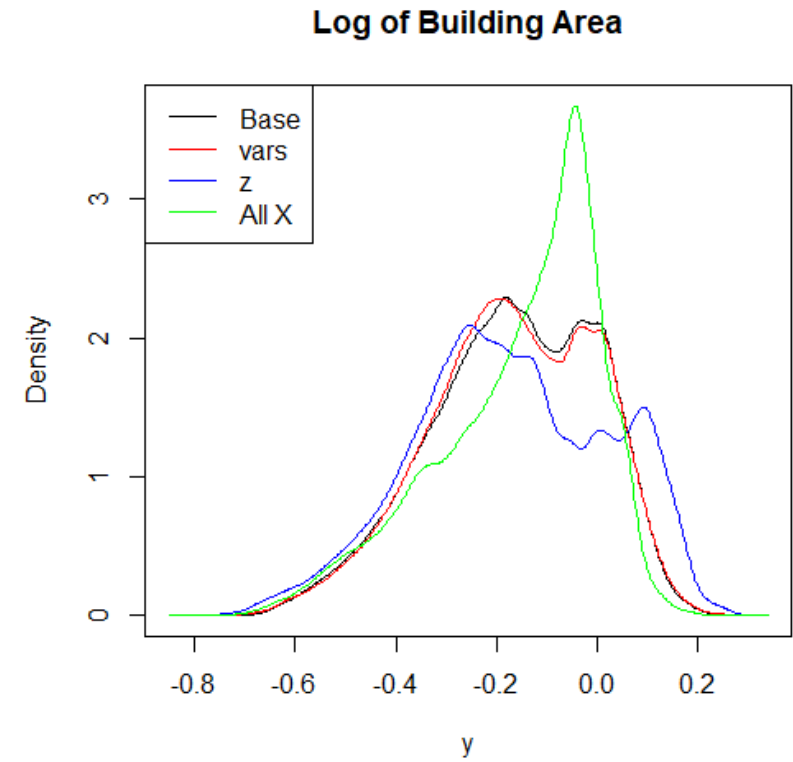
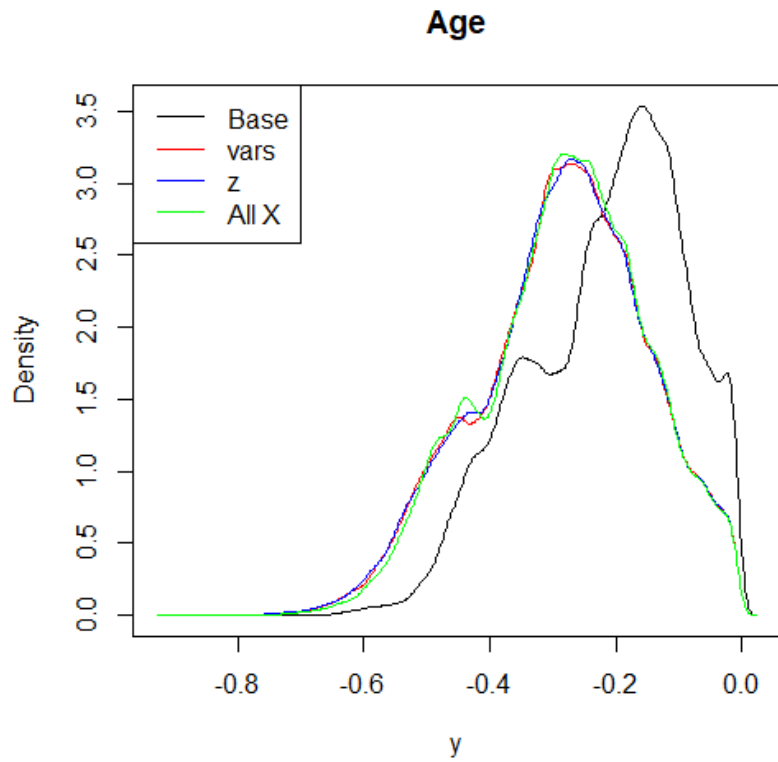
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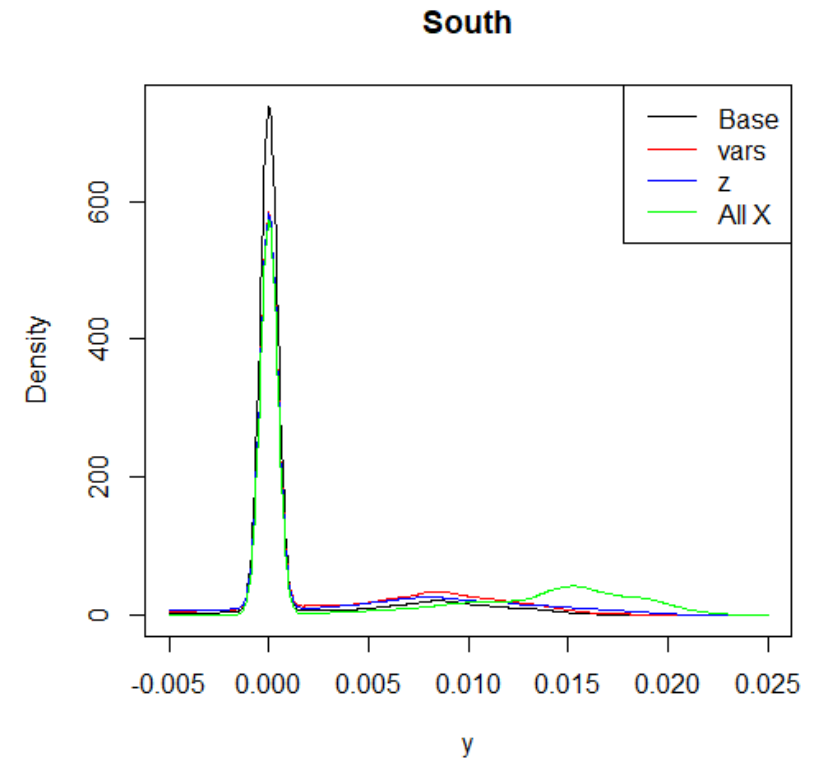
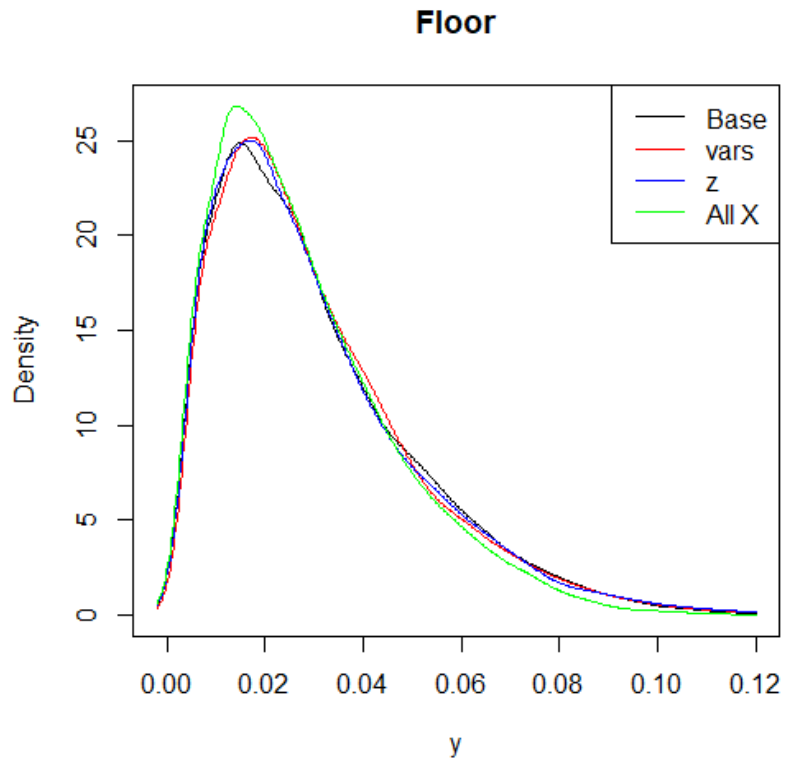
South View



Floor

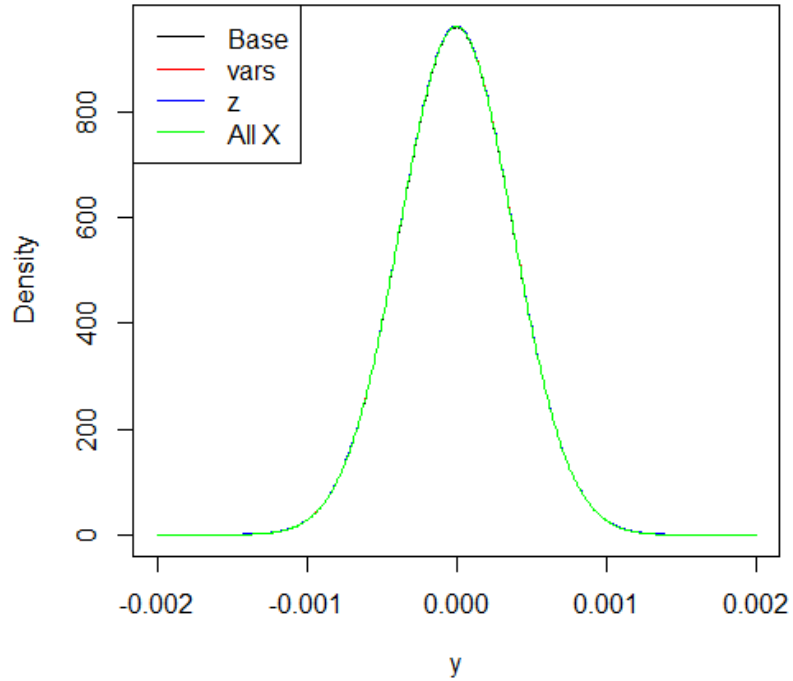


Decomposition for Individual Variables, Tokyo

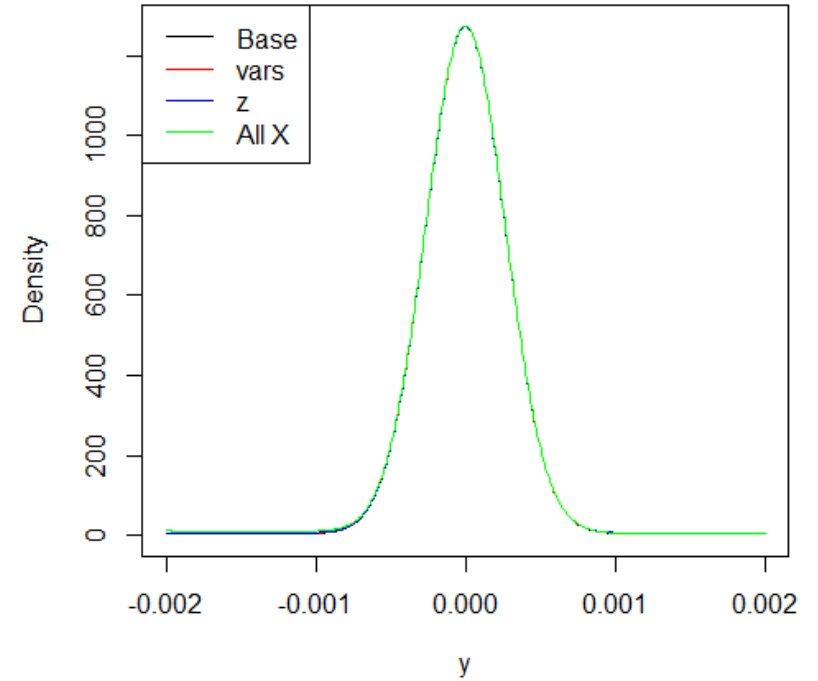


Decomposition for Individual Variables, Tokyo

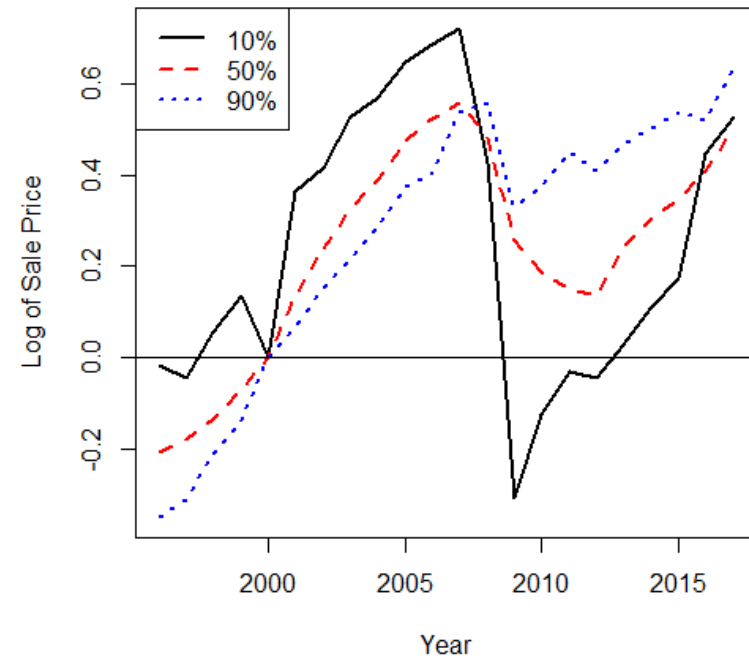
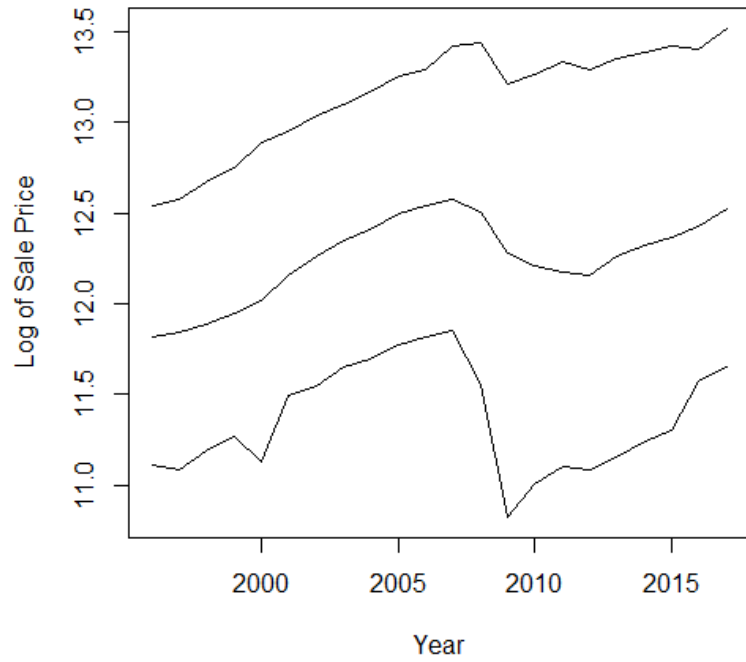
1st Floor



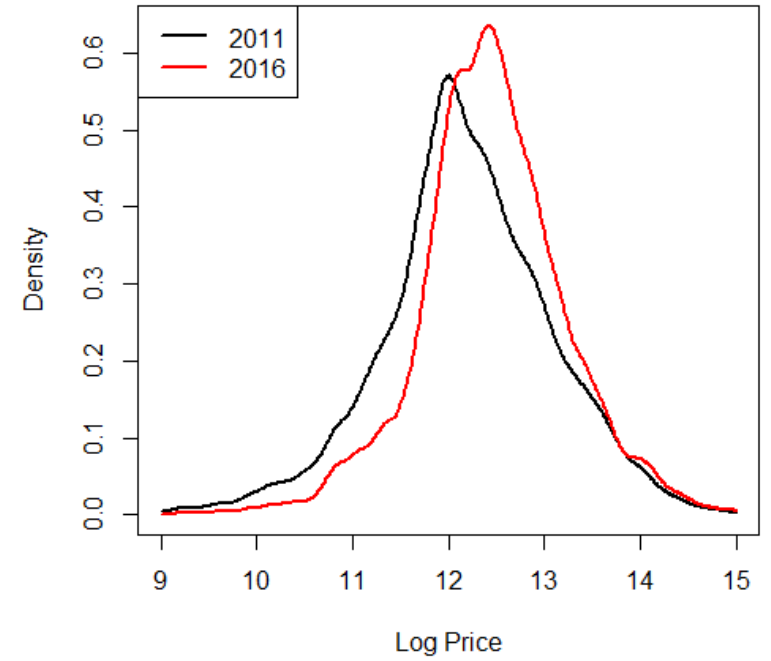
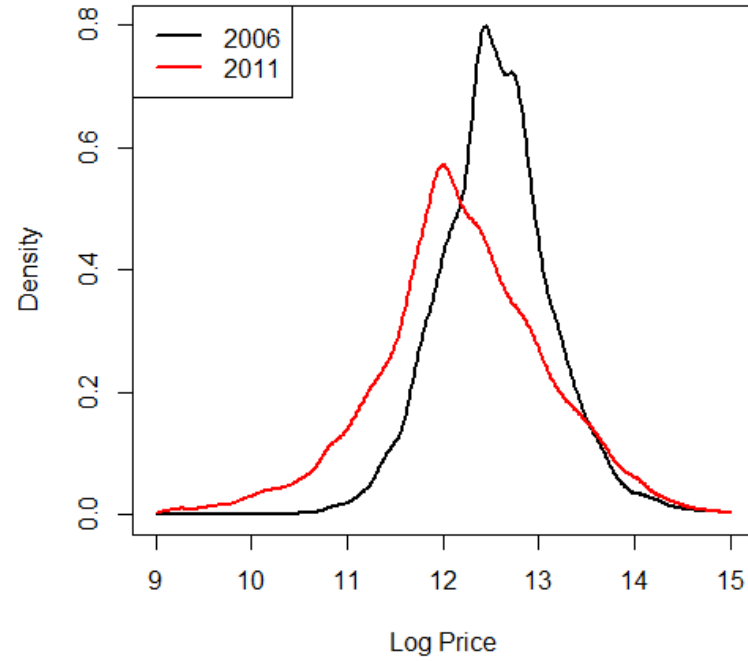
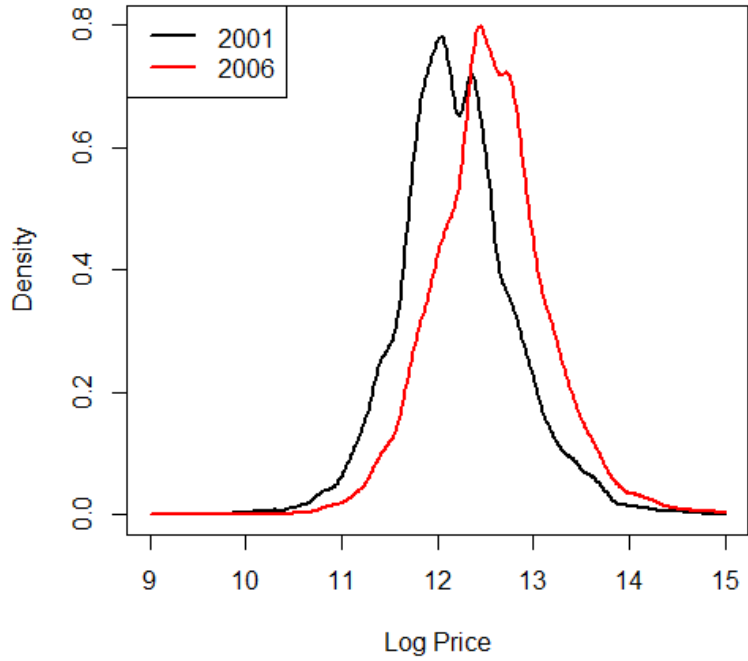
2nd Floor



Decomposition for Individual Variables, Tokyo



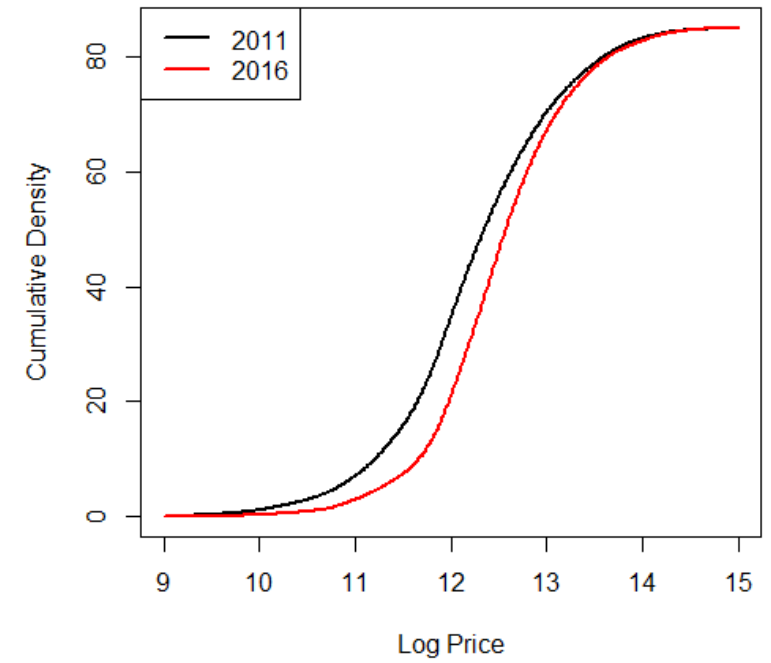
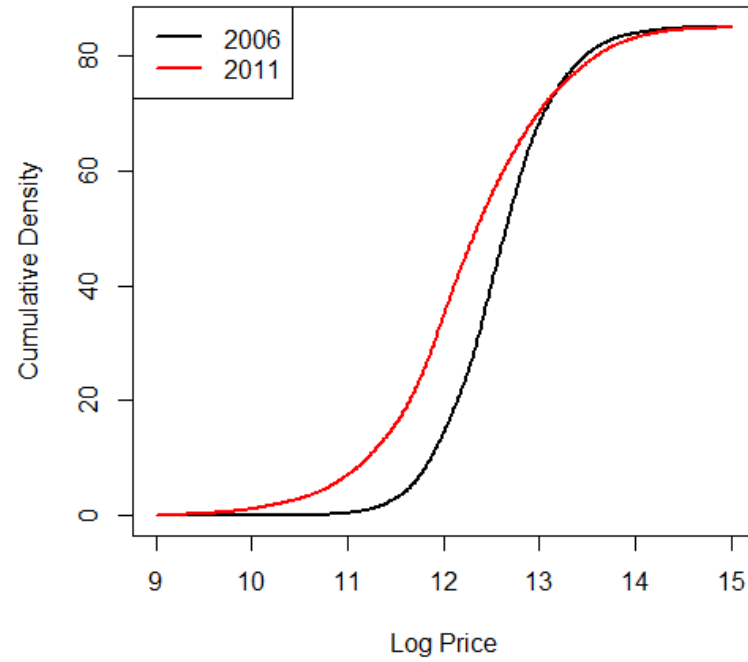
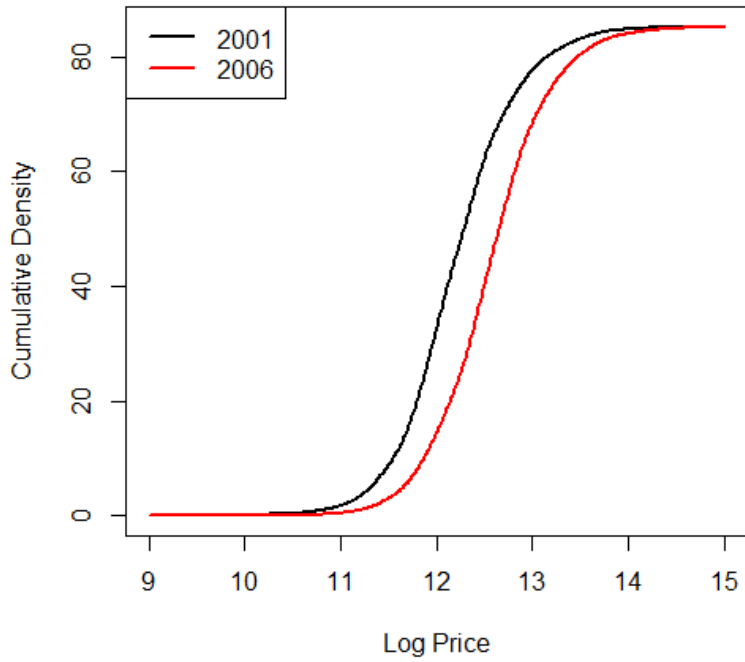
Chicago, 1996-2017



# Density of Sales Prices in Chicago

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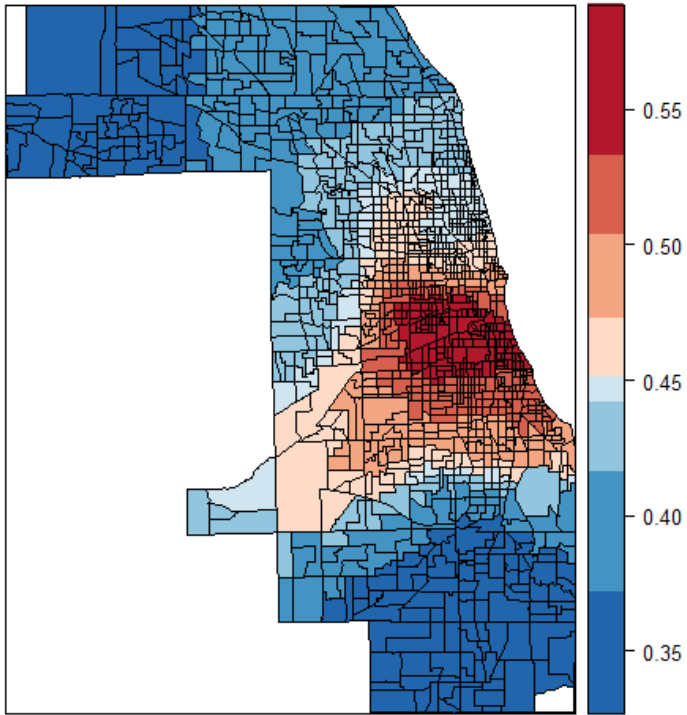




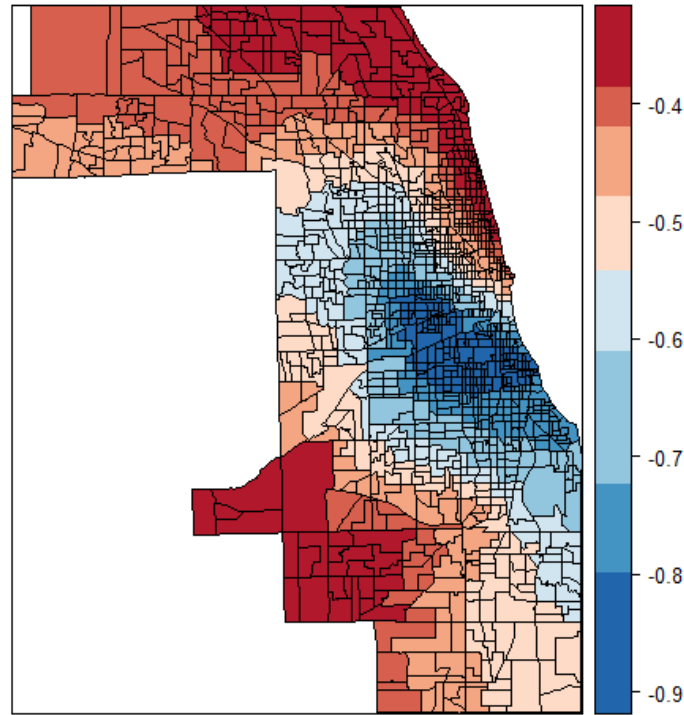
# Cumulative Densities, Chicago

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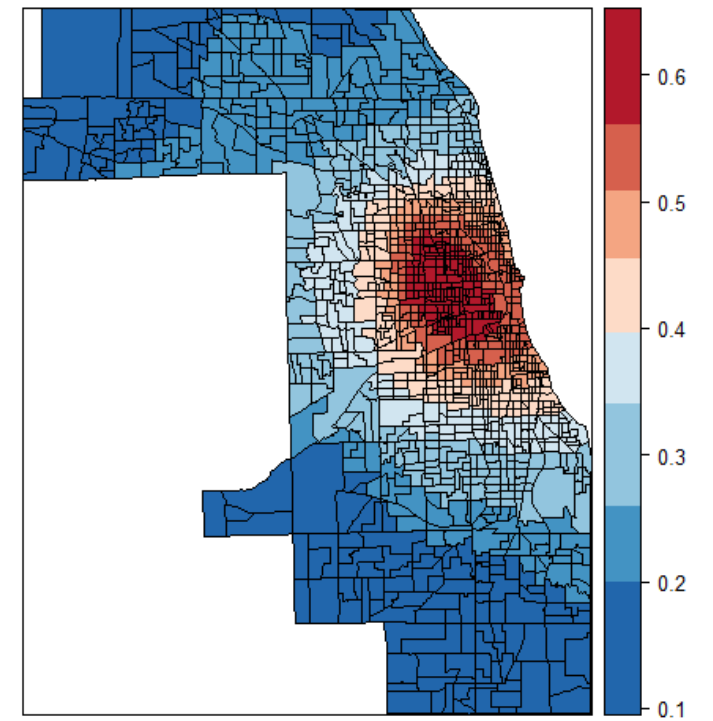
2001 - 2006



2006 - 2011



2011 - 2016



# Appreciation Rates, Chicago

# Chicago

