

# Unsmoothing Real Estate Returns: A Regime-switching Approach

Warapong Wongwachara

University of East Anglia

(Joint with Colin Lizieri and Stephen Satchell  
University of Cambridge)

Bank of Thailand  
September 2011

# Reported Real Estate Return

- Appraisal-based return as opposed to market-traded
- *Smoothing* – Temporal aggregation and lagging effects
  - Serial correlation
  - Dampened volatility
- Real estate perceived as “safe” investment
- Implication in asset allocation / performance measurement

# Unsmoothing Appraisal-based Return

- Conventional unsmoothing methodology: Geltner (1991, 1993), Fisher et al. (1994), Cho et al. (2003), Booth & Marcato (2004), Marcato & Key (2007)
- Fairly successful in that it raises volatility of (unsmoothed) real estate return
- Yet, real estate shown to have significantly better risk hedging characteristics than other asset classes (see e.g. Hudson-Wilson et al. 2003, Worzala & Sirmans 2003, Bond et al. 2007)
- Unsmoothed return still too smooth
- We found that this was only half the truth!

# Our Regime-switching Approach

- Conventional method:
  - True return process
  - Smoothing equation
- Not completely satisfactory as it ignores non-linearity in performance data
- Our approach based on Threshold Autoregressive (TAR) model (Tong 1978, 1990)
  - Switching return: high volatility in “bad regime”, low in “good regime”
  - Switching smoothing: behavioral changes
- Results = direct and important practical implication

# Outline

- 1 Base model
- 2 Regime-switching models
- 3 Estimation and Implementation
- 4 Results and Discussion
- 5 Conclusion

# The Base Model

- Measurement equation (Blundell & Ward 1987)

$$r_t^* = \alpha r_{t-1}^* + (1 - \alpha)r_t$$

where  $r_t^*$  = observed (smoothed) return,  $r_t$  the “true” return, and the smoothing coefficient  $\alpha \in (0, 1)$

- Given  $\alpha$ , can calculate the unsmoothed return by

$$r_t = \frac{1}{1 - \alpha} (r_t^* - \alpha r_{t-1}^*)$$

- Can also show that the “unsmoothed” variance is strictly increasing in  $\alpha$

# The Base Model

## Continued

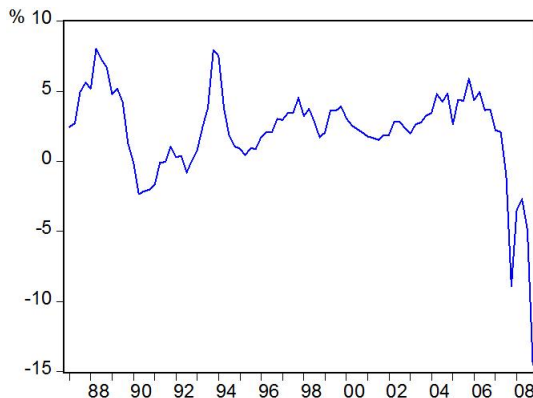
- So far true return process irrelevant in a sense that  $r_t$  may be obtained once  $\alpha$  is known
- When  $\alpha$  unknown, further information on  $r_t$  required
- Practical approach: *iid* return, hence  $\hat{\alpha} = \hat{\rho}_1$
- Return process (State equation)

$$r_t = \gamma + \phi r_{t-1} + \varepsilon_t, \quad \varepsilon_t \sim iid(0, \sigma_\varepsilon^2)$$

- Most studies assume  $|\phi| < 1$
- Actual data probably not stationary

# Non-linearity in Real Estate Return

Figure: Quarterly log-returns on IPD Index (Q4 1986 - Q4 2008)



Source: Investment Property Databank (IPD)

# Regime-switching Approach

## ■ Smoothing equation

$$r_t^* = \alpha_t r_{t-1}^* + (1 - \alpha_t) r_t$$

## ■ State equation

$$r_t = \gamma_t + \phi_t r_{t-1} + \varepsilon_t, \quad \varepsilon_t \sim iid(0, \sigma_\varepsilon^2)$$

## ■ Regime-switching parameters

$$\alpha_t = \begin{cases} \alpha_1 & , \quad z_{1t-1} > c_1 \\ \alpha_2 & , \quad z_{1t-1} \leq c_1 \end{cases} \quad (\gamma_t, \phi_t) = \begin{cases} (\gamma_1, \phi_1) & , \quad z_{2t-1} > c_2 \\ (\gamma_2, \phi_2) & , \quad z_{2t-1} \leq c_2 \end{cases}$$

where  $z_{it}$  = an exogenous observable regime indicator

# Economic Identification

- Switching in the smoothing equation
  - Probably due to behavioural shifts of the appraisal agency
  - Different arrival rates of new information in “good” and “bad” regimes
  - Our model more generalised than that of Chaplin (1997)
- Switching return
  - Tied to changes in the macroeconomic environment
  - Time-varying volatility
- Open to a large number of plausible regime indicators
- Not necessarily the same regime indicator for smoothing and return

# Potential Regime Indicators

## Drivers of UK Real Estate Return

- Three-month LIBOR rate (end of period)
- GDP growth (nominal)
- SA employment
- Inflation – RPI excluding mortgage interest
- Log-return on FT All Share Total Return index
- Initial Yield index (rent to capital value)
- USD-GBP spot rate

# Degree of Restrictions

- Conventional AR (AR-AR):  $\alpha_t = \alpha, \gamma_t = \gamma, \phi_t = \phi$
- Switching return (AR-TAR):  $\alpha_t = \alpha$
- Switching behaviour (TAR-AR):  $\gamma_t = \gamma, \phi_t = \phi$
- Co-switching (TAR-TAR)

# Estimation of AR-AR

- LS which iterates between the two equations (Cochrane-Orcutt-type)
- Implied AR(2) in reported return

$$r_t^* = (1 - \alpha)\gamma + (\alpha + \phi)r_{t-1}^* - \alpha\phi r_{t-2}^* + v_t$$

with  $v_t = (1 - \alpha)\varepsilon_t$

- Given  $(\gamma, \phi)$ , can estimate  $\alpha$  by LS
- Then use  $\hat{\alpha}$  to obtain  $r_t$
- Use this unsmoothed return to estimate  $(\gamma, \phi)$
- Continue until coefficients converge, i.e. differ than previous value by less than 0.01
- Consistent (though not efficient)

# Estimation of TAR-TAR

- Straightforward extension of the previous technique
- Implied TAR(2)

$$(1 - \phi_t L)(1 - \alpha_t L)r_t^* = (1 - \alpha_t)(\gamma_t + \varepsilon_t)$$

- Make use of an indicator function  $I_{it} = \mathbf{1}(z_{it} > c_i)$  such that

$$\alpha_t = \alpha_1 I_{1t-1} + \alpha_2 (1 - I_{1t-1})$$

$$(\gamma_t, \phi_t) = (\gamma_1, \phi_1) I_{2t-1} + (\gamma_2, \phi_2) (1 - I_{2t-1})$$

- Provides perfect discrimination between regimes

# Estimation of TAR-TAR

## Continued

- Initialisation:  $(\gamma_1^0, \gamma_2^0, \phi_1^0, \phi_2^0, c_2^0)$
- Built-in grid search to estimate the threshold level (not required when no switching involved)
- **Comment:** In practice actually easier to search over all possible values of  $z_{it}$  which are of  $O(T)$
- $c_i$  chosen such that the standard error of regression is minimised
- Consistency discussed in Franses & van Dijk (2000)
- Re-iterates till converged

# Estimation Results

Table: AR-TAR (Switching Return)

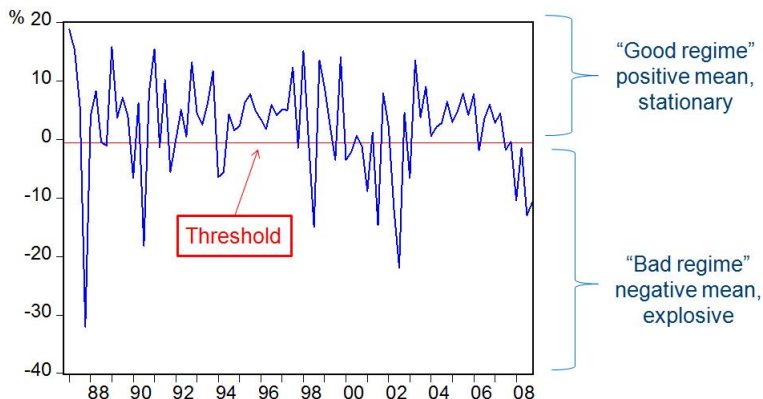
Model	$\alpha$	$\gamma_1$	$\phi_1$	$\gamma_2$	$\phi_2$	$c$	$\pi$	(Min,Max)	SSE
<b>TAR</b>									
LIBOR	0.51** (0.07)	-1.25* (0.48)	1.27** (0.10)	2.38** (0.65)	0.18 (0.11)	6.25	0.56	(2.83,15.25)	233.45
INF	0.77** (0.09)	0.25 (1.67)	-0.09 (0.15)	-0.22 (1.08)	0.95** (0.23)	0.94	0.33	(-0.57,2.73)	259.44
FT	0.53** (0.09)	2.22** (0.32)	0.31** (0.08)	-4.05** (0.94)	1.77** (0.24)	-1.54	0.76	(-32.0,18.84)	179.91
GDP	0.81** (0.12)	1.72 (0.99)	0.11 (0.08)	5.18 (6.66)	4.27* (1.86)	-0.41	0.94	(-1.80,2.20)	208.52
<b>AR</b>	0.94** (0.04)	-1.35 (2.82)	0.12 (0.15)						309.53

Notes: (i) Newey-West HAC s.d. in parenthesis; (ii) \* sig. at 5%, \*\* at 1%.

# TAR on FT Returns

## Implication on Real Estate Return

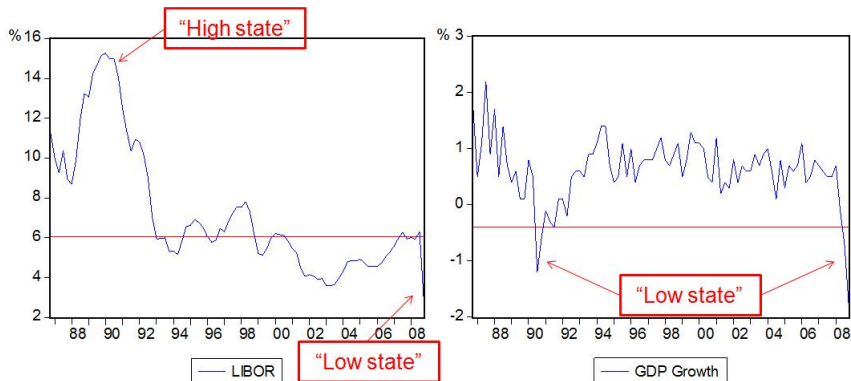
Figure: Quarterly log-returns on FT Index (Q4 1986 - Q4 2008)



# Quality of Regime Indicators

The 1990s & the recent crises

Figure: End-of-quarter LIBOR and Quarterly GDP Growth

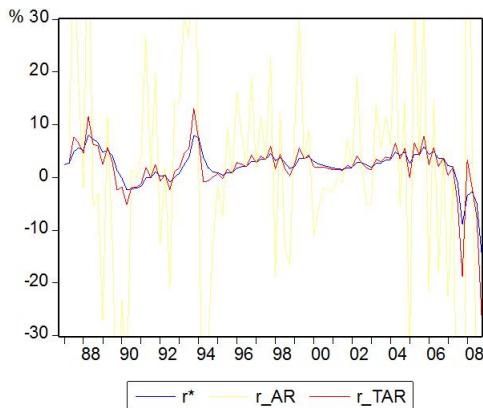


# AR-TAR Results

- Arguably, much more economically sound than AR-AR
- AR-AR: Smoothing = 0.94, i.e. the unsmoothed return will be highly volatile (at all times)
- AR-TAR: “Abnormal return” sieved from “normal return” where volatility is relatively low
- Usually, explosive return in one regime (bad regime), yet steady-state variance exists
- Knight & Satchell (2011):  $\phi_1^2 > 1$  and  $\phi_1^2\pi + (1 - \pi)\phi_2^2 < 1$

# Smoothed and Unsmoothed Returns

## AR-AR vs AR-TAR



	$r^*$	$r_{AR}$	$r_{TAR}$
Mean	2.14	-1.15	1.90
Median	2.56	1.86	2.26
Maximum	8.03	89.19	13.07
Minimum	-14.51	-177.01	-26.48
Std. Dev.	3.27	34.11	4.99
Skewness	-1.96	-2.12	-2.82
Kurtosis	10.54	13.16	16.61

Remark: Regime indicator = FT return

# Switching Smoothing

Table: Estimated TAR-AR model

Model	$\alpha_1$	$\alpha_2$	$\gamma$	$\phi$	$c$	$\pi$	(Min,Max)	SSE
<b>TAR</b>								
LIBOR	1.22** (0.10)	0.75** (0.20)	3.69 (2.71)	-0.04 (0.20)	12.30	0.10	(2.83,15.25)	272.24
INF	0.79** (0.14)	1.33** (0.11)	2.21 (1.97)	-0.09 (0.18)	0.87	0.43	(-0.57,2.73)	269.40
FT	0.21** (0.08)	1.97** (0.25)	1.17 (0.67)	0.60** (0.16)	-1.54	0.76	(-32.00,18.84)	195.51
GDP	0.82** (0.12)	1.86** (0.45)	1.78 (2.90)	0.08 (0.12)	-0.59	0.95	(-1.80,2.20)	246.65
<b>AR</b>								
	0.94** (0.04)		-1.35 (2.82)	0.12 (0.15)				309.53

Notes: (i) Newey-West HAC s.d. in parenthesis; (ii) \* sig. at 5%, \*\* at 1%.

# Switching Smoothing

## Continued

- Basically,

$$\pi\hat{\alpha}_1 + (1 - \pi)\hat{\alpha}_2 = \hat{\alpha}$$

- Excessive smoothing in “bad regime” i.e. high LIBOR / low FT returns / low GDP growth
- Psychological effect (?)
- But fits data slightly worse than AR-TAR (switching return)
- In some cases, lower (unconditional) volatility than smoothed return e.g. for FT return, s.d. according to TAR-AR = 2.83 less than 3.27 = s.d. of smoothed return

# Co-switching Model

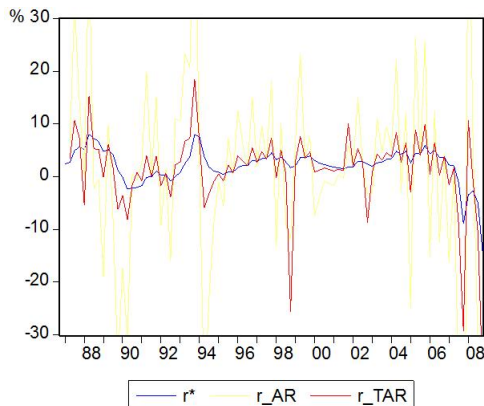
Table: Estimated TAR-TAR model

Model	$\alpha_1$	$\alpha_2$	$\gamma_1$	$\phi_1$	$\gamma_2$	$\phi_2$	$c_1$	$c_2$	SSE
<b>TAR-TAR</b>									
LIBOR-LIBOR	1.42** (0.28)	0.73** (0.07)	-0.37 (1.33)	0.79** (0.14)	3.05** (0.68)	-0.04 (0.15)	6.21	11.31	250.27
FT-FT	0.72** (0.07)	0.96** (0.04)	3.36** (0.70)	0.01 (0.09)	-7.13* (2.61)	1.40** (0.41)	-13.33	-1.20	183.16
LIBOR-FT	1.40** (0.24)	0.56** (0.09)	1.63** (0.61)	0.35** (0.12)	5.28** (1.61)	-0.85* (0.42)	6.21	-1.79	335.93
FT-LIBOR	0.79** (0.10)	2.22** (0.80)	0.24 (1.49)	0.59* (0.29)	3.34** (1.03)	-0.18 (0.13)	-13.34	5.95	211.80
<b>AR-AR</b>									
	0.94** (0.04)		-1.35 (2.82)	0.12 (0.15)					309.53

Notes: (i) Newey-West HAC s.d. in parenthesis; (ii) \* sig. at 5%, \*\* at 1%.

# Unsmoothed Returns under Co-switching

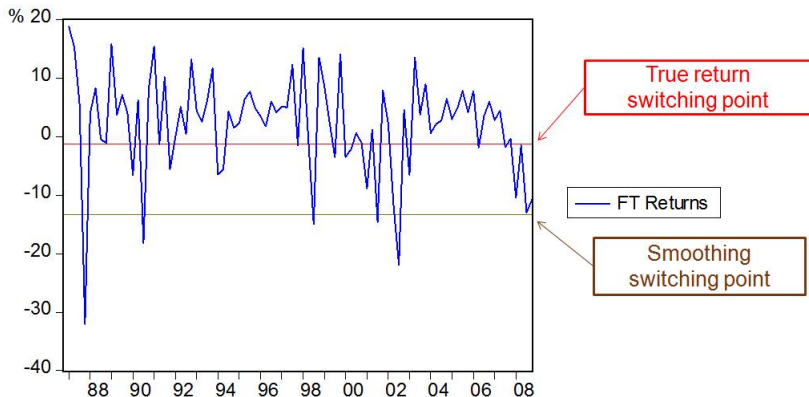
## AR-AR vs TAR-TAR



	$r^*$	$r_{AR}$	$r_{TAR}$
Mean	2.14	-1.15	1.25
Median	2.56	1.86	2.22
Maximum	8.03	89.19	18.52
Minimum	-14.51	-177.01	-39.21
Std. Dev.	3.27	34.11	7.95
Skewness	-1.96	-2.12	-2.47
Kurtosis	10.54	13.16	12.66

Remark: FT return = indicator for both smoothing and return equations

# How does Co-switching work?



# Comments on TAR-TAR

- Least restricted in this paper
- Explosive regime-switching behaviour either in the smoothing equation or in the return process, BUT not simultaneously
- From a theoretical viewpoint: Particular structure on the state probability when a single indicator governs both equations
- Supported by Hansen (1996, 1997) test results, especially when involved with FT returns
- Results driven by different combinations of the exogenous variables used, thereby opening up a new area of research

# Implication on Asset Allocation

- The TAR-TAR unconditional variance still close to that of the conventional smoothing model
- However, **Time-varying behaviour** (conditional smoothing) masked by the latter
- More informative and also crucial to successful dynamic (active) asset allocation
- Also, sheds more light on the nature of real estate risk
- The impact in the extreme regimes, although probably short lived, profound to asset values
- After all, quality of risk measures dependent of accuracy of estimated smoothing coefficient(s)

# Conclusions

- New unsmoothing technique for returns on an appraisal-based valuation index
- Clear evidence of regime effects and time-varying behavior in the commercial real estate returns
- Most promising results from the use of FT equity returns, LIBOR, and to a lesser extent, GDP growth
- TAR-TAR better than AR-AR according to SSE criteria (about 40% reduction)
- Applicable to other “illiquid” asset classes, e.g. hedge fund, venture capital, or even fine art!

Thank you!