

Hitotsubashi-RIETI International Workshop on Real Estate Markets and the Macro Economy

Sticky Rent and Housing Prices

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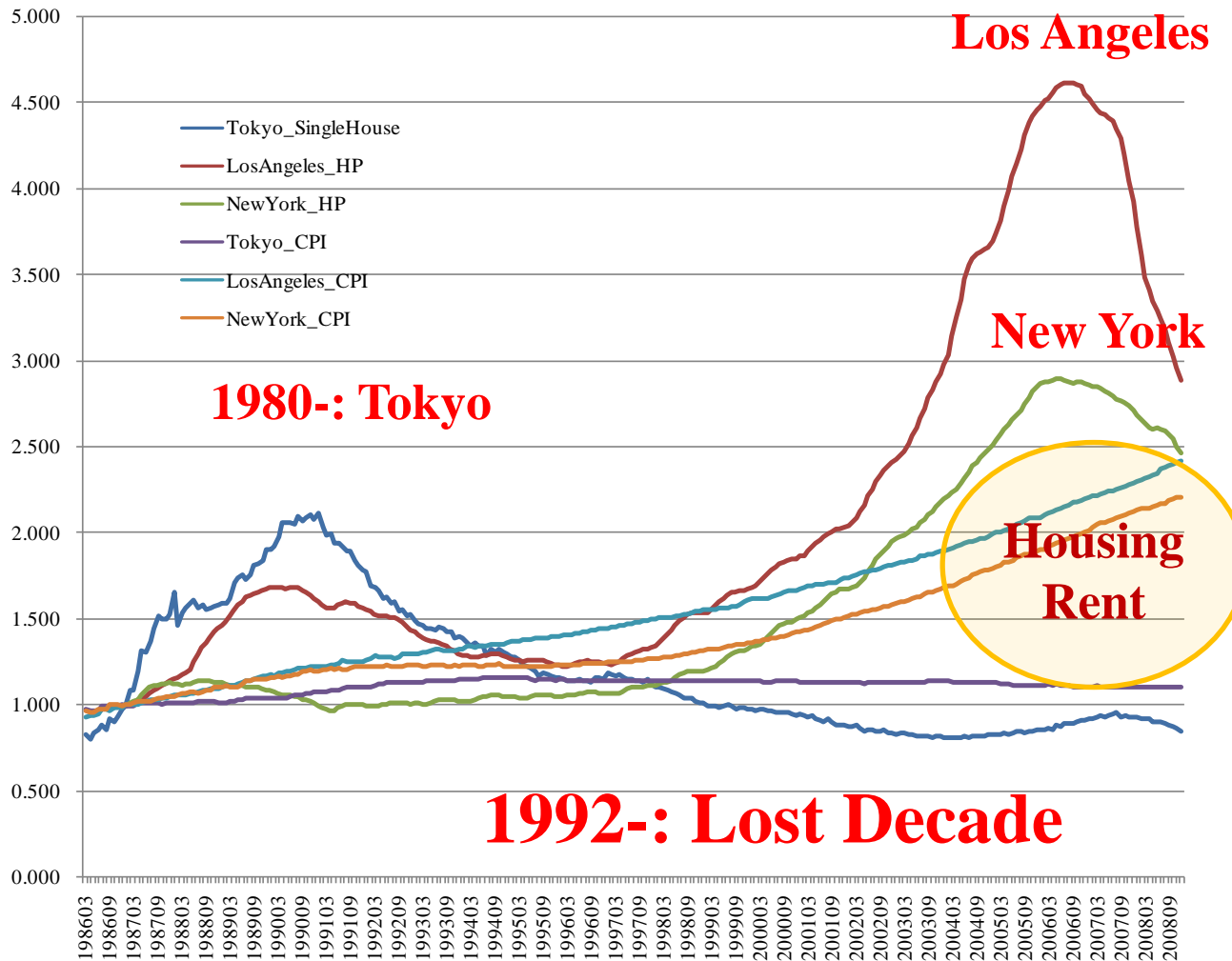
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Comparison House Price and CPI-house in Tokyo, LA, NY



Outline or Questions

- **Why are housing rents in CPI and National Accounts sticky?**
 - Real estate prices and goods & services prices is linked through housing rents. But housing rents did not rise much, and consequently CPI inflation was stable even during the bubble period. This caused delay in monetary tightening
 - How much and why are housing rents sticky?
 - Proposals to make CPI more linked to asset price developments
- **How should we estimate Residential Property Price Indexes or RPPI?**
 - In the wake of the release of this Handbook, how should different countries construct residential property price indexes?
 - Hedonic, Repeat Sales, SPAR, Conjoint, Matching etc.

Background papers.

- **Rigidity of Housing Rent.**

- Shimizu, C., K. G. Nishimura and T. Watanabe (2010), “Residential Rents and Price Rigidity: Micro Structure and Macro Consequences,” *Journal of Japanese and International Economy*, Vol.24, 282-299.
- → Shimizu, C, E.W.Diewert and S.Imai (2015). “Residential Rents and CPI,” *Conference Paper at Ottawa Group Meeting, May 2015.*

- **Comparison of Residential Property Price Indexes.**

- Shimizu, C., K. G. Nishimura and T. Watanabe (2010), “House Prices in Tokyo - A Comparison of Repeat-sales and Hedonic Measures-,” *Journal of Economics and Statistics*, Vol. 230 (6), 792-813.
- → Shimizu, C, and K.Karato (2014). “Estimation of Residential Property Price Indexes in Tokyo-,” *Conference Paper for Hitotsubashi-RIETI International Workshop on Real Estate Markets and the Macro Economy.*

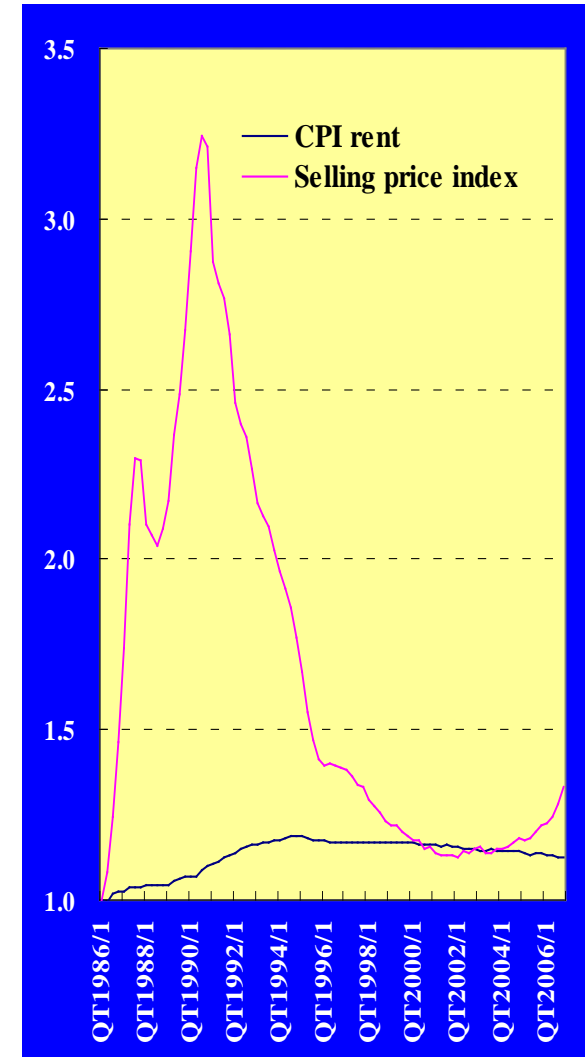
1. Macroeconomic Policy and Housing Market.

Expenditures for housing services: 26.4%

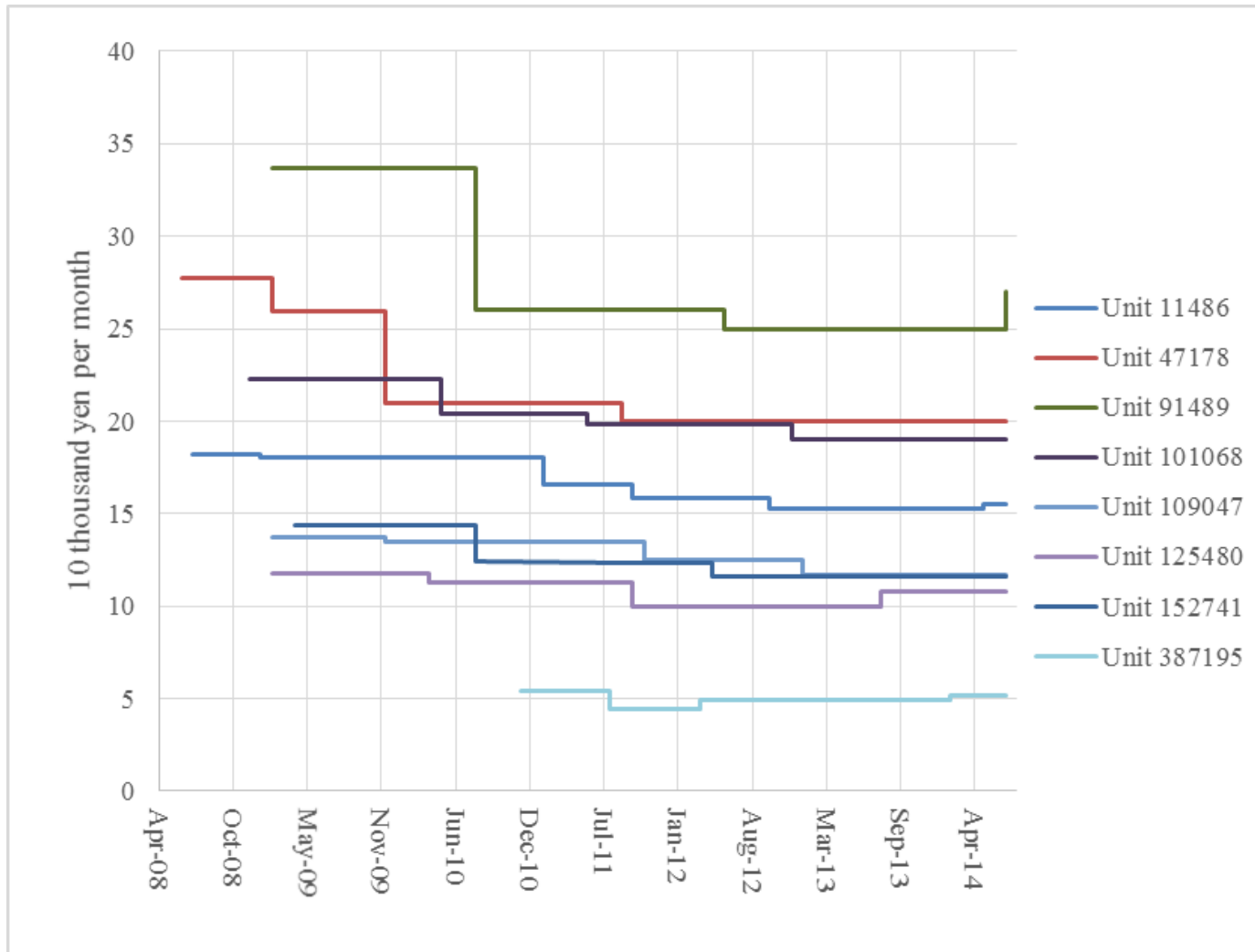
Housing rents:	4.9%
Imputed rents from owner occupied housing:	19.4%
Housing maintenance and others:	2.3%

“Consumer Price Index (CPI) in Tokyo, 2010”

- The most important link between *asset prices* and *goods & services prices* is the one through **housing rents** (Goodhart 2001)
- Housing rents account for more than **one fourth** of personal spending



Panel data of rental prices



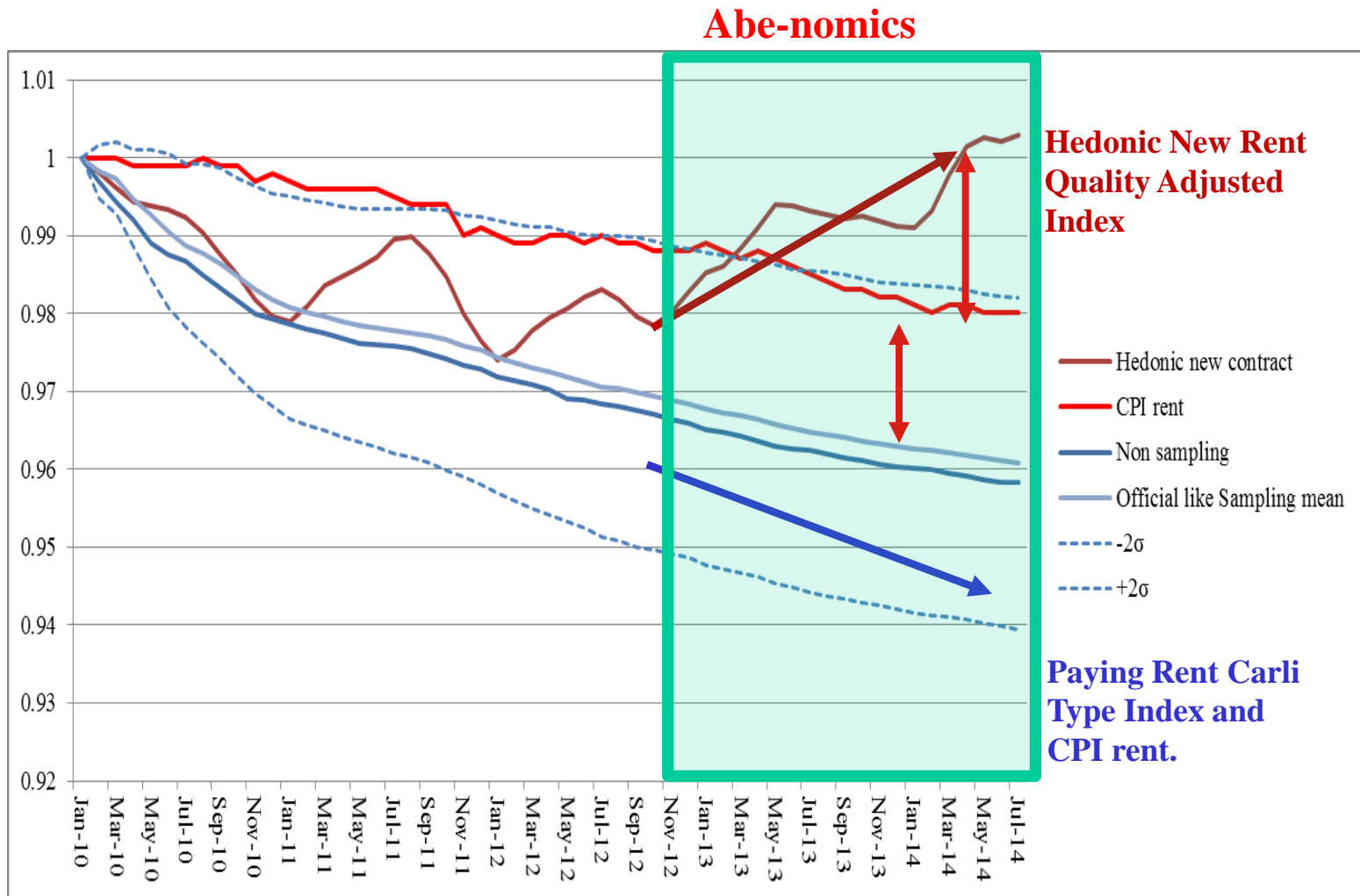
Summary Statistics of Housing Rent

Sample period	January 2010 - July 2014					
Frequency	Monthly					
Geographical scope	Tokyo 23 wards					
Data type	Actual rent					
Coverage	New contract and renewed contract					
Provider	Recruit					
No. of rooms	53,746					
Sample size	Whole sample					
	1,492,542		New contract		Renewed contract	
	40,398		35,139			
	mean	s.d.	mean	s.d.	mean	s.d.
Rent	104,783.6	57,252.5	102,620.2	52,697.4	105,586.8	59,275.9
m²	33.1	17.7	32.7	17.3	33.6	18.2
m² unit rent	3,312.6	924.9	3,295.1	867.3	3,289.4	942.8
Actual age	13.0	9.9	12.2	10.0	13.7	9.8
Time to near station (m)	5.1	3.8	5.0	3.7	5.2	3.9
Time to urban center (m)	12.3	6.4	12.1	6.3	12.5	6.5

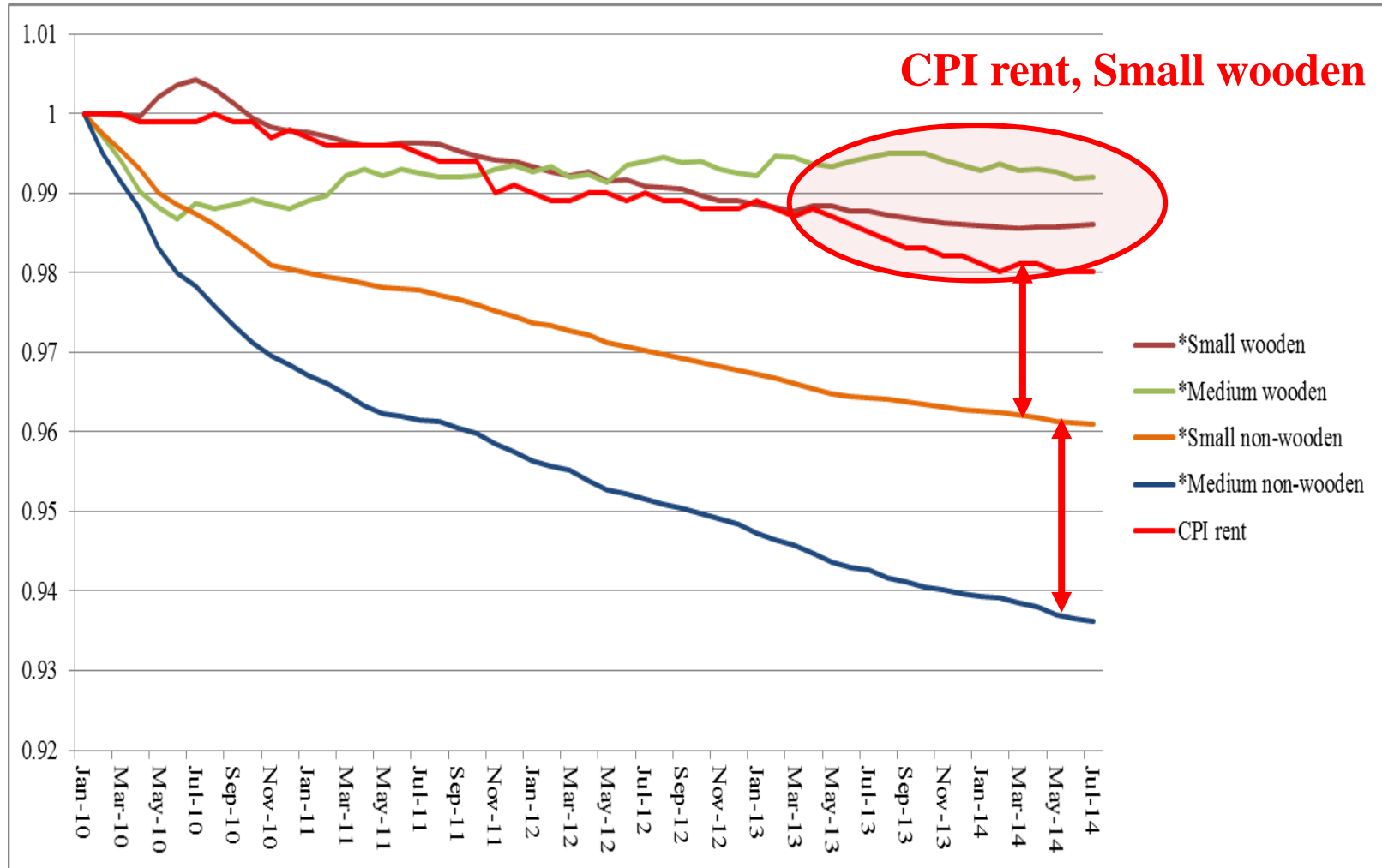
Official Residential Rent Index or CPI Rent

- CPI rent survey districts are selected from enumeration districts of the Population Census by probability sampling. The number of house rent survey districts is **1,221**.
- Statistics Bureau of Japan says that about **28,000 households** are surveyed (see Annual report on the Retail Price Survey 2013).
- The survey districts are allocated according to scale of sample cities, the Tokyo metropolitan area is allocated **54** districts.
- The survey districts are grouped to three groups and one group is surveyed **every 3 months**.
- Rent index is calculated separately by 4 classification.-
Wooden small house, wooden medium house, non-wooden small house and non-wooden medium house.

Housing Rent Index and CPI Rent 1



Housing Rent Index and CPI Rent 2



Frequency of Rent Adjustments

Price Change

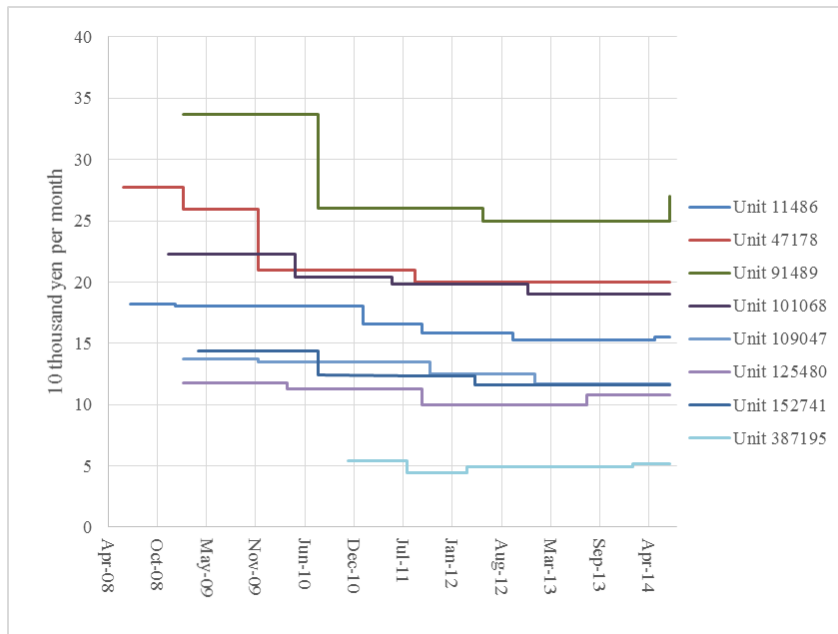
$$\Delta R_{it} \equiv R_{it} - R_{it-1}$$

Probability of event on *New Contract* (I^N) and *Renewed Contract* (I^R)

$$\Pr(\Delta R_{it} = 0) = [1 - \Pr(I_{it}^N = 1) - \Pr(I_{it}^R = 1)]$$

$$+ \Pr(\Delta R_{it} = 0 | I_{it}^N = 1) \Pr(I_{it}^N = 1)$$

$$+ \Pr(\Delta R_{it} = 0 | I_{it}^R = 1) \Pr(I_{it}^R = 1)$$



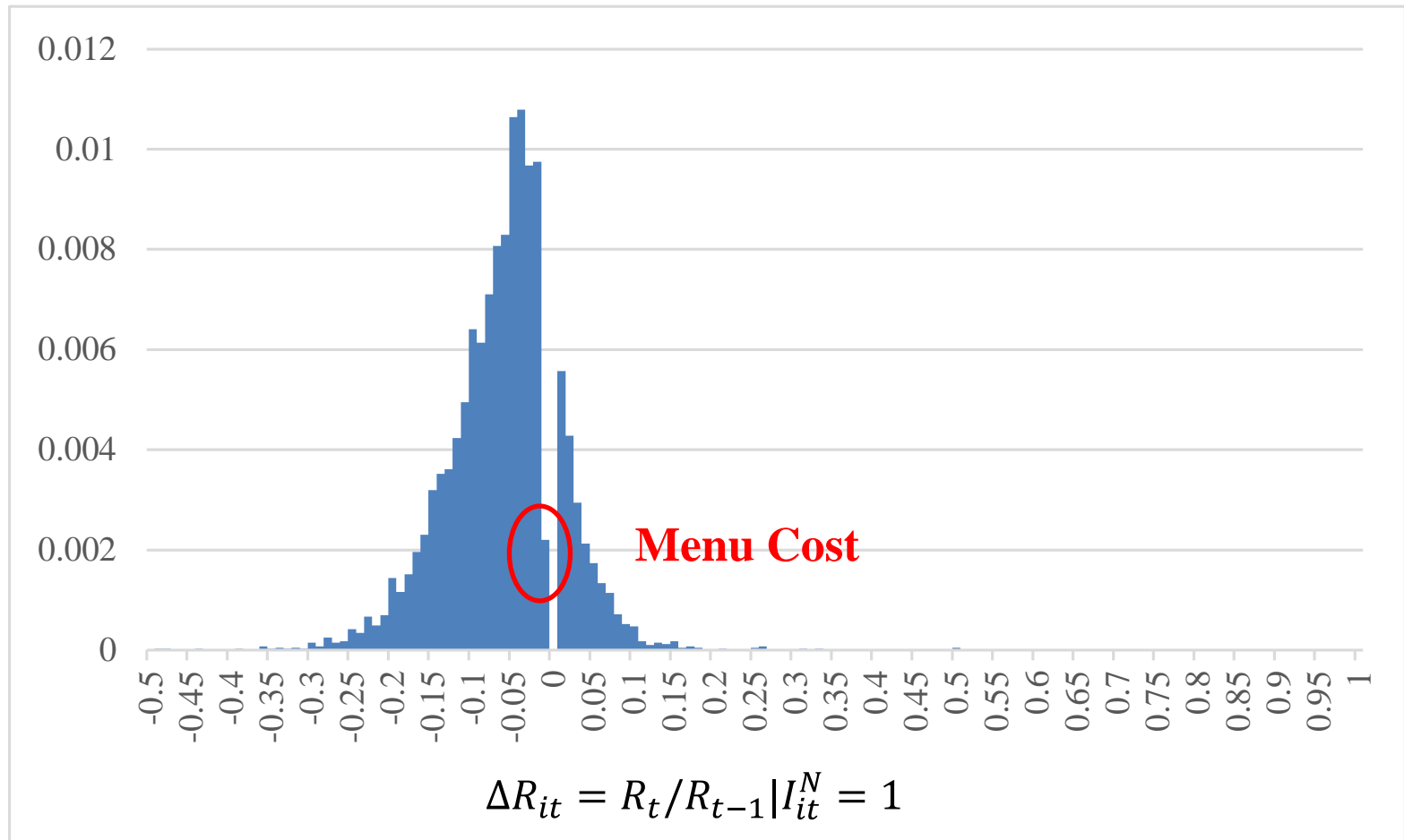
Frequency of Rent Adjustments:

	Negative	Zero	Positive	Number of Observations	(Change)
Turnover units	4,473	34,958	967	40,398	5,440
	(0.111)	(0.865)	(0.024)	(0.535)	(0.135)
Rollover units	1,169	33,667	303	35,139	1,472
	(0.033)	(0.958)	(0.009)	(0.465)	(0.042)
All units	5,642	68,625	1,270	75,537	6,912
	(0.075)	(0.908)	(0.017)	(1.000)	(0.092)

Fraction of housing units without no rent change per year

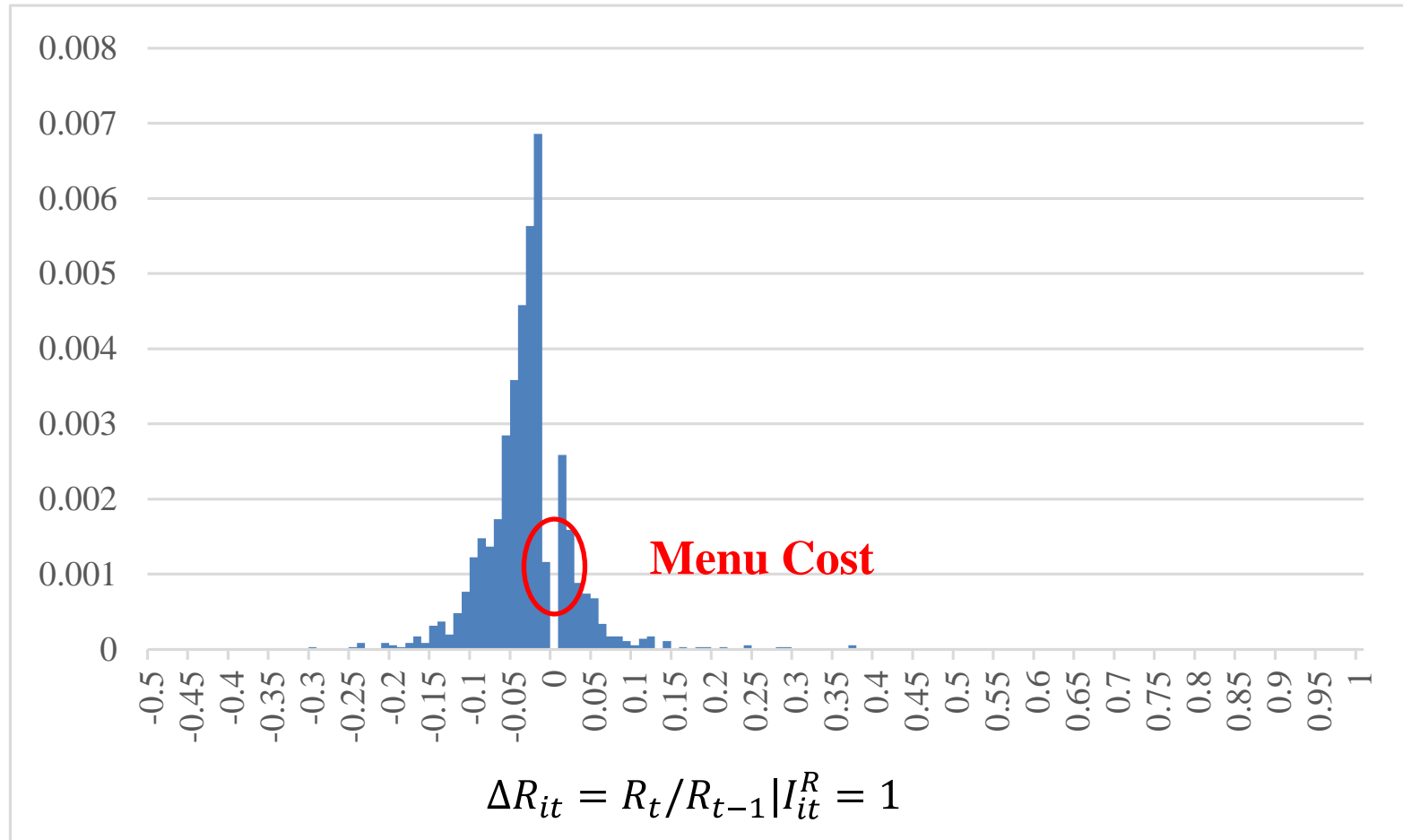
US	29%	Estimated by Genesove (2003)
Germany	78%	Estimated by Kurz-Kim (2006)
Japan	90%	Estimated by this research

Monthly rent change distribution in Turnover Contracts



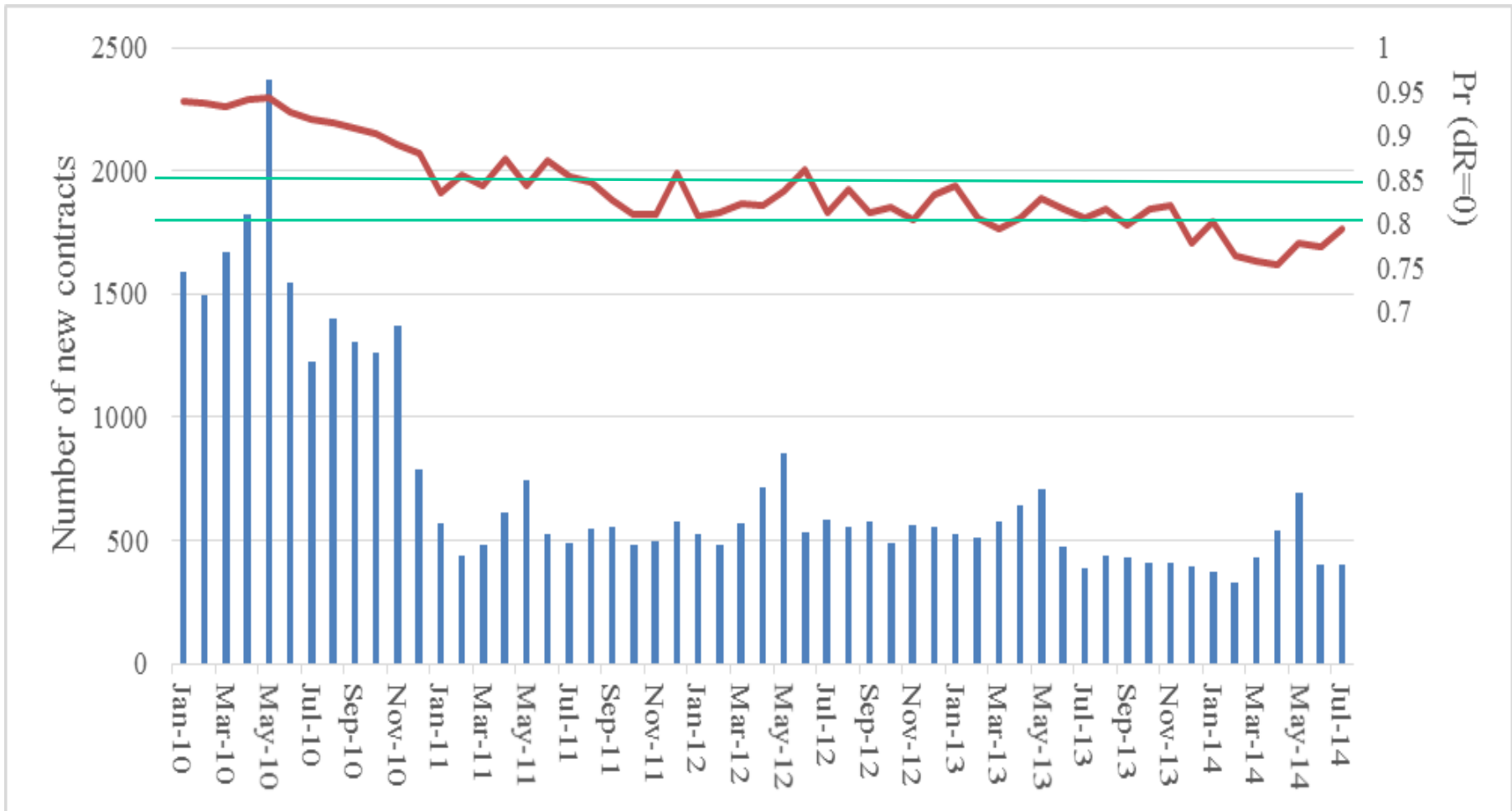
n=5,440

Monthly rent change distribution in Rollover Contracts

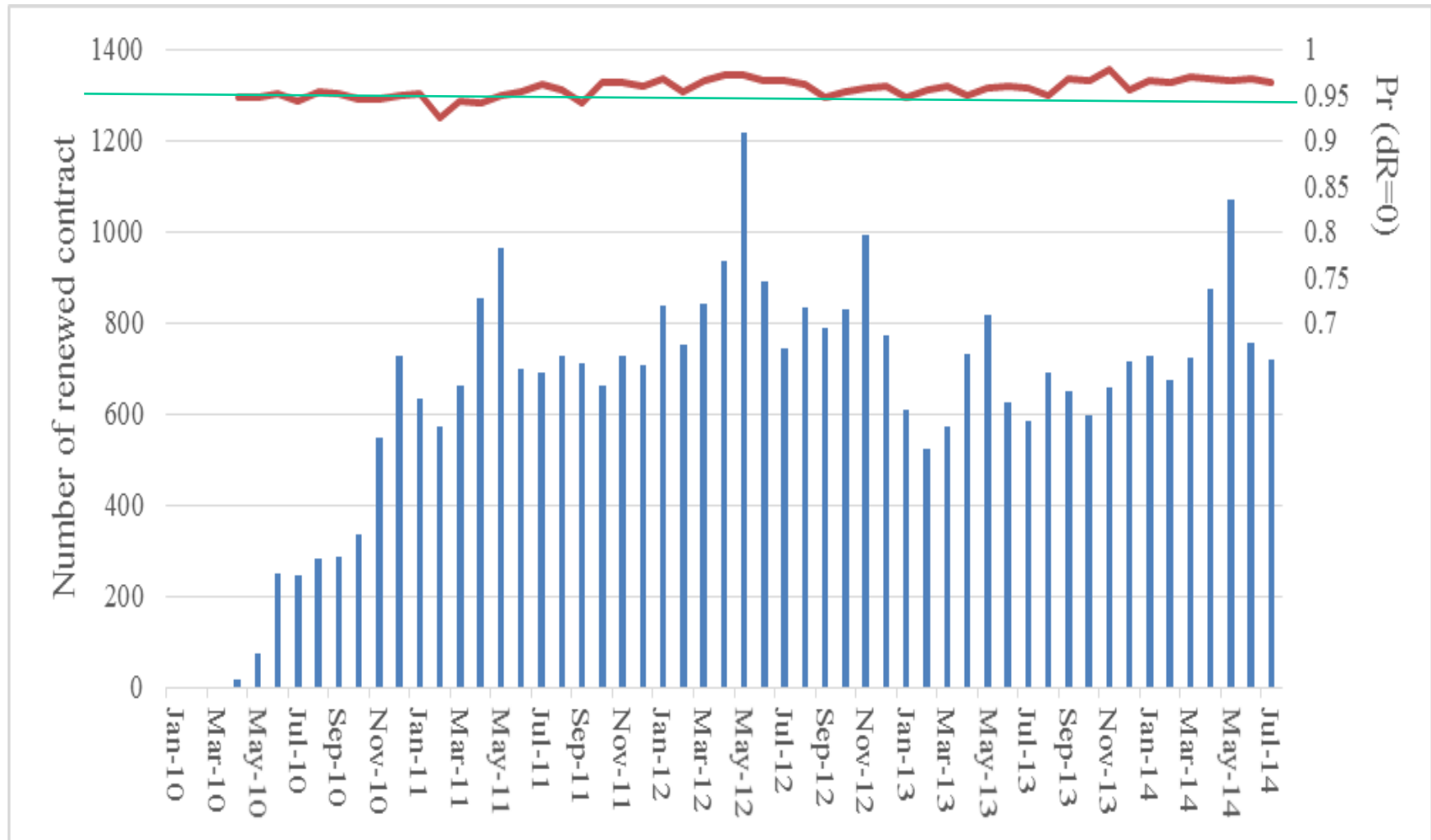


n=18,582,863

Probability of No Rent Adjustments in New Contracts : 2010-2014



Probability of No Rent Adjustments in Rollover Contracts : 2010-2014



State-Dependent or Time-Dependent Pricing: Caballero-Engel's definition of price flexibility

-Target Rent Level

-Price Gap

-Probability of rent adjustments

$$\Delta \log R_{it}^* = \Delta \xi_t + v_{it}$$

$$X_{it} \equiv \log R_{it-1} - \log R_{it}^*$$

$$\Lambda(x) \equiv \Pr(\Delta R_{it} \neq 0 \mid X_{it} = x)$$

Caballero-Engel(1993)
:Adjustment Hazard Function

$$\lim_{\Delta \xi_t \rightarrow 0} \frac{\Delta \log R_t}{\Delta \xi_t} = \int \Lambda(x)h(x)dx + \int x\Lambda'(x)h(x)dx$$

Caballero-Engel's
measure of price flexibility

Intensive margin

Extensive margin

Caballero-Engel(2007)

State-dependent or time-dependent pricing.

$$\Lambda(x) = \Pr(\Delta R_{it} \neq 0 \mid I_{it}^N = 1, X_{it} = x) \Pr(I_{it}^N = 1 \mid X_{it} = x)$$

$$+ \Pr(\Delta R_{it} \neq 0 \mid I_{it}^R = 1, X_{it} = x) \Pr(I_{it}^R = 1 \mid X_{it} = x)$$

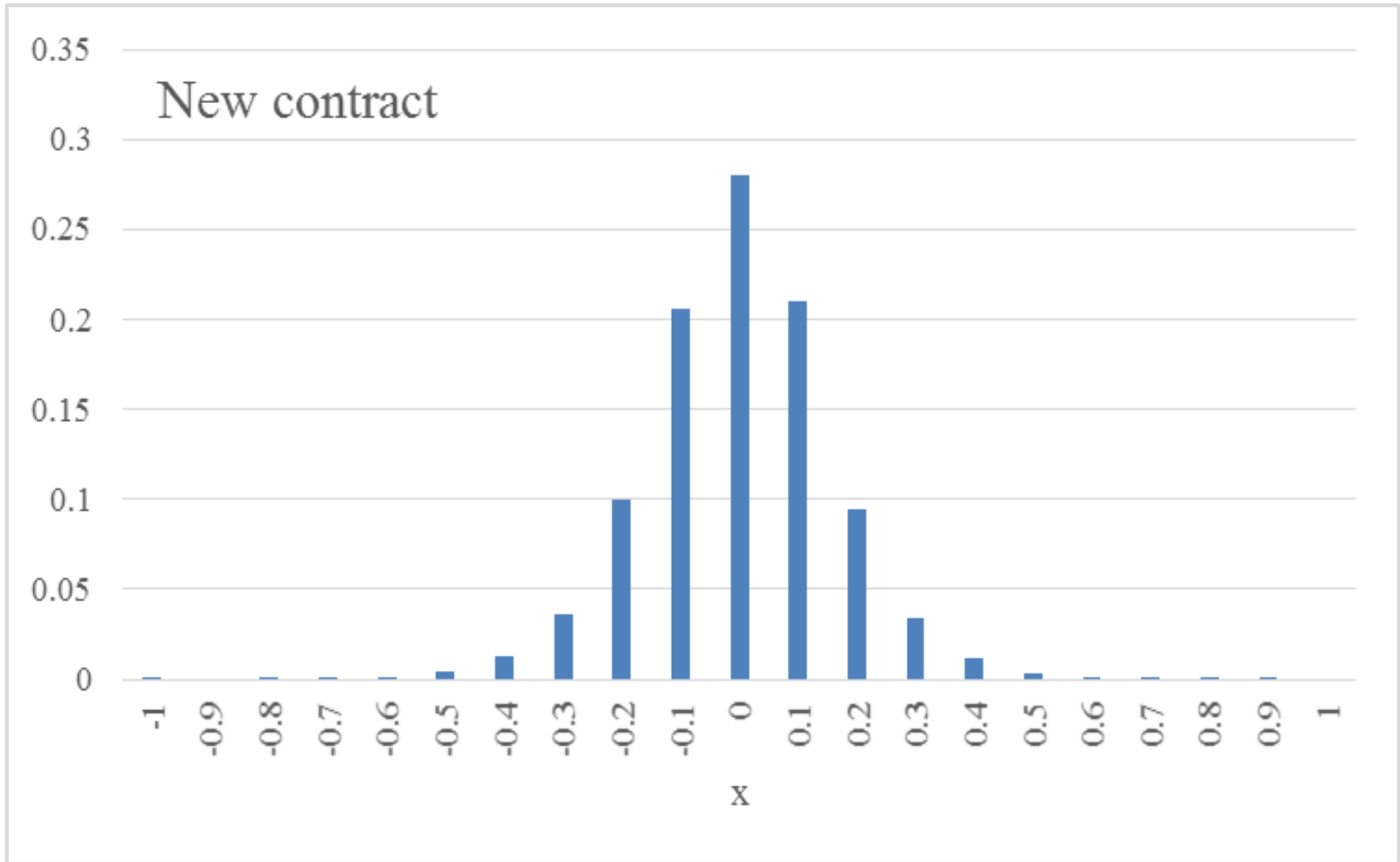
Estimated Coefficients in Hedonic Regressions

$$R_{it}^* = \begin{cases} R_{it} & \text{if } I_{it} = 1 \\ \hat{R}_{it} & \text{if } I_{it} = 0 \end{cases}$$

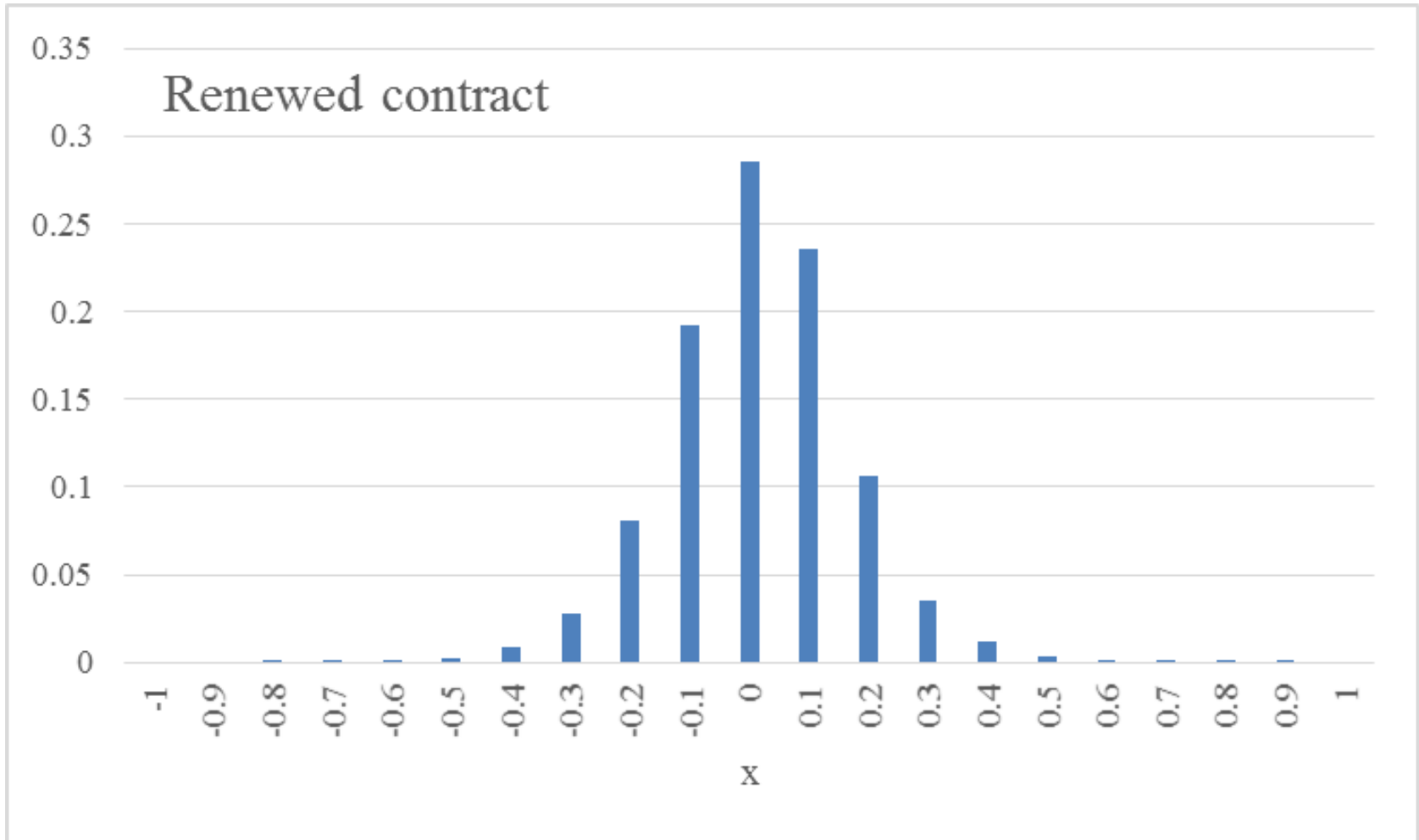
where $\hat{R}_{it} \equiv \hat{\beta}_t x_i + \hat{\gamma}_t TD_t$ $\hat{\beta}_t$ and $\hat{\gamma}_t$ are hedonic estimates

Estimation Window	Floor space	Age of building	Time to nearest station	Commuting time to CBD	Adjusted R ²	Number of observations
201001 - 201012	0.0188	-0.0109	-0.0087	-0.0058	0.917	17,697
201002 - 201101	0.0188	-0.0109	-0.0088	-0.0058	0.916	16,707
201003 - 201102	0.0188	-0.0109	-0.0089	-0.0059	0.917	15,670
201004 - 201103	0.0188	-0.0110	-0.0090	-0.0059	0.917	14,504
201005 - 201104	0.0188	-0.0110	-0.0092	-0.0058	0.916	13,303
201006 - 201105	0.0189	-0.0111	-0.0094	-0.0058	0.915	11,684
201007 - 201106	0.0189	-0.0112	-0.0096	-0.0060	0.914	10,667
201008 - 201107	0.0190	-0.0114	-0.0097	-0.0062	0.916	9,942
201009 - 201108	0.0189	-0.0115	-0.0095	-0.0065	0.918	9,099
201010 - 201109	0.0190	-0.0114	-0.0099	-0.0065	0.919	8,346
201011 - 201110	0.0191	-0.0113	-0.0104	-0.0067	0.922	7,571
201012 - 201111	0.0191	-0.0113	-0.0105	-0.0066	0.924	6,698
201101 - 201112	0.0191	-0.0114	-0.0104	-0.0067	0.924	6,490

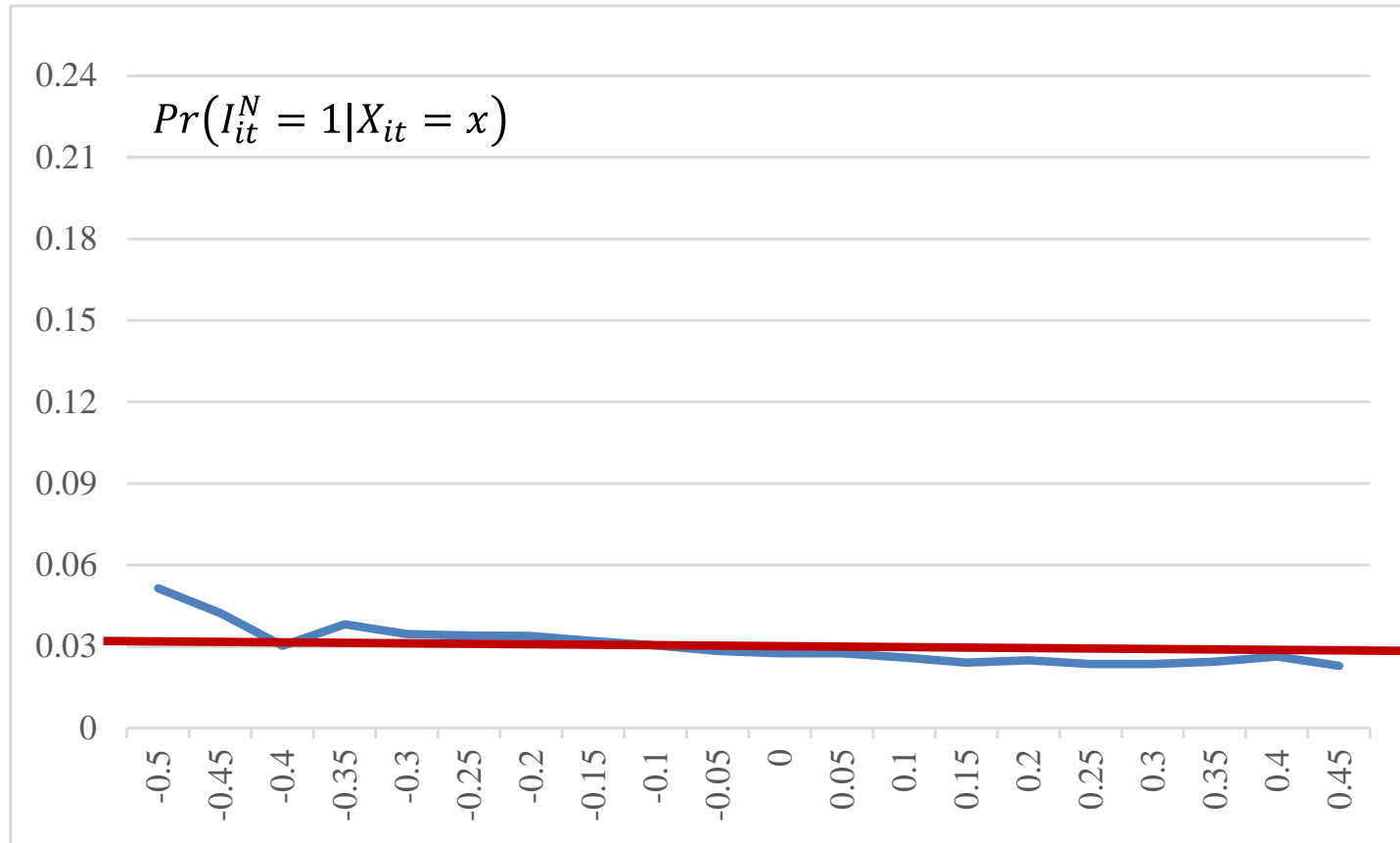
Distribution of Price Gap in New Contract



Distribution of Price Gap in Renewed Contract

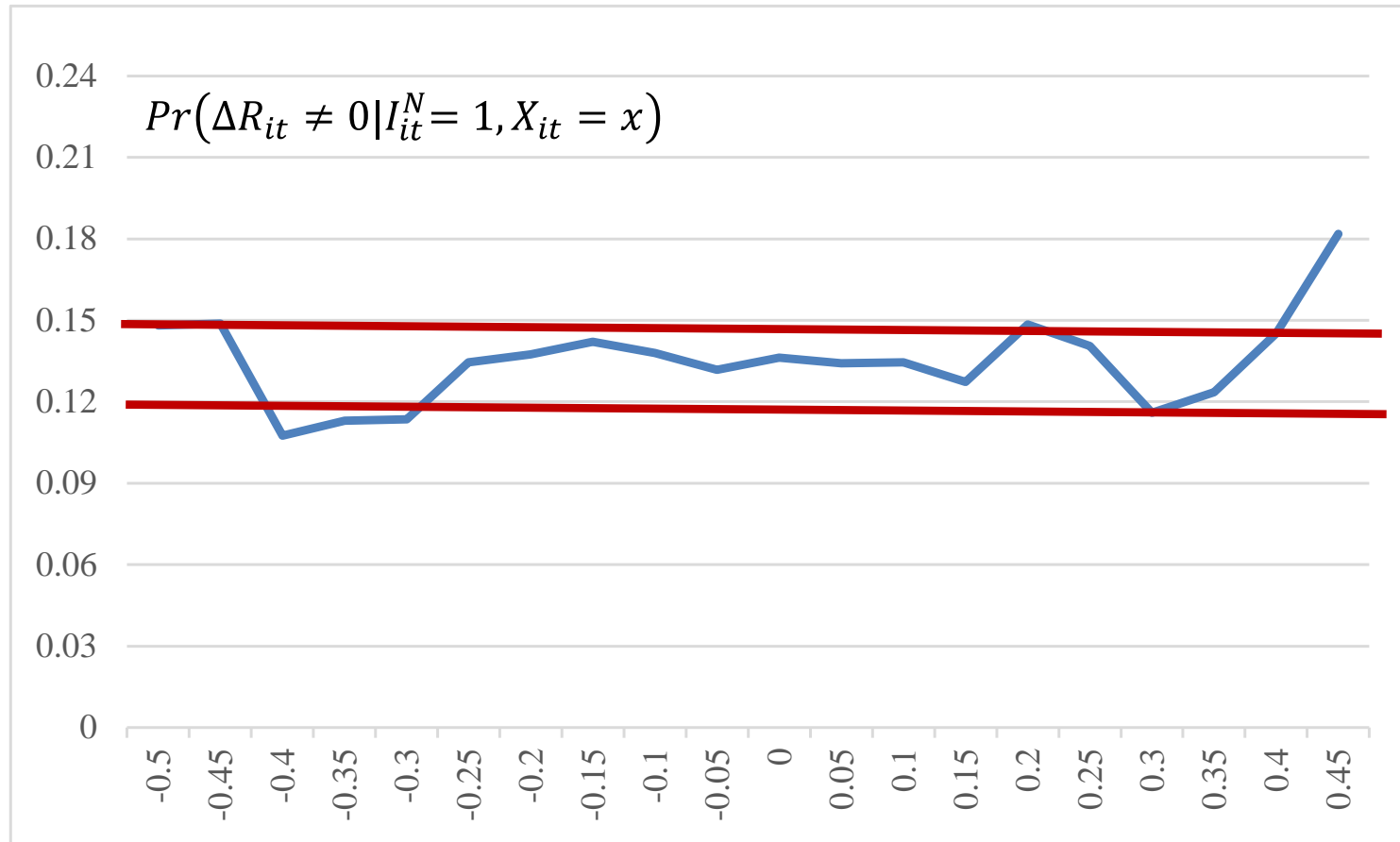


Adjustment Hazard Function for *Turnover Units*: Probability of Unit Turnover



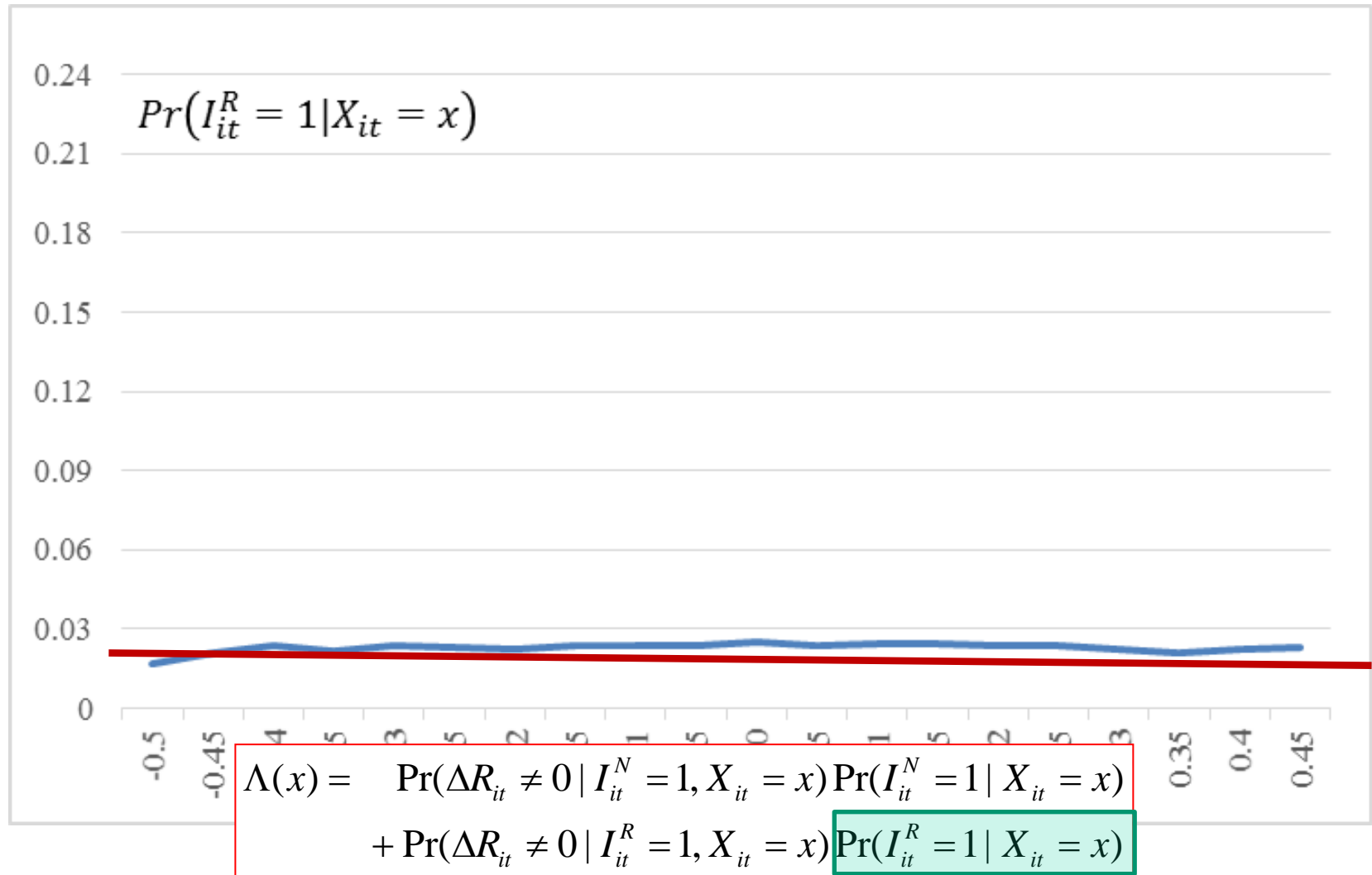
$$\Lambda(x) = \Pr(\Delta R_{it} \neq 0 | I_{it}^N = 1, X_{it} = x) \Pr(I_{it}^N = 1 | X_{it} = x) \\ + \Pr(\Delta R_{it} \neq 0 | I_{it}^R = 1, X_{it} = x) \Pr(I_{it}^R = 1 | X_{it} = x)$$

Adjustment Hazard Function for *Turnover Units*

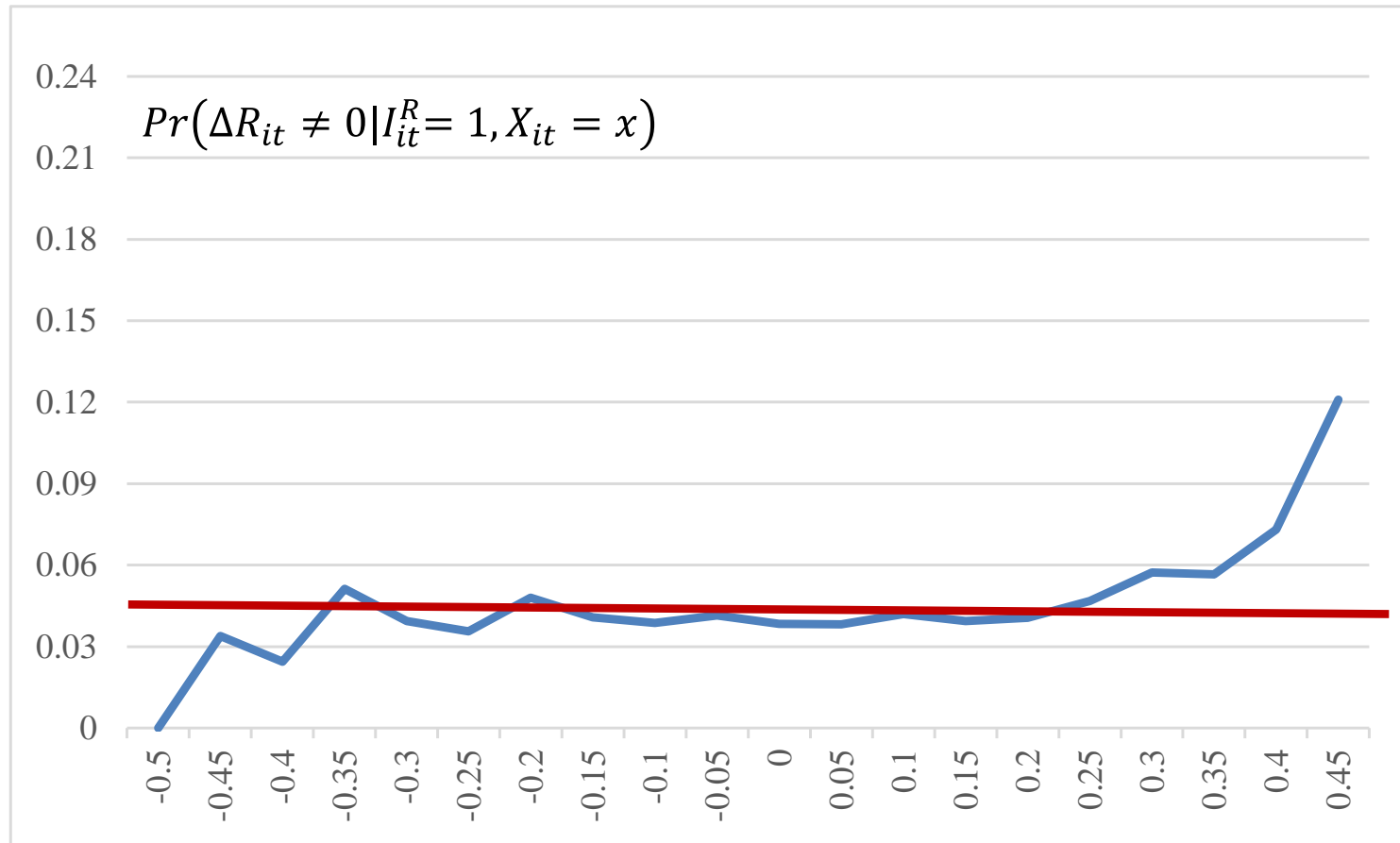


$$\Lambda(x) = \Pr(\Delta R_{it} \neq 0 | I_{it}^N = 1, X_{it} = x) \Pr(I_{it}^N = 1 | X_{it} = x) + \Pr(\Delta R_{it} \neq 0 | I_{it}^R = 1, X_{it} = x) \Pr(I_{it}^R = 1 | X_{it} = x)$$

Adjustment Hazard Function for *Rollover Units*



Adjustment Hazard Function for *Rollover Units*



$$\Lambda(x) = \Pr(\Delta R_{it} \neq 0 | I_{it}^N = 1, X_{it} = x) \Pr(I_{it}^N = 1 | X_{it} = x) + \Pr(\Delta R_{it} \neq 0 | I_{it}^R = 1, X_{it} = x) \Pr(I_{it}^R = 1 | X_{it} = x)$$

Adjustment Hazard Functions

	$x \in [-0.4, -0.2)$	$x \in [-0.2, 0.0)$	$x \in [0.0, 0.2)$	$x \in [0.2, 0.4)$
$\Pr(I_{it}^N = 1 X_{it} = x)$	0.035	0.030	0.026	0.024
$\Pr(I_{it}^R = 1 X_{it} = x)$	0.023	0.024	0.024	0.023
$\Pr(\Delta R_{it} \neq 0 I_{it}^N = 1, X_{it} = x)$	0.123	0.136	0.134	0.138
$\Pr(\Delta R_{it} \neq 0 I_{it}^R = 1, X_{it} = x)$	0.038	0.041	0.039	0.046
$\Lambda(x)$	0.005	0.005	0.004	0.004
$h(x)$	0.039	0.279	0.513	0.147

Intensive margin:

$$\int \Lambda(x)h(x)dx = 0.055$$

Extensive margin:

$$\int x\Lambda'(x)h(x)dx = 0.00580$$



**Caballero-Engel's measure
of price flexibility**

$$\lim_{\Delta \xi_t \rightarrow 0} \frac{\Delta \log R_t}{\Delta \xi_t} = 0.0607$$

Summary in “Sticky Housing Rent”

1. About ninety percent of the units in our dataset had no change in rents per year, indicating that rent stickiness is three times as high as in the US.
2. The probability of rent adjustment depends little on the deviation of the actual rent from its target level, suggesting that rent adjustments are not state dependent but time dependent.
3. These two results indicate that both intensive and extensive margins of rent adjustments are small, resulting in a slow response of the CPI rent to aggregate shocks.
4. The CPI inflation rate would have been higher by one percentage point during the bubble period, and lower by more than one percentage point during the period of bubble bursting, if the Japanese housing rents were as flexible as in the US.

→**Future work: Depreciation bias in Housing Rent in CPI.**

2. How should different countries construct residential property price indexes?

- In the wake of the release of this *Handbook*,
- **How should different countries construct residential property price indexes?**
- With the start of the *Residential Property Price Indices Handbook project*, the government of Japan set up an Advisory Board through the Ministry of Land, Infrastructure, Transport, and Tourism (MLIT) in 2012 and is proceeding with the development of a new Japanese official residential property price index.

Alternative Methods for Constructing Residential Property Price Indexes

- **Housing:** The location, history and facilities of each house are *different* from each other in **varying degrees**.
- Houses have “*particularity with few equivalents.*”
- **Quality-Adjustment Methodology** in Residential Property Price Indexes.
- → **Hedonic method and Repeat sales method**

The Standard Hedonic Regression Model : (1/3)

- House prices and property characteristics for periods $t = 1, 2, \dots, T$. The price of house i in period t , P_{it} , is given by a ***Cobb-Douglas function*** of the lot size of the house, L_i , and the amount of structures capital in constant quality units, K_{it} :
- $(1) P_{it} = P_t L_i^\alpha K_{it}^\beta$
- where P_t is the logarithm of the *quality adjusted house price index for period t* and α and β are positive parameters.
- Housing capital, K_{it} , is subject to *generalized exponential depreciation* so that the housing capital in period t is given by
- $(2) K_{it} = B_i \exp[-\delta A_{it}^\lambda]$
- where B_i is the floor space of the structure, A_{it} is the age of the structure in period t , δ is a parameter between 0 and 1, and λ is a positive parameter.

The Standard Hedonic Regression Model : (2/3)

- By substituting (2) into (1), and taking the logarithm of both sides of the resulting equation:
- (3) $\ln P_{it} = \ln P_t + \alpha \ln L_i + \beta \ln B_i - \beta \delta A_{it}^\lambda$.
- (4) $\ln P_{it} = d_t + \alpha \ln L_i + \beta \ln B_i - \beta \delta A_{it}^\lambda + \gamma \cdot x_i + v_{it}$
- where **$d_t \equiv \ln P_t$ is the logarithm of the constant quality population price index for period t** , P_t , γ is a vector of parameters associated with the vector of house i characteristics x_i , $\gamma \cdot x_i$ is the inner product of the vectors γ and x_i and v_{it} is an iid normal disturbance.

The Standard Hedonic Regression Model : (3/3)

- Running an OLS regression of equation (4) yields estimates for the coefficients on the time dummy variables, d_t for $t = 1, \dots, T$ as well as for the parameters α , β , γ , and δ .
- After making the normalization $d_1 = 0$, the series of estimated coefficients for the time dummy variables, d_t^* for $t = 1, \dots, T$, can be exponentiated to yield **the time series of constant quality price indexes, $P_t \equiv \exp[d_t^*]$ for $t = 1, \dots, T$** . Note that the coefficients α , β , γ , and δ are all identified in this regression model.

The Standard Repeat Sales Model : (1/3)

- The underlying price determination model is basically the same as in equation (4).
- (4) $\ln P_{it} = d_t + \alpha \ln L_i + \beta \ln B_i - \beta \delta A_{it}^\lambda + \gamma \cdot x_i + u_{it}$
- Suppose that house i is transacted twice, and that the transactions occur in periods s and t with $s < t$.
-
- (5) $\ln(P_{it}/P_{is}) = d_t - d_s - \beta \delta (A_{it}^\lambda - A_{is}^\lambda) + u_{it} - u_{is}$.
-
- Note that the terms that do not include time subscripts in equation (4), namely $\alpha \ln L_i$, $\beta \ln B_i$ and γx_i , all disappear by taking differences with respect to time, so that the resulting equation is simpler than the original one.

The Standard Repeat Sales Model : (2/3)

- Furthermore, assuming no renovation expenditures between the two time periods and no depreciation of housing capital so that $\delta = 0$, equation (5) reduces to:
 - (6) $\ln(P_{it}/P_{is}) = d_t - d_s + v_{it} - v_{is}$.
 - The above equation can be rewritten as the following linear regression model:
 - (7) $\ln(P_{it}/P_{is}) = \sum_{j=1}^T D_j^{its} d_j + v_{its}$
 - where $v_{its} \equiv v_{it} - v_{is}$ is a consolidated error term and D_j^{its} is a dummy variable that takes on the value 1 when $j = t$ (where t is the period when house i is resold), the value -1 when $j = s$ (where s is the period when house i is first sold) and D_j^{its} takes on the value 0 for j not equal to s or t .

The Standard Repeat Sales Model : (3/3)

- In order to identify all of the parameters d_j , a normalization is required such as $d_1 \equiv 0$. this normalization will make the house price index equal to unity in the first period.
- The *standard repeat sales house price indexes* are then defined by $P_t \equiv \exp[d_t^*]$ for $t = 1, 2, \dots, T$, where the d_t^* are the least squares estimators for the d_t .
- Thus the regression model defined by (5) does not require characteristics information on the house(except that information on the age of the house at the time of each transaction is required).

Problems for the hedonic method

- (i) there is an *omitted variable bias* (Case and Quigley 1991; Ekeland, Heckman and Nesheim 2004; Shimizu 2009);
- (ii) the assumption of *no structural change* is unrealistic (Case et al. 1991; Clapp et al. 1991; Clapp and Giaccotto 1992, 1998; Shimizu and Nishimura 2006, 2007, Shimizu, Takatsuji, Ono, and Nishimura 2007).

Problems for the repeat sales method

- (i) there is *sample selection bias* (Clapp and Giaccotto 1992);
- (ii) *heteroscedasticity*: error terms are likely to become larger when two transaction dates are further apart.
- Case and Shiller (1987, 1989) have proposed a model in which a GLS estimation is performed taking account of *heteroscedasticity*.
- (iii) the assumption that there are *no changes in property characteristics* and their parameters during the transaction period is unrealistic (Case and Shiller, 1987, 1989; Clapp and Giaccotto, 1992, 1998, 1999; Goodman and Thibodeau, 1998; Case et al. 1991). → **Depreciation and Renovation.**

Rolling Window Hedonic Regressions: Structural Change Adjustments to the Hedonic Index: (1/3)

- **Structural-change adjustment to hedonic :**
- We estimated hedonic model *considering structural change* by **Overlapping Period Hedonic Model**; OPHM, proposed by Shimizu et.al(1997), (2007) or **Rolling Window Hedonic Regressions**.
- OPHM may be more appropriate to estimate regression coefficients on the basis of a process of successive changes by taking a certain **length as the estimation “window”**, by shifting this period in a way of **rolling regressions**, in essence similar to moving averages.

Rolling Window Hedonic Regressions: Structural Change Adjustments to the Hedonic Index: (2/3)

- The sequence of final price levels P_t for the first τ periods is obtained by exponentiating the estimated time dummy parameters taken from the first Rolling Window regression; i.e., $P_t \equiv \exp[d_t^*]$ for $t = 1, 2, \dots, \tau$.
- The next Rolling window regression the data for periods 2, 3, ..., $\tau+1$ generates the new set of estimated time parameters, $d_2^{2*} \equiv 1, d_3^{2*}, \dots, d_{\tau+1}^{2*}$ and the new set of price levels P_t^2 for periods 2 to $\tau+1$, defined as $P_t^2 \equiv \exp[d_t^{2*}]$ for $t = 2, 3, \dots, \tau+1$.

Rolling Window Hedonic Regressions: Structural Change Adjustments to the Hedonic Index: (3/3)

- Now use only the last two price levels generated by the new regression to define the *final price level* for period $\tau+1$, $P_{\tau+1}$, as the period τ price level generated by the first regression, P_{τ} , times (one plus) the rate of change in the price level over the last two periods using the results of the second regression model; i.e., define $P_{\tau+1} \equiv P_{\tau} (P_{\tau+1}^2/P_{\tau}^2)$.
- The next step is to repeat the τ period regression model using the data for the periods 3, 4, ..., $\tau+2$ and obtain a new set of estimated time parameters, $d_3^{3*} \equiv 1, d_4^{3*}, \dots, d_{\tau+2}^{3*}$.
- $P_t^3 \equiv \exp[d_t^{3*}]$ for $t = 3, 4, \dots, \tau+2$ and update $P_{\tau+1}$ by multiplying it by $(P_{\tau+2}^3/P_{\tau+1}^3)$ so that the final price level for period $\tau+2$ is defined as $P_{\tau+2} \equiv P_{\tau+1} (P_{\tau+2}^3/P_{\tau+1}^3)$. Carry on with the same process until P_T has been defined.

Case-Shiller adjustment:

- Case and Shiller (1987) (1989) address the heteroskedasticity problem in the disturbance term by assuming that the variance of the residual v_{its} in (7) increases as t and s are further apart; i.e., they assume that $E(v_{its}) = 0$ and $E(v_{its})^2 = \xi_0 + \xi_1(t-s)$ where ξ_0 and ξ_1 are positive parameters.
- First, equation (7) is estimated, and the resulting squared disturbance term is regressed on $\xi_0 + \xi_1(t-s)$ in order to obtain estimates for ξ_0 and ξ_1 .
- Then equation (7) is reestimated by Generalized Least Squares (GLS) where observation i , $\ln(P_{it}/P_{is})$, is adjusted by the weight $[\xi_0^* + \xi_1^*(t-s)]^{1/2}$.
- The Case-Shiller *heteroskedasticity adjusted repeat sales indexes* are then defined by $P_t \equiv \exp[d_t^*]$ for $t = 1, 2, \dots, T$.

Previous researches about Age-adjusted RS

- Palmquist (1980) proposed a **two-stage method**: first obtain an independent estimate of the depreciation rate from the hedonic price.
- Case and Quigley (1991), Hill, et al (1997), and Englund, et al (1998) offered a similar idea but combined the hedonic and repeat sales regressions into a **hybrid model** for joint estimation.
- Chau, Wong, and Yiu (2005) , Shimizu, Nishimura and Watanabe(2010), Karato, Movshuk and Shimizu (2010) found another instrument to **separate the age and time effects. (cohort effect)**
- Cannaday, Munneke, and Yang (2005) , had to drop two age dummies arbitrarily in order to avoid perfect collinearity, although a high degree of collinearity still remains.

Age-adjustment to repeat sales index: (1/2)

- To take account of the depreciation effect, we go back to equation (5) and rewrite it as follows:
- (8) $\ln(P_{it}/P_{is}) = d_t - d_s - \beta\delta[(A_{is} + t - s)^\lambda - A_{is}^\lambda] + v_{its}$.
- Note that repeat sales indexes that do not include an age term (such as the term involving A_{is} on the right hand side of the above equation) will suffer from a downward bias.
- McMillen (2003) considered a simpler version of this model with $\lambda = 1$, so that the depreciation rate is constant over time. When $\lambda = 1$, (8) reduces to (9):
- (9) $\ln(P_{it}/P_{is}) = d_t - d_s - \beta\delta(t - s) + v_{its}$.
- Note that there is exact collinearity between $d_t - d_s$ and $t - s$, so that it is impossible to obtain estimates for the coefficients on the time dummies.

Age-adjustment to repeat sales index: (2/2)

- Once the d_t parameters have been estimated by maximum likelihood or nonlinear least squares (denote the estimates by d_t^* with d_1^* set equal to 0), then the *Shimizu, Nishimura and Watanabe repeat sales indexes* P_t are defined as $P_t \equiv \exp[d_t^*]$ for $t = 1, 2, \dots, T$.
- This is in sharp contrast with the hedonic regression model defined by (4), in which β appears not only as a coefficient of the age term but also as a coefficient on $\ln B_i$, so that β and δ are identified.
- **Shimizu, Nishimura and Watanabe (2010), Chau, Wong and Yui (2005), Wong, Chau, Karato, Shimizu(2013)**
“Separating the Age Effect from a Repeat Sales Index: Land and structure decomposition”

Alternative models: Hybrid Model

- Hybrid Model and Hedonic Model

$$y_{it} \equiv \ln P_{it} = X_i' \beta + \delta A_{it} + \alpha_t + \varepsilon_{it}$$

Hill, Knight, and Sirmans (1997) distinguished the time effect and age effect by refining Case and Quigley's (1991).

$$Y_i = y_{it} - y_{is} = \ln \frac{P_{it}}{P_{is}} = \tau_i \delta + \alpha_t - \alpha_s + v_i \quad (i=1, 2, \dots, N_R)$$

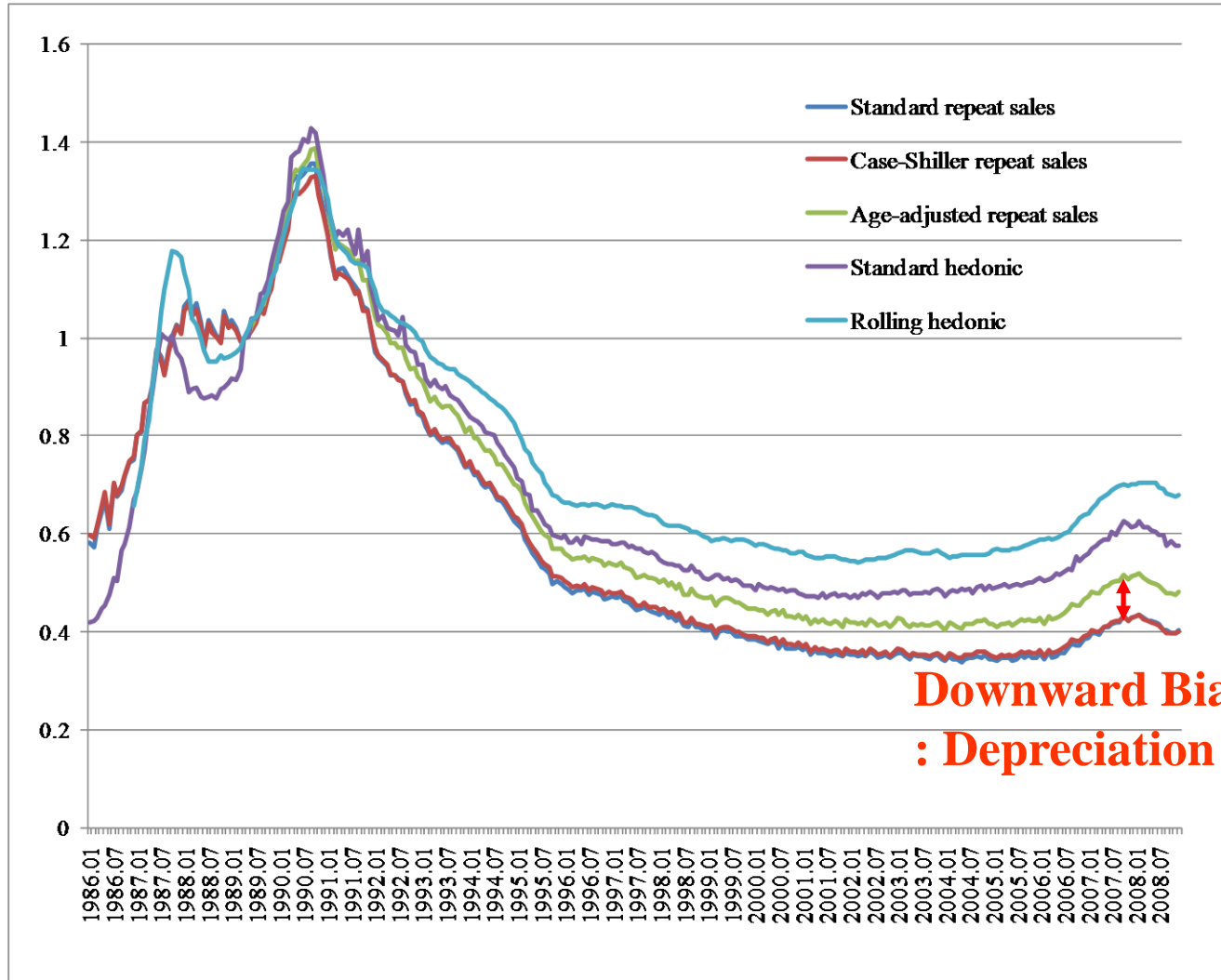
$$\begin{pmatrix} y \\ Y \end{pmatrix} = \begin{pmatrix} X & A & d \\ 0 & \tau & D \end{pmatrix} \begin{pmatrix} \beta \\ \delta \\ \alpha \end{pmatrix} + \begin{pmatrix} \varepsilon \\ v \end{pmatrix}$$

$$\ln P_{it} = X_i' \beta + \delta \frac{A_{it}^\lambda - 1}{\lambda} + \alpha_t + \varepsilon_{it}$$

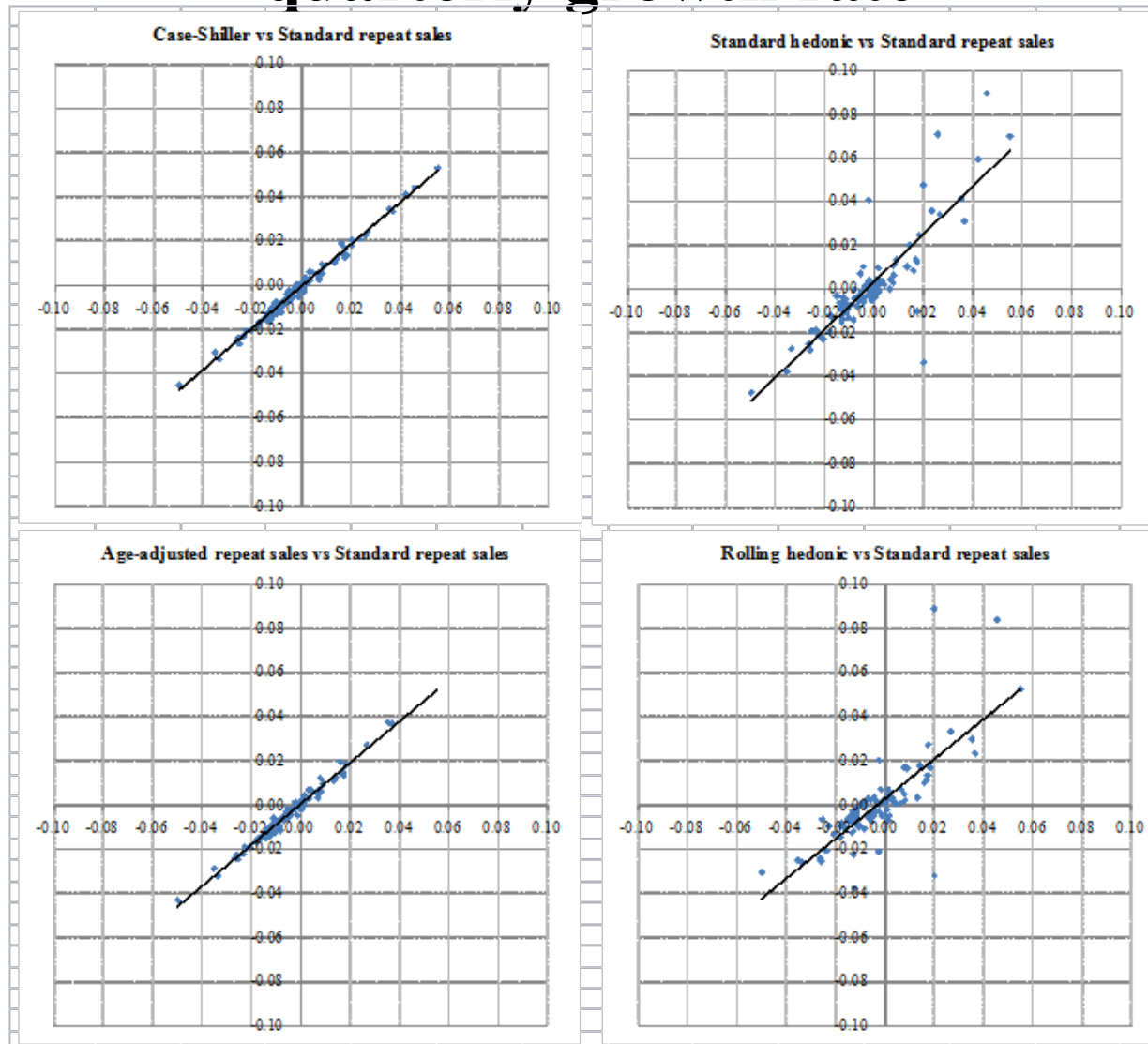
Alternative models: Builders Model

- *“At the national level, statistical agencies need to construct overall values of land and structures for the National Balance Sheets for the nation. If a user cost approach is applied to the valuation of Owner Occupied Housing services, it is necessary to have a decomposition of housing values into land and structures components **since structures depreciate while land does not.**”*
- Diewert, W. E. and C. Shimizu (2013), “Residential Property Price Indexes for Tokyo,” Discussion Paper 13-07, Vancouver School of Economics, University of British Columbia. *Macroeconomic Dynamics*, forthcoming.

Estimated five indices for condominiums



Comparison of the five indexes in terms of the quarterly growth rate



Pairwise Granger-causality tests

Condominiums

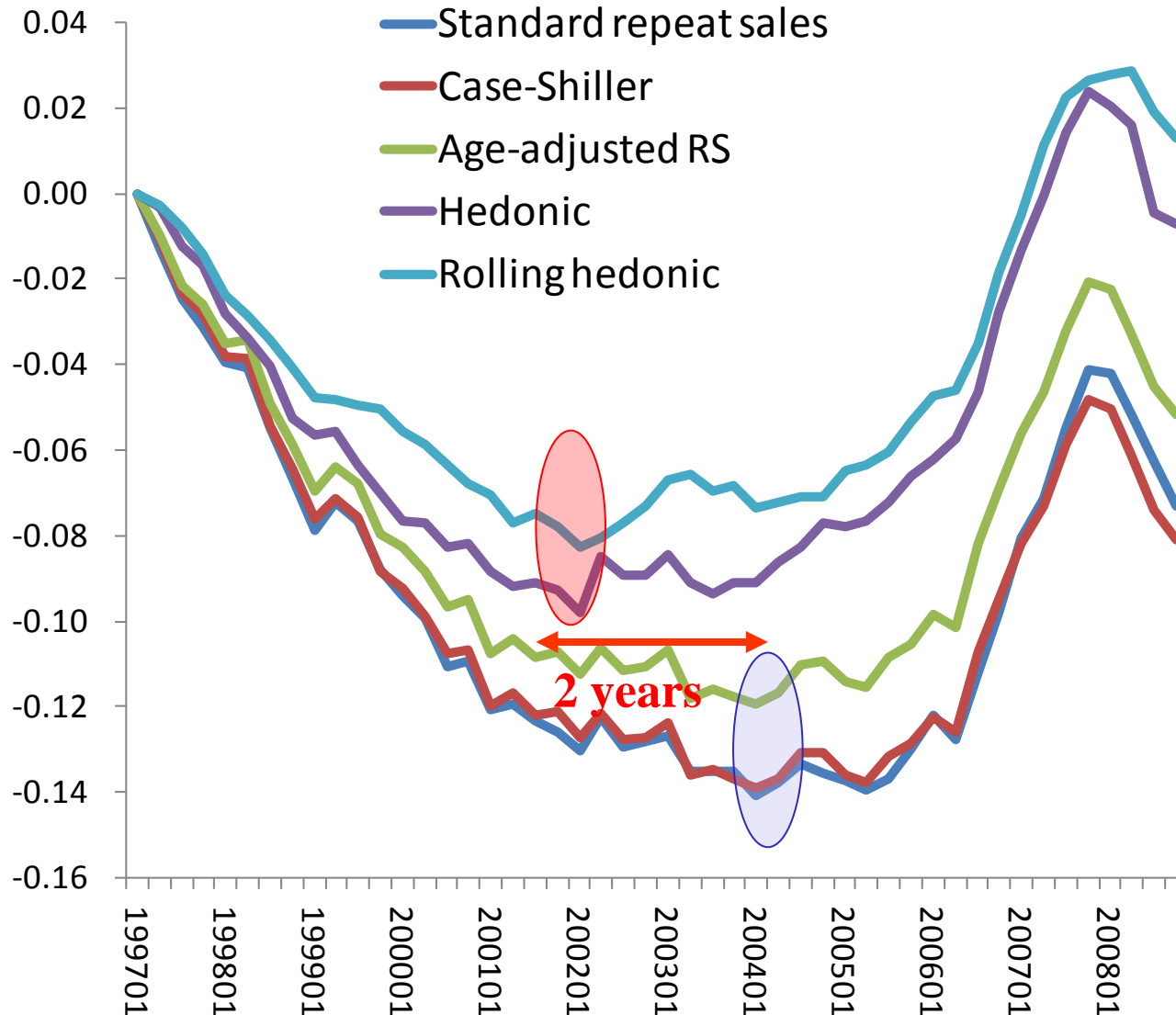
	Standard repeat sales	Case-Shiller repeat sales	Age-adjusted repeat sales	Standard hedonic	Rolling hedonic
Standard repeat sales		0.0120	0.0019	0.0039	0.0000
Case-Shiller RS	0.2018		n.a.	0.0398	0.0000
Age-adjusted RS	0.0568	n.a.		0.1258	0.0000
Standard hedonic	0.0004	0.0001	0.0000		0.0000
Rolling hedonic	0.0053	0.0082	0.0022	0.1528	

Single family houses

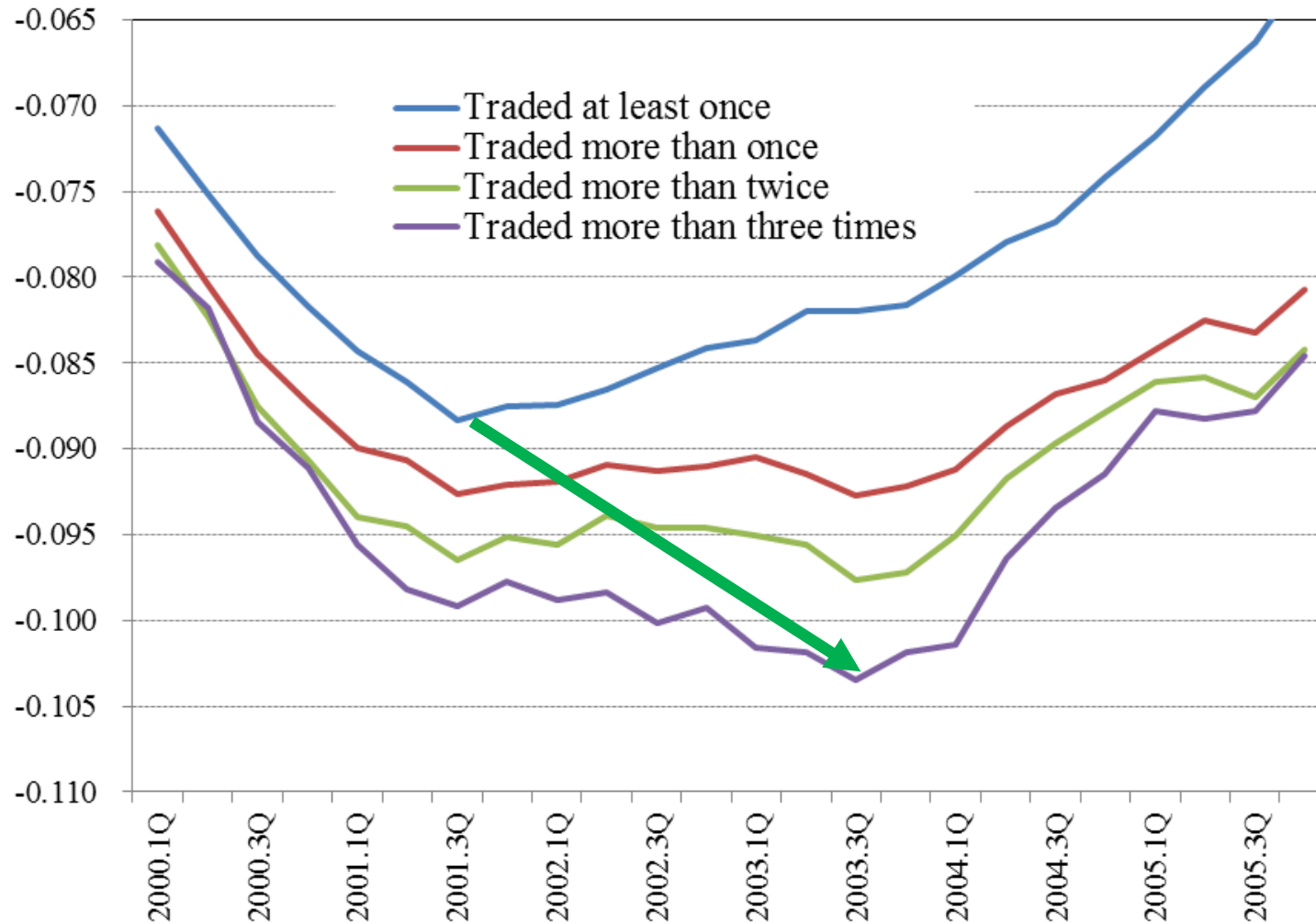
	Standard repeat sales	Case-Shiller repeat sales	Age-adjusted repeat sales	Standard hedonic	Rolling hedonic
Standard repeat sales		0.2726	0.4345	0.1919	0.0048
Case-Shiller RS	0.2397		n.a.	0.1810	0.0088
Age-adjusted RS	0.3275	n.a.		0.1962	0.0078
Standard hedonic	0.0028	0.0028	0.0027		0.0048
Rolling hedonic	0.0812	0.0784	0.0781	0.1089	

Note: The number in each cell represents the p-value associated with the null hypothesis that the variable in the row does not Granger-cause the variable in the column.

When did the condominium price hit bottom?



Hedonic indexes estimated using repeat-sales samples



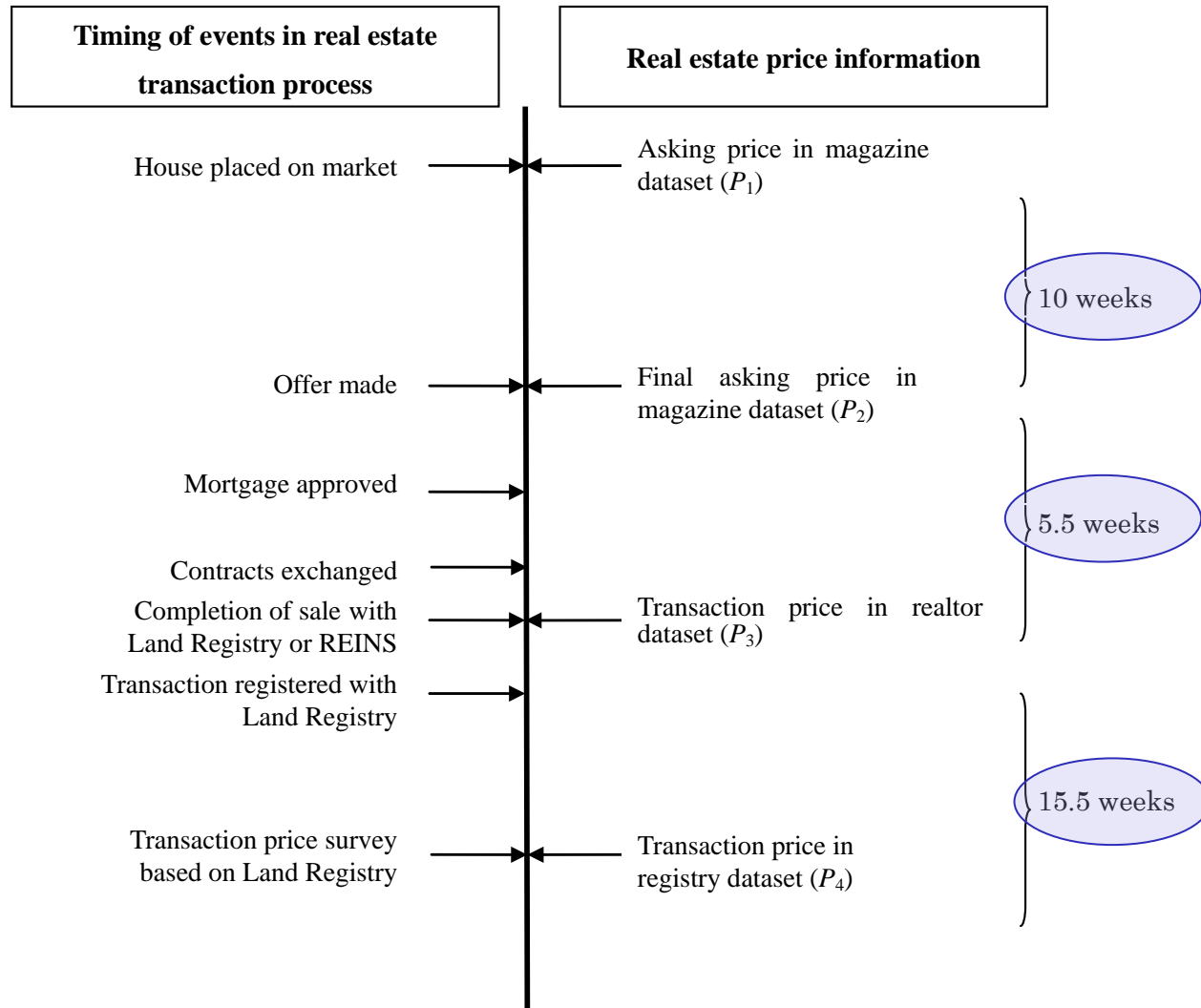
The Selection of Data Sources for the Construction of Housing Price Indexes

- Are house prices different depending on the stages of the buying/selling process?

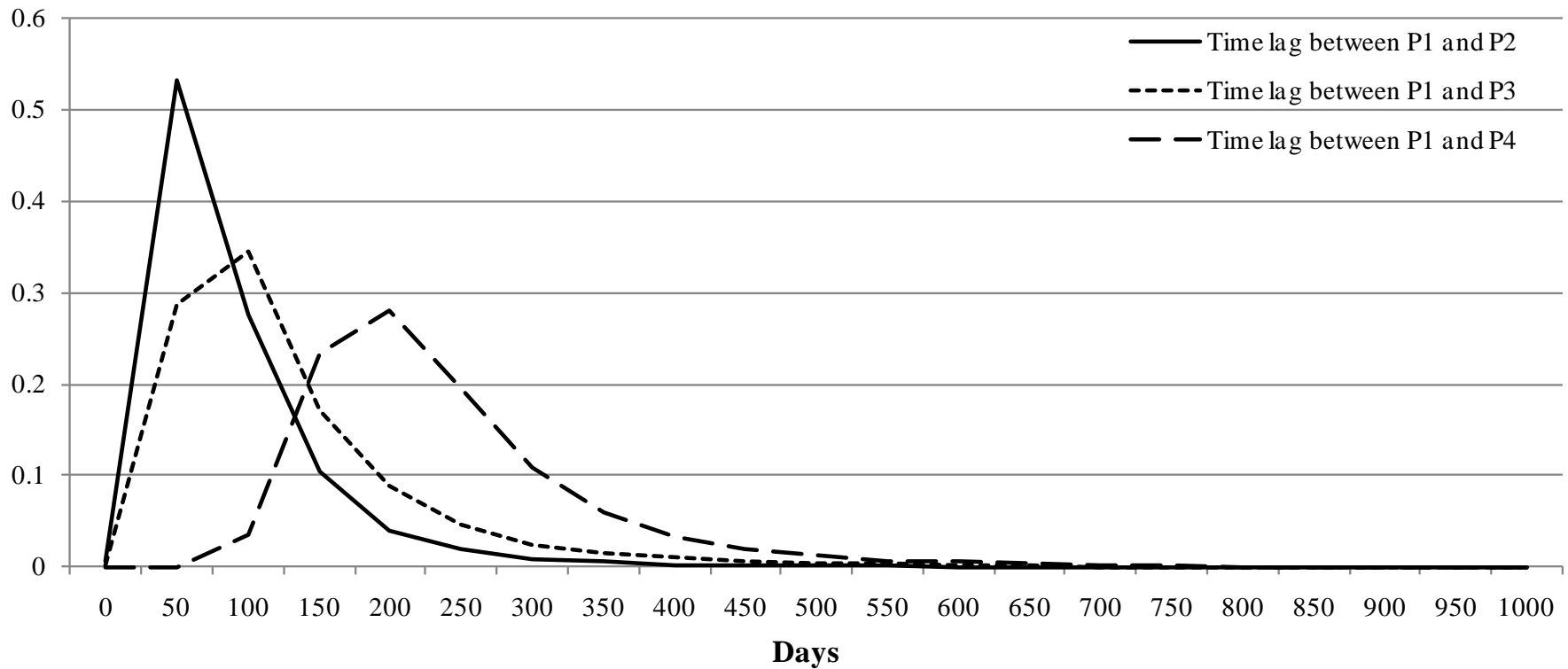


- We address this question by comparing the distributions of prices collected at different stages of the buying/selling process, including:
 - **(1) initial asking prices listed on a magazine,**
 - **(2) asking prices at which an offer is made by a buyer,**
 - **(3) contract prices reported by realtors after mortgage approval,**
 - **(4) registry prices.**

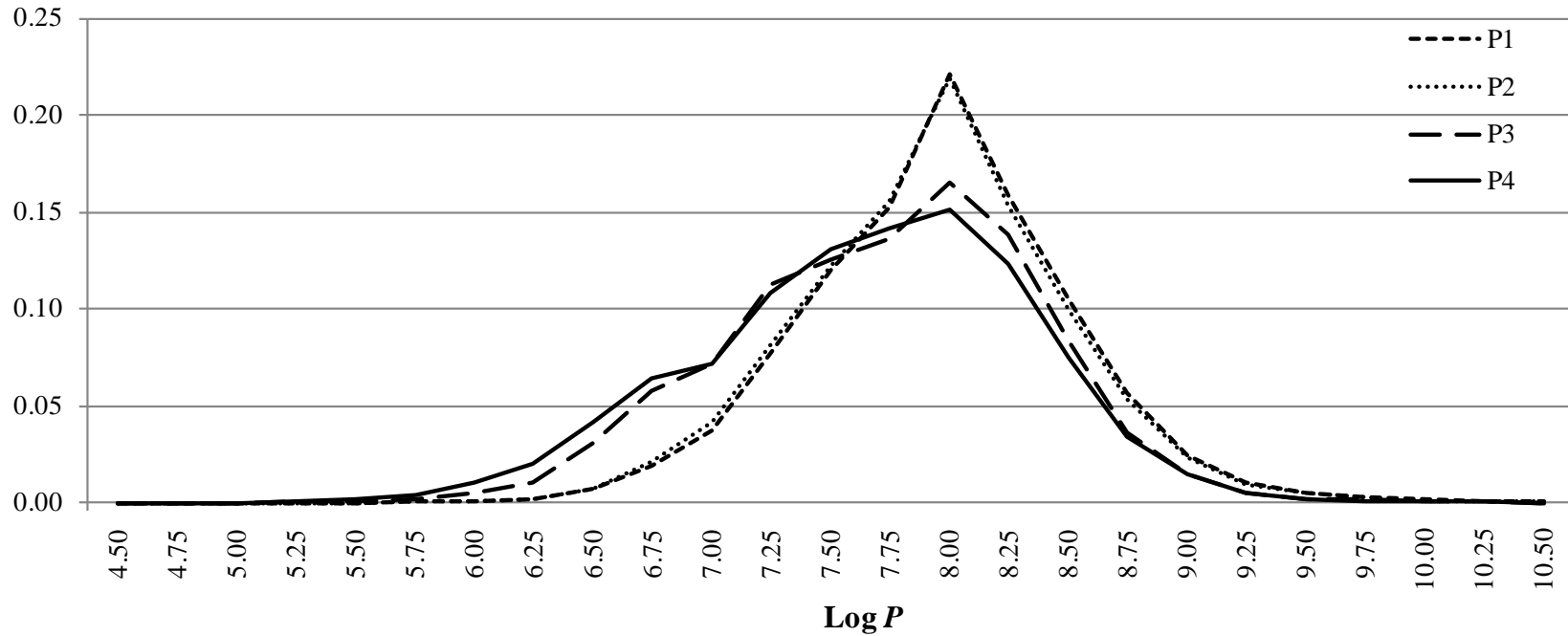
House purchase timeline



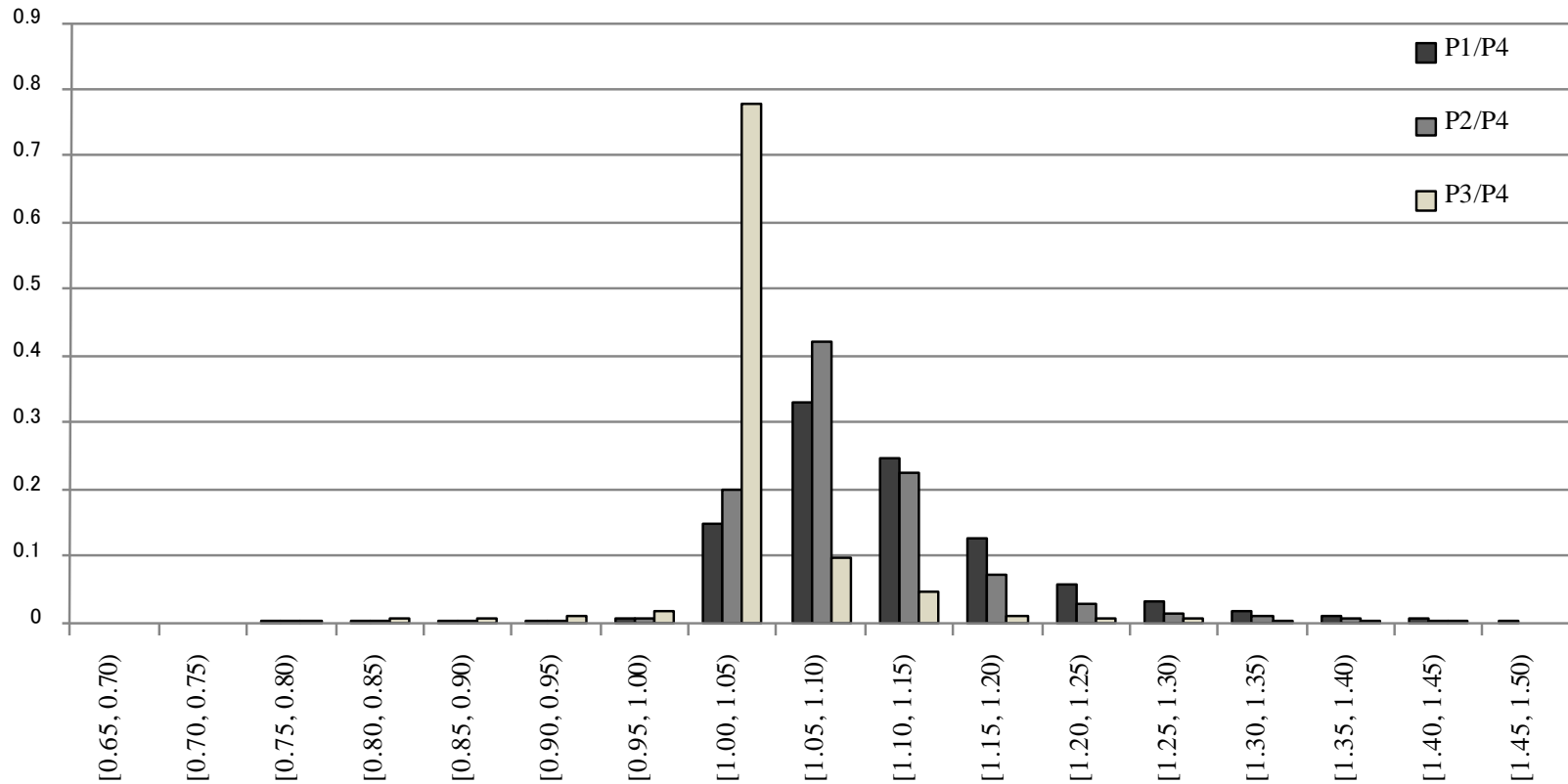
Intervals between events in the house buying/selling process



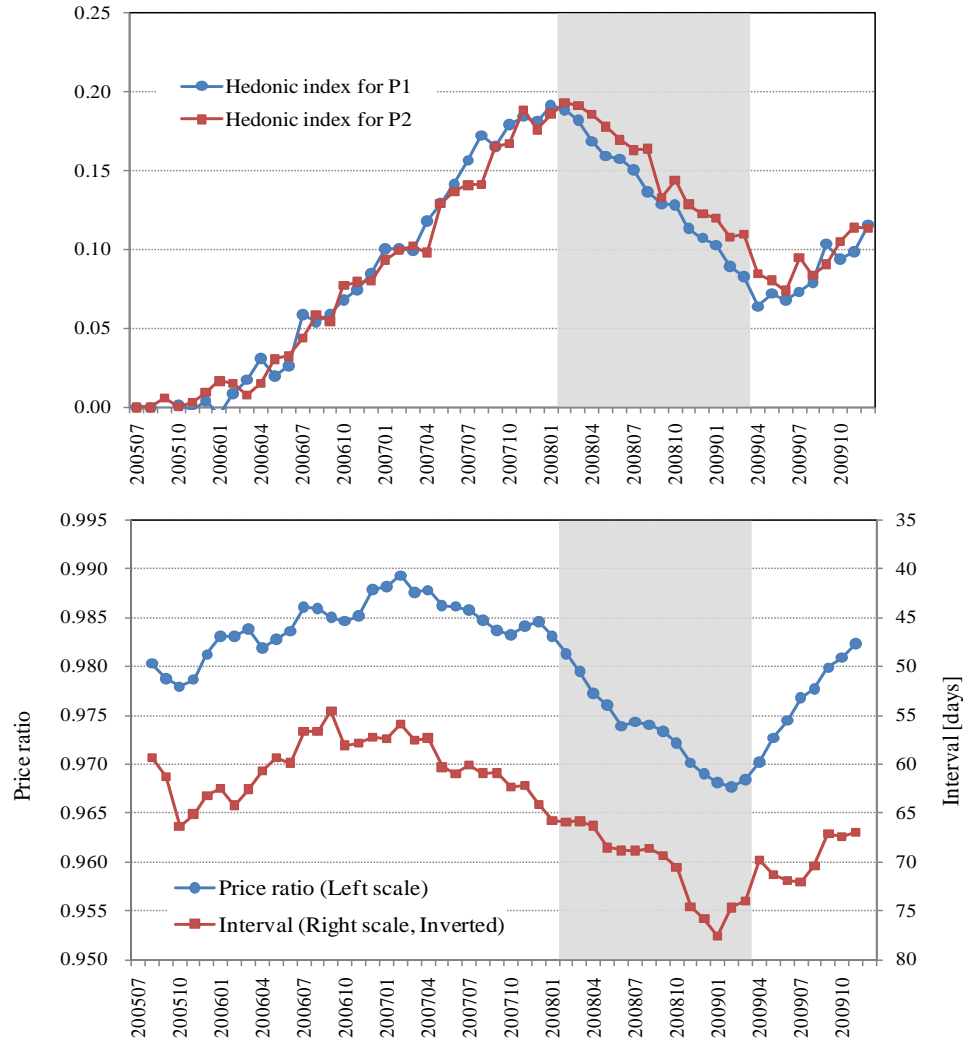
Price densities for P_1 , P_2 , P_3 , and P_4



Densities for relative prices



Fluctuations in the price ratio and the interval for P_1 and P_2



Turning point of two indicators

- The hedonic index for P1 declined by more than ten percent during the period between March 2008 and April 2009 indicated by the shaded area.
-
- During this downturn period, the price ratio exhibited a substantial decline, and more interestingly, changes in the price ratio preceded changes in the hedonic indexes.
- The price ratio started to decline in December 2007, three months earlier than the hedonic index for P1, and bottomed out in February 2009, two months earlier than the hedonic index for P1.

3. Conclusions:

- By the stickiness or rigidity of housing rent, CPI does not work well as a mirror of housing prices.
- In the wake of the release of the *Residential Property Price Indices Handbook*, the following questions arise:
- Q1: Do the different methods suggested in the Eurostat Handbook lead to different estimates of housing price changes?
- Q2: If the methods do generate different results, which method should be chosen.
- Q3: Which data source should be used for housing information?
- **RPPI to CPPI, Commercial Property Price Indexes.**

Response to Q1 and Q2: Methods for RPPI

- We find that there remains a *substantial discrepancy* between **the repeat sales measure and the hedonic measure**, even though we have made various adjustments to both indexes.
- Especially, *we find a substantial discrepancy in terms of turning points: the repeat sales measure tends to exhibit a delayed turn compared with the hedonic measure.*
 - For example, the hedonic measure of condominium prices hit bottom at **the beginning of 2002**, while the corresponding repeat-sales measure exhibits reversal only in **the spring of 2004**.
 - The lead-lag relationships between the two indices may come from *the omitted variable problem* in **the hedonic measure** and/or the problem of *non-random sampling* in **the repeat sales measure**.

Case of Japanese RPPI: Methods

- The question of which method is “*best*” remains open but the depreciation bias in the standard repeat sales method tends to lead us to prefer *hedonic methods*.
- **Given these results, the government of Japan decided to prepare an official residential property price index based on the hedonic method.**
- In particular, it has been determined that it will be estimated with the *rolling window hedonic method* proposed by Shimizu, Takatsugi, Ono and Nishimura (2010) and Shimizu, Nishimura and Watanabe (2010) and system development is underway.

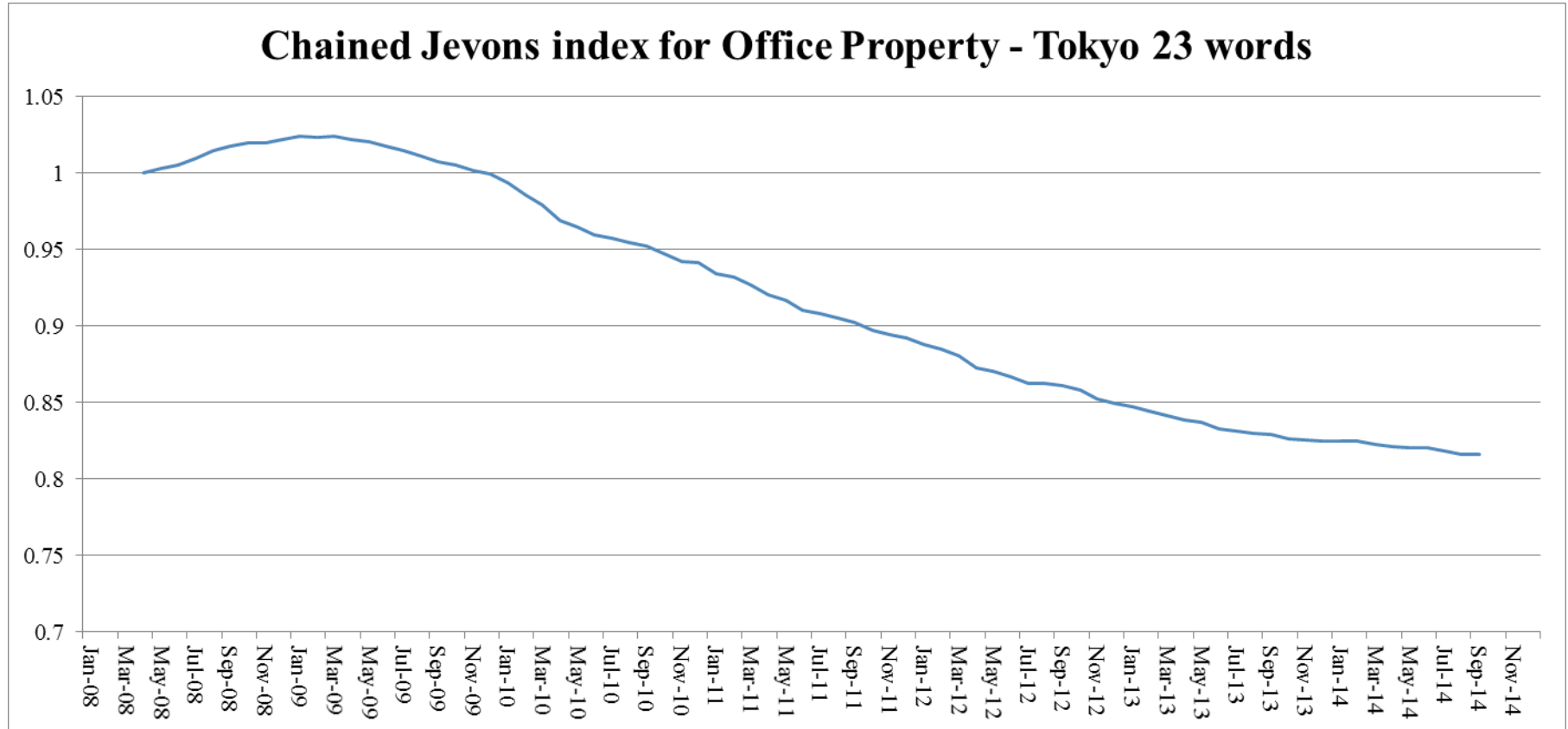
Annex:

Future Work:

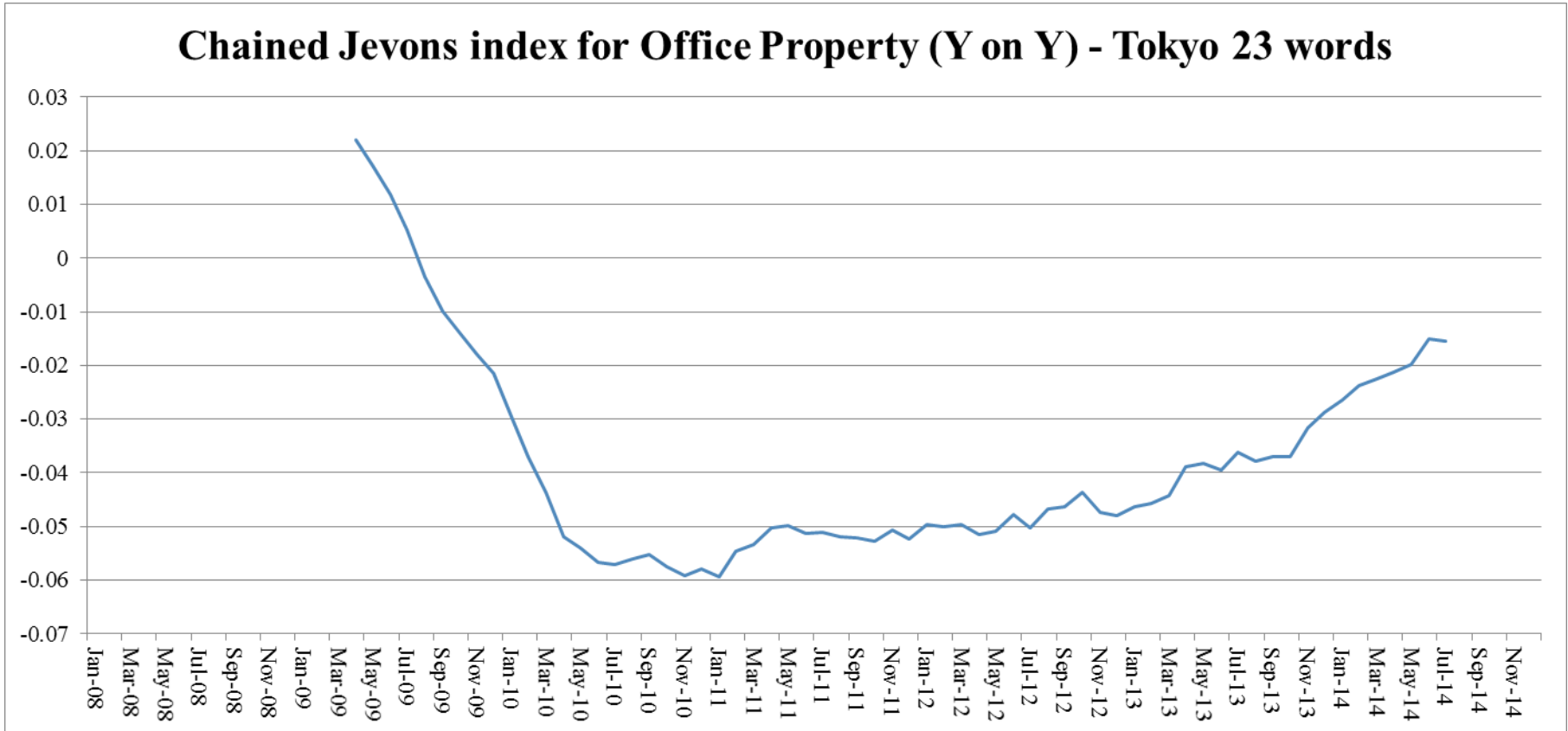
“Rigidity of *Office* Rent” and “*Office* Rent and PPI”

Sample period	April 2008 - September 2014					
Frequency	Monthly					
Geographical scope	Tokyo 23 wards					
Data type	Actual rent					
Coverage	New contract and renewed contract					
Provider	XYMAX					
No. of rooms	2,975					
Sample size	Whole sample		New contract		Renewed contract	
	115,529		3,774		3,056	
	mean	s.d.	mean	s.d.	mean	s.d.
Rent	2,119,543.2	3,815,365.2	2,051,398.8	4,260,888.4	2,144,011.2	3,014,086.9
m ²	405.8	415.4	383.3	395.6	424.0	434.9
m ² unit rent	5,048.5	15,568.8	5,132.8	18,285.7	4,815.4	3,300.3

Chained Jevons Index for Office Property



Chained Jevons Index for Office Property (Y on Y)



Comparison of Frequency of Rent Adjustments: Office vs. Residential

Office

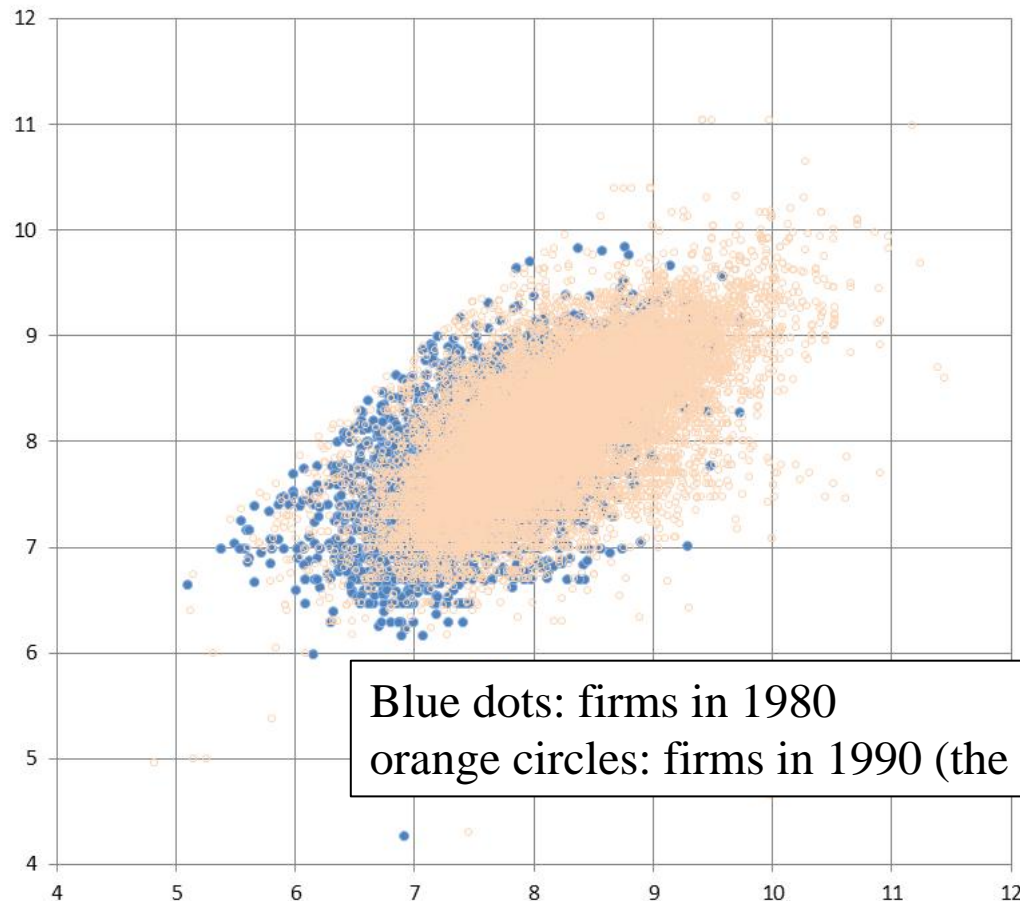
	Negative	Zero	Positive	Number of Observations	(Change)
Turnover units	226 (0.060)	3,032 (0.803)	516 (0.137)	3,774 (0.553)	742 (0.197)
Rollover units	677 (0.222)	2,214 (0.724)	165 (0.054)	3,056 (0.447)	842 (0.276)
All units	903 (0.132)	5,246 (0.768)	681 (0.100)	6,830 (1.000)	1,584 (0.232)

Residential

	Negative	Zero	Positive	Number of Observations	(Change)
Turnover units	4,473 (0.111)	34,958 (0.865)	967 (0.024)	40,398 (0.535)	5,440 (0.135)
Rollover units	1,169 (0.033)	33,667 (0.958)	303 (0.009)	35,139 (0.465)	1,472 (0.042)
All units	5,642 (0.075)	68,625 (0.908)	1,270 (0.017)	75,537 (1.000)	6,912 (0.092)

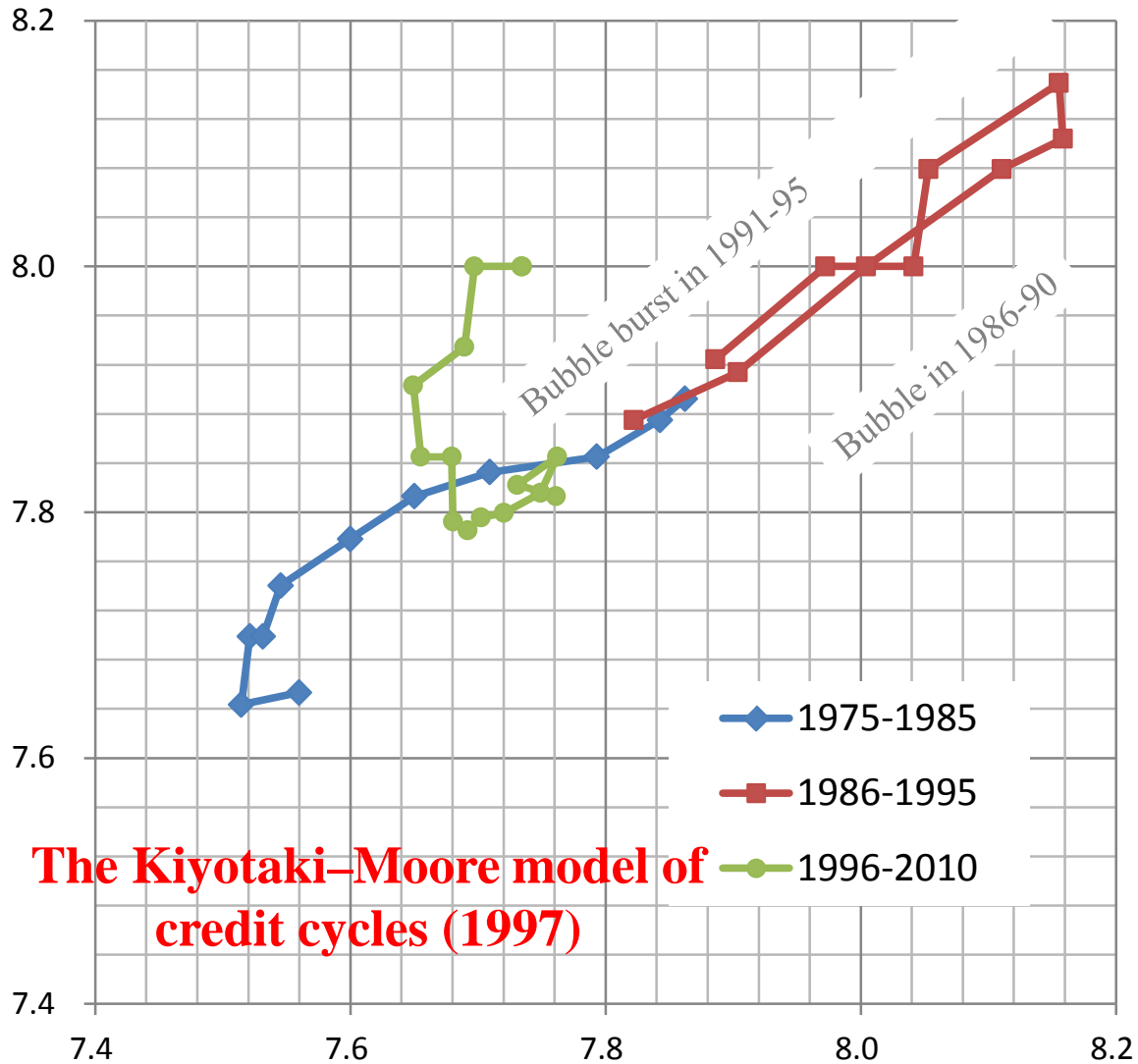
Scatter plot of property values and bank lending in 1980 and 1990

Bank lending
to firms [in log]



Value of land pledged as collateral by firms [in log]

Bank lending
to median firm
[in log]



The Kiyotaki-Moore model of credit cycles (1997)

Value of land pledged as collateral by median firm [in log]

Reestimates of CPI Inflation

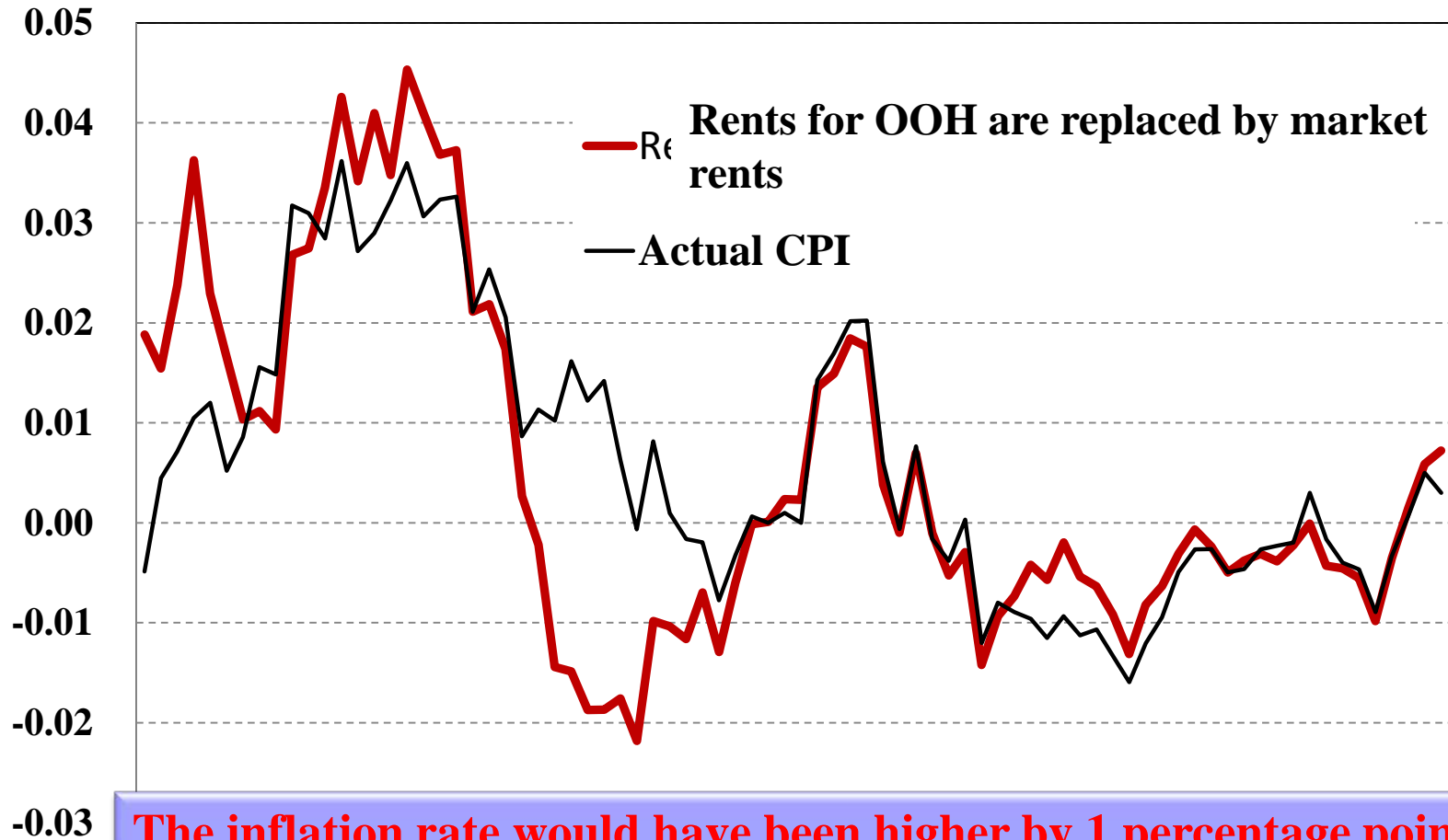
Imputed rent for owner occupied housing

The rental equivalence approach:

“The rental equivalence approach values the services yielded by the use of a dwelling by the corresponding *market value* for the same sort of dwelling for the same period of time (if such a rental value exists).” (Diewert and Nakamura 2008)

“If someone were to rent your home *today*, how much do you think it would rent for monthly, unfurnished and without utilities?” (BLS)

Alternative measures of CPI inflation



The inflation rate would have been higher by 1 percentage point during the bubble period and lower by 2 percentage points during the post-bubble period, if the market price were used in estimating the rent index for owner occupied housing


An age-adjusted repeat sales model

- Value (P) is the sum of land value (L) and structure value (S).
The value of a new property is:
- $P(0) = L + S(0)$
- After A periods, the property reaches age A.
- $P(A) = L + S(0) \times (1 - \delta A)$
- $P(A) = L + S(0) \times [1 - \delta g(A)]$

Depreciation for Property

***Ratio of new structure value to new property value**

$$\bullet \quad \frac{P(A) - P(0)}{P(0)} = - \frac{S(0)}{P(0)} \delta g(A)$$



 $R(0)$

- (a) The same δ (e.g. building technology) for all structures, properties in a high land value area would depreciate more slowly than those in a low land value area.
- (b) If the structure depreciates at a constant rate, the property's depreciation rate is likely to be time-varying because structure and land values do not move at the same pace.

Price Structure

- Hedonic price equation:

- Age term:
$$\frac{P(A)-P(0)}{P(0)} = -\frac{S(0)}{P(0)} \delta g(A)$$

- $$\ln P_{it} = X_i \beta + \alpha_t - \delta R_t g(A_{it}) + \varepsilon_{it}$$

- A non-linear depreciation pattern of the structure, the age function, $g(A_{it})$, adopts the Box-Cox transformation as shown in Equation 6

- $$g(A_{it}) = \frac{A_{it}^\lambda - 1}{\lambda}$$

An age-adjusted repeat sales model

- Property i is sold twice at time s and t (where $t > s$) and there is no change in property attributes between the sales.
- Age-adjusted repeat sales model can be derived from the $(t-s)^{\text{th}}$ difference of Equation 5:

- $$\ln \left(\frac{P_{it}}{P_{is}} \right) = (\alpha_t - \alpha_s) - \delta [R_t g(A_{it}) - R_s g(A_{is})]$$
- $$+ (\varepsilon_{it} - \varepsilon_{is})$$

- $$(t) \quad \ln P_{it} = X_i \beta + \alpha_t - \delta R_t g(A_{it}) + \varepsilon_{it}$$
- $$(s) \quad \ln P_{is} = X_i \beta + \alpha_s - \delta R_s g(A_{is}) + \varepsilon_{is}$$