

Exiting from QE

by

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Some comments from Mark Watson

Questions:

- How can we estimate effects of monetary policy when the policy instrument changes endogenously from the short term interest rate to QE?
- How can we estimate the effect of changing instruments, i.e., exiting from QE?

Answer:

- Use standard recursive SVAR that allows for discrete breaks associated with changes in policy instruments.
 - Model breaks in terms of observables
 - Do some careful data work
 - Estimate parameters using appropriate methods (breaks, truncation associated with bounds).
 - Make some sensible empirical choices
 - Compute IRFs and counterfactuals using nonlinear methods.

Models and VARs

Notation: $\begin{bmatrix} X_t \\ P_t \end{bmatrix}$, $X_t = \begin{bmatrix} \text{Output gap} \\ \text{Inflation} \end{bmatrix}$ and $P_t = \begin{bmatrix} \text{Bank Rate} \\ \text{Excess Reserves} \end{bmatrix}$

Model:
$$\begin{cases} X_t = f_X(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^X), \eta_{X,t}) \\ P_t = f_P(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^P), \eta_{P,t}) \end{cases}$$

Linearize model and solve

SVAR:
$$\begin{cases} X_t = \Phi X_{t-1} + \mathbf{B}P_{t-1} + \Gamma_{XX}\eta_{X,t} + \mathbf{\Gamma}_{XP}\eta_{P,t} \\ P_t = \Lambda X_{t-1} + \Psi P_{t-1} + \Gamma_{PX}\eta_{X,t} + G_{PP}\eta_{P,t} \end{cases}$$

2 Models

$$\text{Model 0: } \begin{cases} X_t = f_X(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^X), \eta_{X,t}) \\ P_t = f_{0,P}(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^P), \eta_{P,t}) \end{cases}$$

$$\text{Model 1: } \begin{cases} X_t = f_X(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^X), \eta_{X,t}) \\ P_t = f_{1,P}(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^P), \eta_{P,t}) \end{cases}$$

$$\text{SVAR 0: } \begin{cases} X_t = \Phi_0 X_{t-1} + B_0 P_{t-1} + \Gamma_{0,XX} \eta_{X,t} + \Gamma_{0,XP} \eta_{P,t} \\ P_t = \Lambda_0 X_{t-1} + \Psi_0 P_{t-1} + \Gamma_{0,PX} \eta_{X,t} + G_{0,PP} \eta_{P,t} \end{cases}$$

$$\text{SVAR 1: } \begin{cases} X_t = \Phi_1 X_{t-1} + B_1 P_{t-1} + \Gamma_{1,XX} \eta_{X,t} + \Gamma_{1,XP} \eta_{P,t} \\ P_t = \Lambda_1 X_{t-1} + \Psi_1 P_{t-1} + \Gamma_{1,PX} \eta_{X,t} + G_{1,PP} \eta_{P,t} \end{cases}$$

1 Model, 2 Regimes ($s_t = 0$ or $s_t = 1$)

$$\text{Model}(s_t = 0): \begin{cases} X_t = f_X(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^X), \eta_{X,t}) \\ P_t = f_{0,P}(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^P), \eta_{P,t}) \end{cases}$$

$$\text{Model}(s_t = 1): \begin{cases} X_t = f_X(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^X), \eta_{X,t}) \\ P_t = f_{1,P}(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t^P), \eta_{P,t}) \end{cases}$$

$$\text{SVAR}(s_t = 0): ?? \begin{cases} X_t = \Phi_0 X_{t-1} + B_0 P_{t-1} + \Gamma_{0,XX} \eta_{X,t} + \Gamma_{0,XP} \eta_{P,t} \\ P_t = \Lambda_0 X_{t-1} + \Psi_0 P_{t-1} + \Gamma_{0,PX} \eta_{X,t} + G_{0,PP} \eta_{P,t} \end{cases}$$

$$\text{SVAR}(s_t = 1): ?? \begin{cases} X_t = \Phi_1 X_{t-1} + B_1 P_{t-1} + \Gamma_{1,XX} \eta_{X,t} + \Gamma_{1,XP} \eta_{P,t} \\ P_t = \Lambda_1 X_{t-1} + \Psi_1 P_{t-1} + \Gamma_{1,PX} \eta_{X,t} + G_{1,PP} \eta_{P,t} \end{cases}$$

What matters: $E(P_t, X_{t+1}, P_{t+1}, \dots | \Omega_t)$

which depends on forecasts of $s_t, s_{t+1}, s_{t+2}, s_{t+3}, \dots$

A special case: $P\left(s_{t+k} \mid s_t, X_t, P_t, \left\{s_{t-j}, X_{t-j}, P_{t-j}\right\}_{j \geq 1}\right) = P(s_{t+k} | s_t)$

That is, s_t is an exogenous first-order Markov process

Special Case: $P\left(s_{t+k} \mid s_t, X_t, P_t, \left\{s_{t-j}, X_{t-j}, P_{t-j}\right\}_{j \geq 1}\right) = P\left(s_{t+k} \mid s_t\right)$

Model:
$$\begin{cases} X_t = f_X\left(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots \mid \Omega_t^X), \eta_{X,t}\right) \\ P_t = f_{(s_t=0),P}\left(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots \mid \Omega_t^P), \eta_{P,t}\right) \\ P_t = f_{(s_t=1),P}\left(X_{t-1}, P_{t-1}, E(P_t, X_{t+1}, P_{t+1}, \dots \mid \Omega_t^P), \eta_{P,t}\right) \end{cases}$$

Suppose $\Omega_t^X = \Omega_t^P$ (and both include s_t). Things are simple

SVAR($s_t = 0$):
$$\begin{cases} X_t = \Phi_0 X_{t-1} + B_0 P_{t-1} + \Gamma_{0,XX} \eta_{X,t} + \mathbf{\Gamma}_{0,XP} \eta_{P,t} \\ P_t = \Lambda_0 X_{t-1} + \Psi_0 P_{t-1} + \Gamma_{0,PX} \eta_{X,t} + G_{0,PP} \eta_{P,t} \end{cases}$$

SVAR($s_t = 1$):
$$\begin{cases} X_t = \Phi_1 X_{t-1} + B_1 P_{t-1} + \Gamma_{1,XX} \eta_{X,t} + \mathbf{\Gamma}_{1,XP} \eta_{P,t} \\ P_t = \Lambda_1 X_{t-1} + \Psi_1 P_{t-1} + \Gamma_{1,PX} \eta_{X,t} + G_{1,PP} \eta_{P,t} \end{cases}$$

Suppose Ω_t^X includes s_{t-1} but not s_t

Ω_t^P include s_t

$$\text{SVAR}(s_t = 0, s_{t-1} = 0): \begin{cases} X_t = \Phi_{0,0}X_{t-1} + B_{0,0}P_{t-1} + \Gamma_{0,0,XX}\eta_{X,t} + \Gamma_{0,0,XP}\eta_{P,t} \\ P_t = \Lambda_{0,0}X_{t-1} + \Psi_{0,0}P_{t-1} + \Gamma_{0,0,PX}\eta_{X,t} + G_{0,0,PP}\eta_{P,t} \end{cases}$$

$$\text{SVAR}(s_t = 1, s_{t-1} = 0): \begin{cases} X_t = \Phi_{1,0}X_{t-1} + B_{1,0}P_{t-1} + \Gamma_{1,0,XX}\eta_{X,t} + \Gamma_{1,0,XP}\eta_{P,t} \\ P_t = \Lambda_{1,0}X_{t-1} + \Psi_{1,0}P_{t-1} + \Gamma_{1,0,PX}\eta_{X,t} + G_{1,0,PP}\eta_{P,t} \end{cases}$$

$$\text{SVAR}(s_t = 0, s_{t-1} = 1): \begin{cases} X_t = \Phi_{0,1}X_{t-1} + B_{0,1}P_{t-1} + \Gamma_{0,1,XX}\eta_{X,t} + \Gamma_{0,1,XP}\eta_{P,t} \\ P_t = \Lambda_{0,1}X_{t-1} + \Psi_{0,1}P_{t-1} + \Gamma_{0,1,PX}\eta_{X,t} + G_{0,1,PP}\eta_{P,t} \end{cases}$$

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Truth: $\text{SVAR}(s_t = 0, s_{t-1} = 0)$, $\text{SVAR}(s_t = 0, s_{t-1} = 1)$,
 $\text{SVAR}(s_t = 1, s_{t-1} = 0)$, $\text{SVAR}(s_t = 1, s_{t-1} = 1)$

Estimate: $\text{SVAR}(s_t = 0)$, $\text{SVAR}(s_t = 1)$ model.

What will I find ?

$\text{SVAR}(s_t = 0)$ will be a mixture of $\text{SVAR}(s_t = 0, s_{t-1} = 0)$, $\text{SVAR}(s_t = 0, s_{t-1} = 1)$

$\text{SVAR}(s_t = 1)$ will be a mixture of $\text{SVAR}(s_t = 1, s_{t-1} = 0)$, $\text{SVAR}(s_t = 1, s_{t-1} = 1)$

but

$P(s_t = 0, s_{t-1} = 0) \gg P(s_t = 0, s_{t-1} = 1)$, so $\text{SVAR}(s_t = 0) \approx \text{SVAR}(s_t = 0, s_{t-1} = 0)$

$P(s_t = 1, s_{t-1} = 1) \gg P(s_t = 1, s_{t-1} = 0)$, so $\text{SVAR}(s_t = 1) \approx \text{SVAR}(s_t = 1, s_{t-1} = 1)$

Some questions for this misspecified model

(1) IRFs of Policy variables under $s = 1$ (R – shocks) or $s = 0$ (m – shocks)

$$\text{SVAR}(s_t = 1) \approx \text{SVAR}(s_t = 1, s_{t-1} = 1)$$

$$\text{SVAR}(s_t = 0) \approx \text{SVAR}(s_t = 0, s_{t-1} = 0)$$

Approximately correct “continuing regime” answers.

(2) Counterfactual “Exit from Regime”

	s_{t-1}	s_t	s_{t+1}	s_{t+2}
Factual	0	1	1	1
Counterfactual	0	0	1	1

Correct Answers from

	$t-1$	t	$t+1$	$t+2$
Factual		$\text{SVAR}(s_t = 1, s_{t-1} = 0)$	$\text{SVAR}(s_t = 1, s_{t-1} = 1)$	
Counterfactual		$\text{SVAR}(s_t = 0, s_{t-1} = 0)$	$\text{SVAR}(s_t = 0, s_{t-1} = 1)$	

Misspecified Model Answers from

	$t-1$	t	$t+1$	$t+2$
Factual		$\text{SVAR}(s_t = 1, s_{t-1} = 1)$	$\text{SVAR}(s_t = 1, s_{t-1} = 1)$	
Counterfactual		$\text{SVAR}(s_t = 0, s_{t-1} = 0)$	$\text{SVAR}(s_t = 1, s_{t-1} = 1)$	

Special Case: $P\left(s_{t+k} \mid s_t, X_t, P_t, \left\{s_{t-j}, X_{t-j}, P_{t-j}\right\}_{j \geq 1}\right) = P\left(s_{t+k} \mid s_t\right)$

Hyashi-Koeda:

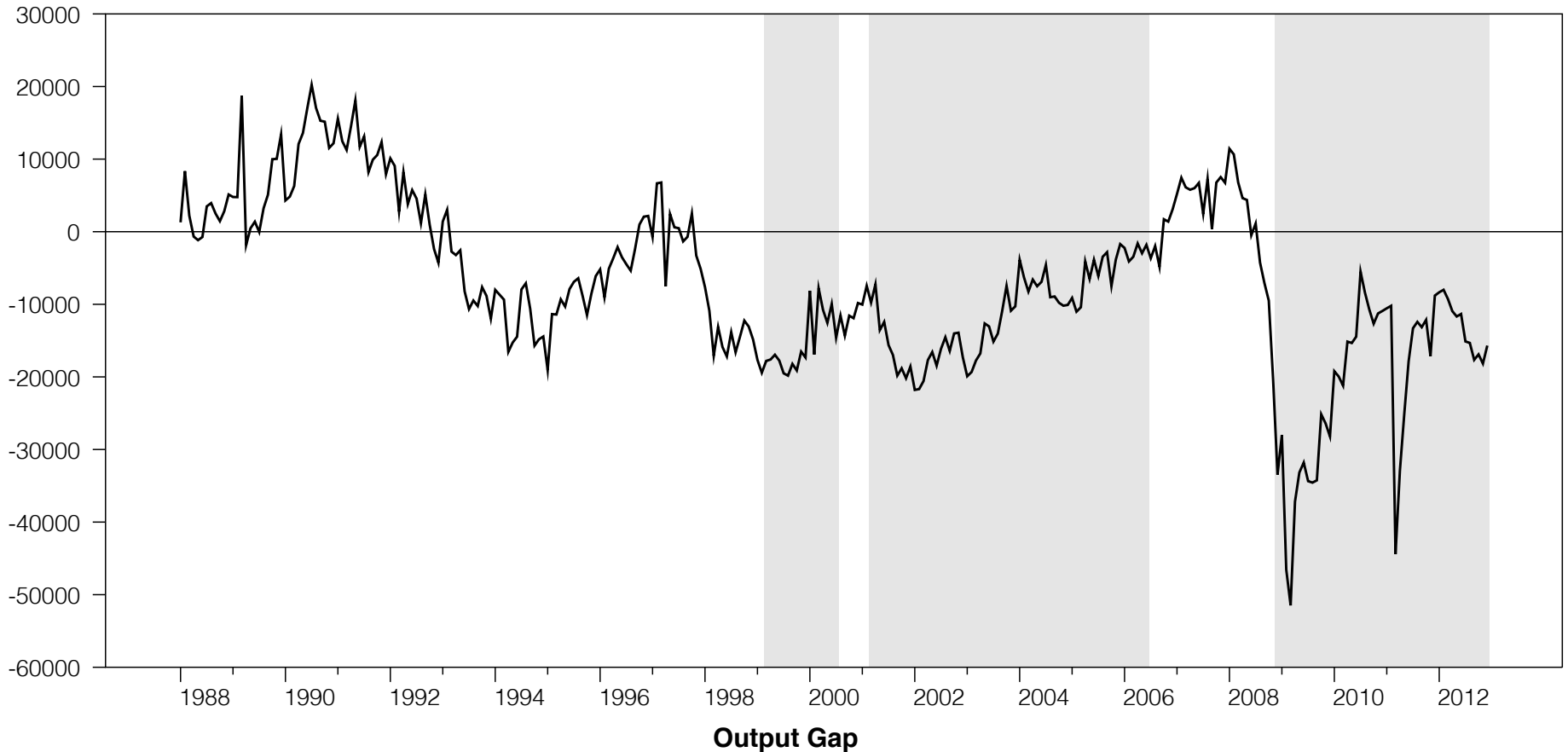
$$s_t = \begin{cases} 1 & \text{if } \begin{cases} s_{t-1} = 0, \rho r^*(X_t, \dots, X_{t-11}) + (1-\rho)r_{t-1} > \bar{r}_t > 0 \text{ and } \pi_t > q_t \sim N(\bar{\pi}, \sigma^2) \\ s_{t-1} = 1, \rho r^*(X_t, \dots, X_{t-11}) + (1-\rho)r_{t-1} > \bar{r}_t > 0 \end{cases} \\ 0 & \text{otherwise} \end{cases}$$

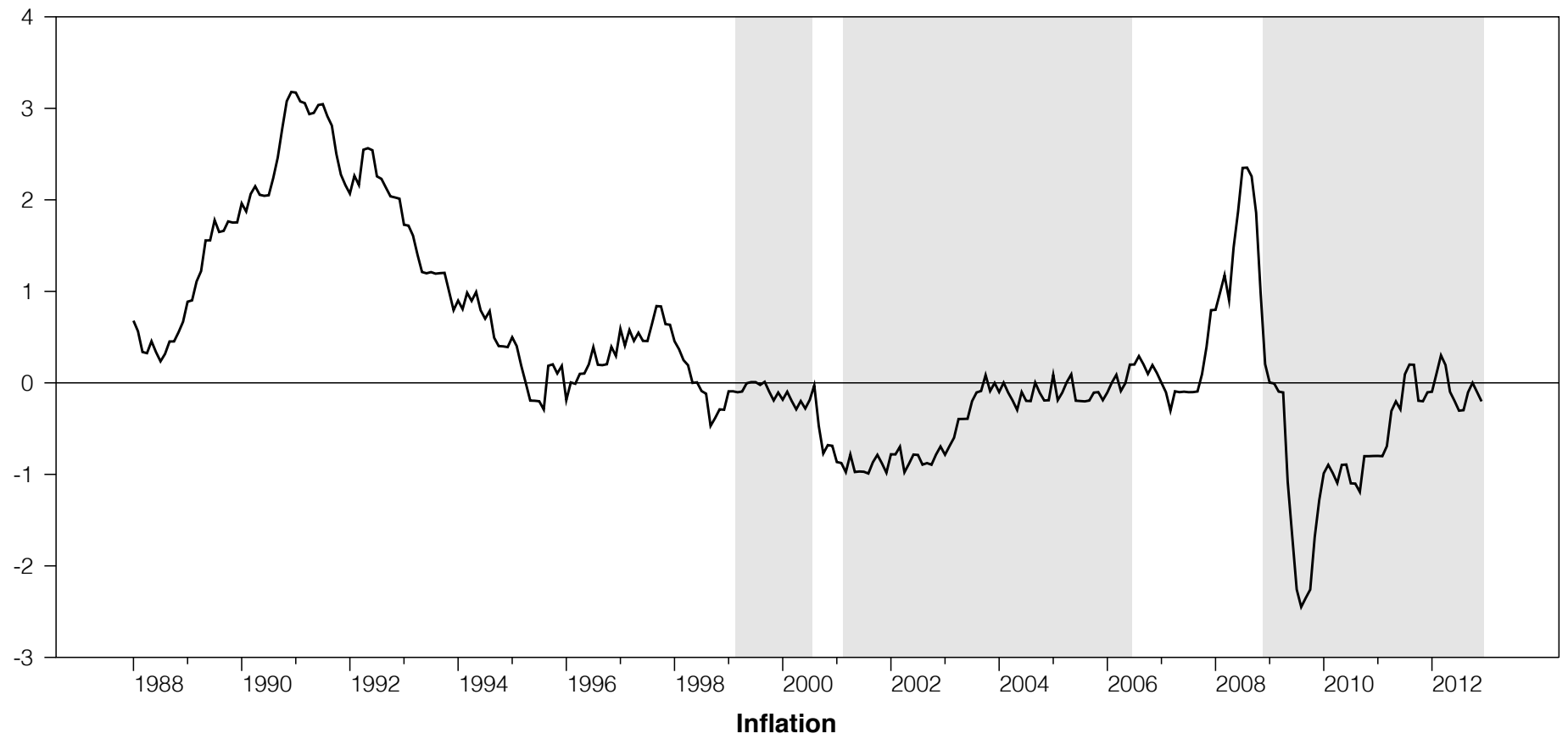
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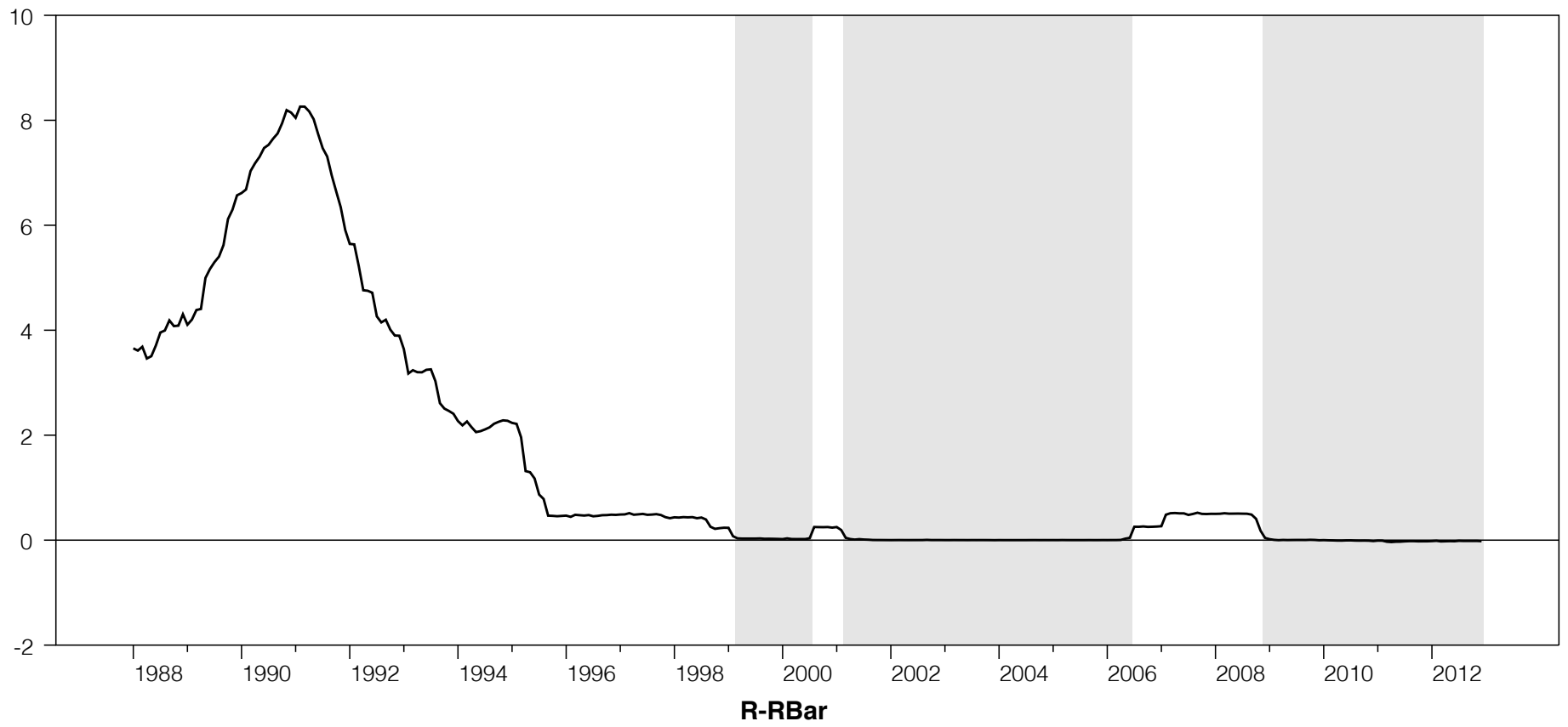
SVAR ... more complicated (!)

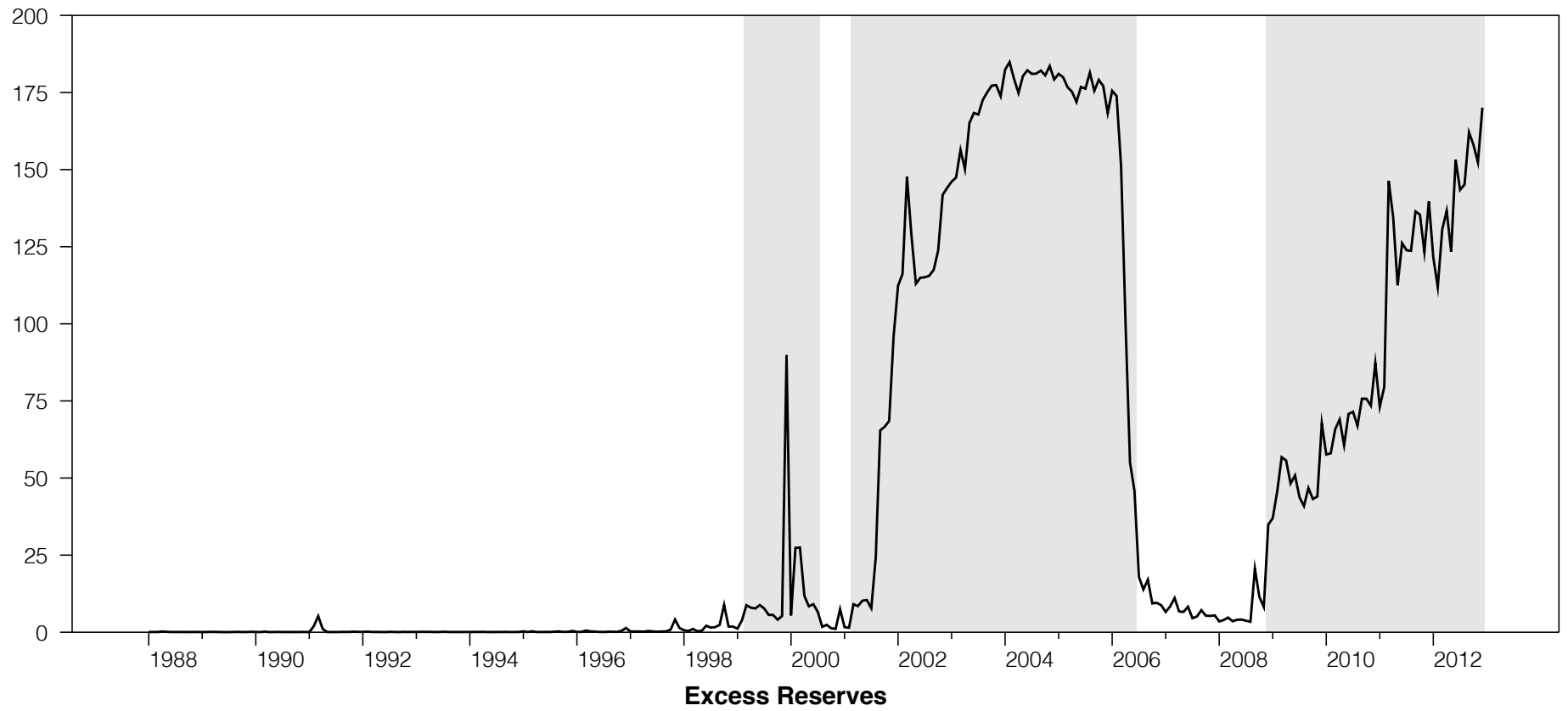
Approximations: SVAR($s_t = 0$) and SVAR($s_t = 1$) are averages over these regimes

Data:









Approximations: $\text{SVAR}(s_t = 0)$ and $\text{SVAR}(s_t = 1)$ are averages over these regimes

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