

MACROECONOMETRICS

Master in Economics

Markov Regime Switching Models

Markov Regime Switching Models

- Generalisation of the simple dummy variables approach to model structural changes in the parameters of a model
- Allow regimes (called states) to occur several periods over time
- In each period t the state is denoted by s_t .
- There can be m possible states: $s_t = 1, \dots, m$

Markov Regime Switching Models

Example: Dating Business Cycles

Let y_t denote the GDP growth rate

- Simple model with $m=2$ (only 2 regimes)

$$y_t = \mu_1 + e_t \quad \text{when } s_t = 1$$

$$y_t = \mu_2 + e_t \quad \text{when } s_t = 2$$

$$e_t \text{ i.i.d. } N(0, \sigma^2)$$

- Notation using dummy variables:

$$y_t = \mu_1 D_{1t} + \mu_2 (1 - D_{1t}) + e_t$$

$$\begin{aligned} \text{where } D_{1t} &= 1 \text{ when } s_t = 1, \\ &= 0 \text{ when } s_t = 2 \end{aligned}$$

Markov Regime Switching Models

- How does s_t evolve over time?

- Model: Markov switching :

$$P[s_t | s_1, s_2, \dots, s_{t-1}] = P[s_t | s_{t-1}]$$

- Probability of moving from state i to state j :

$$p_{ij} = P[s_t = j | s_{t-1} = i]$$

Markov Regime Switching Models

- Example with only 2 possible states: $m=2$
- Switching (or conditional) probabilities:

$$\text{Prob}[s_t = 1 \mid s_{t-1} = 1] = p_{11}$$

$$\text{Prob}[s_t = 2 \mid s_{t-1} = 1] = 1 - p_{11}$$

$$\text{Prob}[s_t = 2 \mid s_{t-1} = 2] = p_{22}$$

$$\text{Prob}[s_t = 1 \mid s_{t-1} = 2] = 1 - p_{22}$$

Markov Regime Switching Models

- Unconditional probabilities:

$$\text{Prob}[s_t = 1] = (1 - p_{22}) / (2 - p_{11} - p_{22})$$

$$\text{Prob}[s_t = 2] = 1 - \text{Prob}[s_t = 1]$$

Examples:

If $p_{11} = 0.9$ and $p_{22} = 0.7$ then $\text{Prob}[s_t = 1] = 0.75$

If $p_{11} = 0.9$ and $p_{22} = 0.9$ then $\text{Prob}[s_t = 1] = 0.5$

Markov Regime Switching Models

Estimation:

- Maximize Log-Likelihood
- Hamilton filter - Filtered probabilities:

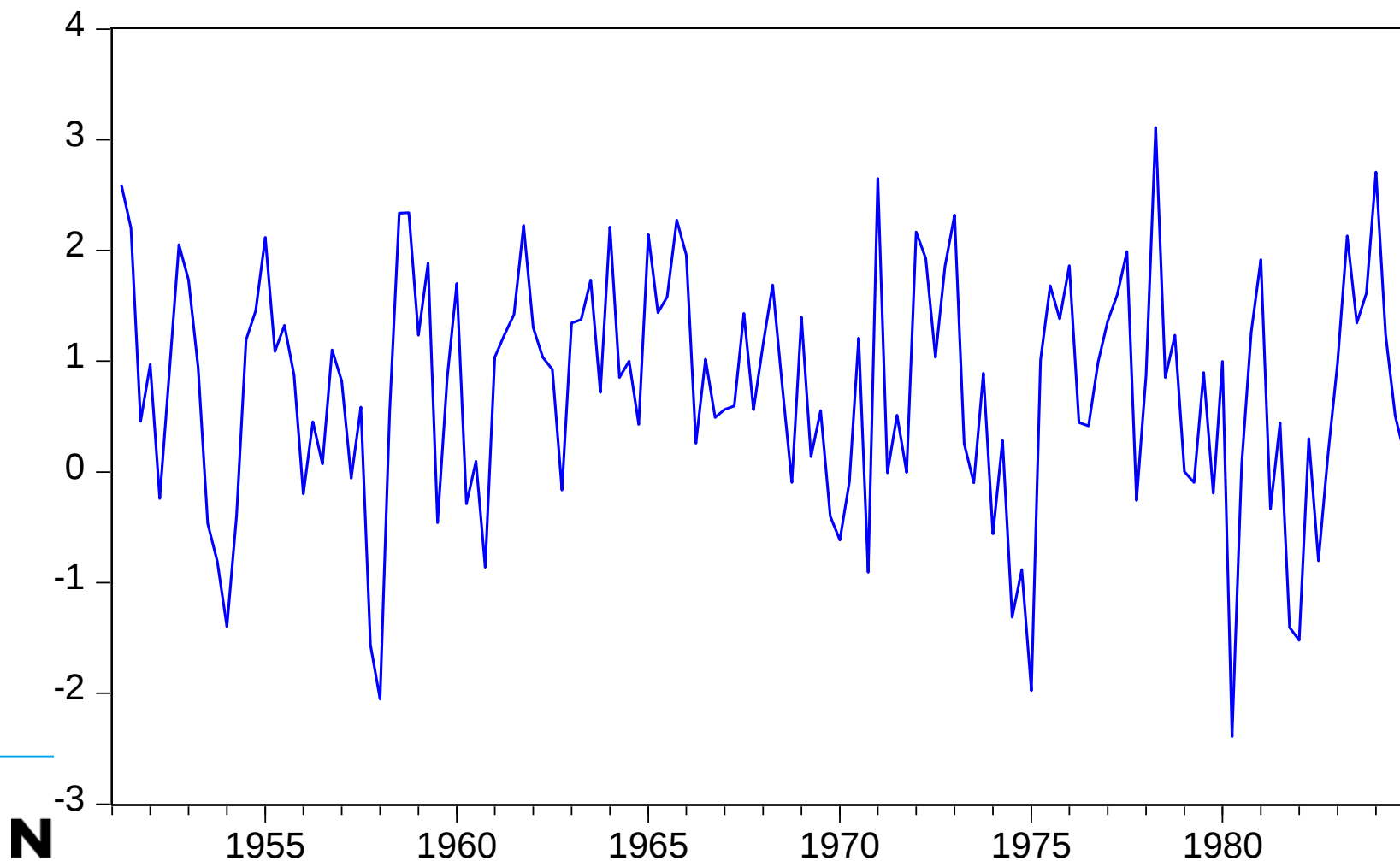
$$\pi_{1,t} = \text{Prob}[s_t = 1 \mid y_1, y_2, \dots, y_t]$$

$$\pi_{2,t} = \text{Prob}[s_t = 2 \mid y_1, y_2, \dots, y_t] = 1 - \pi_{1,t}$$

for $t=1, \dots, T$

Markov Regime Switching Models

Hamilton(1989)-GNP growth: gnp_hamilton.wf1



Markov Regime Switching Models

Variable	Coefficient	Std. Error	z-Statistic	Prob.
Regime 1				
C	1.104278	0.130904	8.435813	0.0000
Regime 2				
C	-0.486848	0.344455	-1.413388	0.1575
Common				
LOG(SIGMA)	-0.182101	0.075204	-2.421439	0.0155
Transition Matrix Parameters				
P11-C	2.314974	0.558760	4.143056	0.0000
P21-C	-0.785850	0.607785	-1.292973	0.1960

Markov Regime Switching Models

Estimated model with quarterly data,

$y_t = 100 * \text{dlog}(\text{gnp})$:

$$y_t = 1.1 + e_t \quad \text{when } s_t = 1$$

$$y_t = -0.5 + e_t \quad \text{when } s_t = 2$$

$$e_t \text{ i.i.d. } N(0, 0.8^2)$$

$$p_{11} = 0.91$$

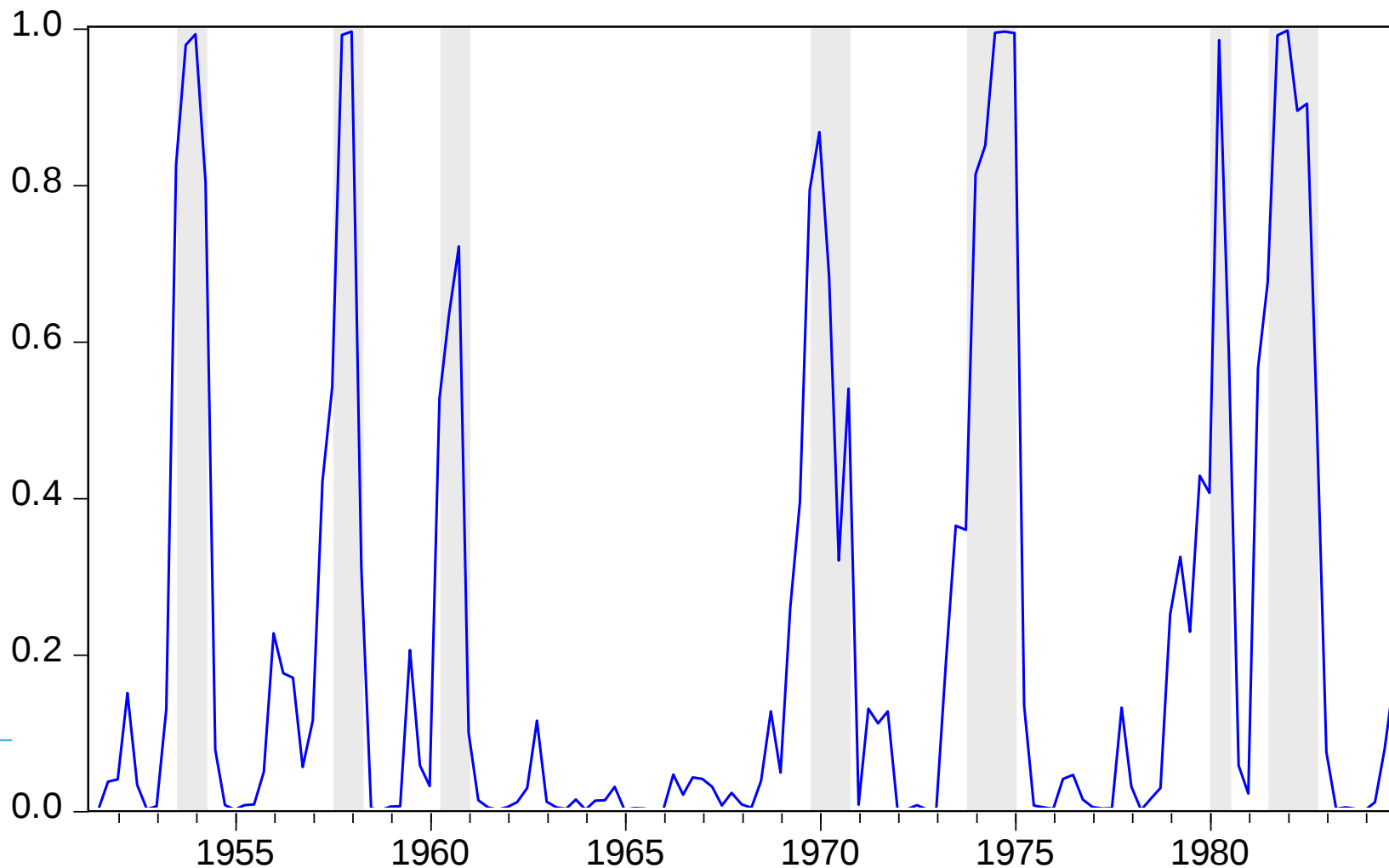
$$p_{22} = 0.69$$

Expected durations of regimes 1 and 2: 11 and 3 quarters respectively

Markov Regime Switching Models

Smoothed Regime Probabilities

$P(S(t) = 2)$



Markov Regime Switching Models

Time-Varying Transition Probabilities

- Example with only 2 possible states: $m=2$
- Switching (or conditional) probabilities:

$$\text{Prob}[s_t = 1 \mid s_{t-1} = 1] = p_{11} = F_{11} (a_{11} + b_{11} x_t)$$

$$\text{Prob}[s_t = 2 \mid s_{t-1} = 2] = p_{22} = F_{22} (a_{22} + b_{22} x_t)$$

- EViews uses multinomial logit for function $F(.)$

Markov Regime Switching Models

Time-Varying Transition Probabilities

- Example: Kim and Nelson (1999)

kimnelson_tvp.wf1

Y_t = log growth rate of industrial production (DLOGIP)

X_t = log growth rate of the Composite Index of Eleven Leading Indicators (DLOGIDX) → as a business-cycle predictor

Markov Regime Switching Models

Time-Varying Transition Probabilities

Variable	Coefficient	Std. Error	z-Statistic	Prob.
Regime 1				
C	0.517304	0.077907	6.640016	0.0000
Regime 2				
C	-0.865887	0.154375	-5.608983	0.0000
Common				
AR(1)	0.189474	0.050835	3.727263	0.0002
AR(2)	0.079344	0.051193	1.549889	0.1212
AR(3)	0.110945	0.052365	2.118690	0.0341
AR(4)	0.122252	0.050906	2.401551	0.0163
LOG(SIGMA)	-0.362469	0.038331	-9.456332	0.0000

Markov Regime Switching Models

Time-Varying Transition Probabilities

Time-varying transition probabilities:

$$P(i, k) = P(s(t) = k \mid s(t-1) = i)$$

(row = i / column = j)

Mean		1	2
	1	0.951975	0.048025
	2	0.200479	0.799521

Std. Dev.		1	2
	1	0.112804	0.112804
	2	0.145240	0.145240

Time-varying expected durations:

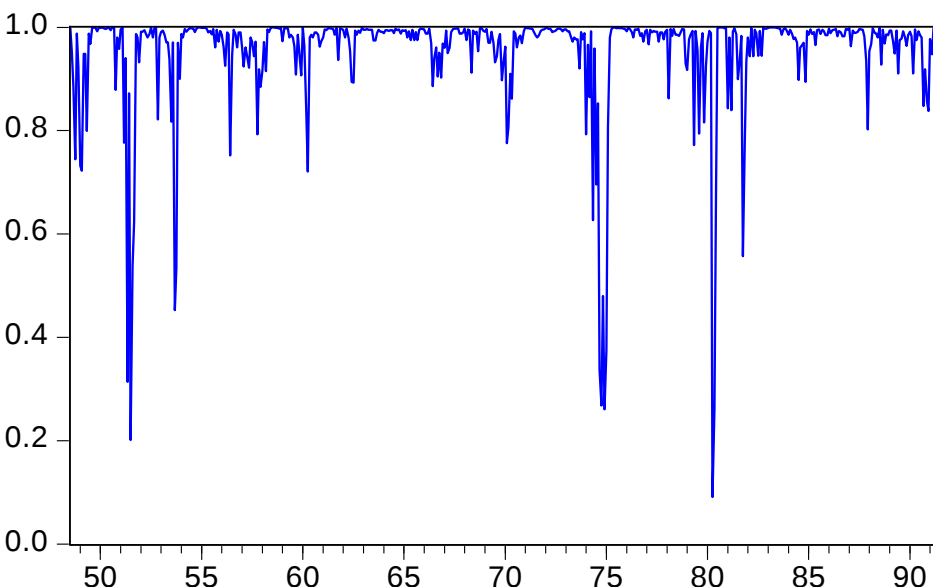
Mean		1	2
	1	506.3750	10.47908
Std. Dev.	2	2959.387	16.94910

Markov Regime Switching Models

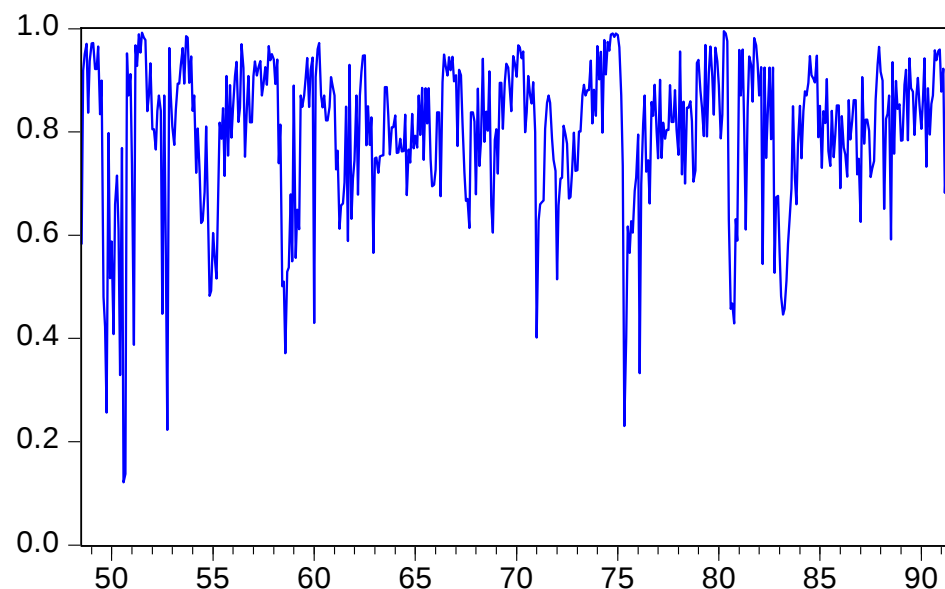
Time-Varying Transition Probabilities

Time-varying Markov Transition Probabilities

$\Pr(S(t)=1 | S(t-1)=1)$



$\Pr(S(t)=2 | S(t-1)=2)$

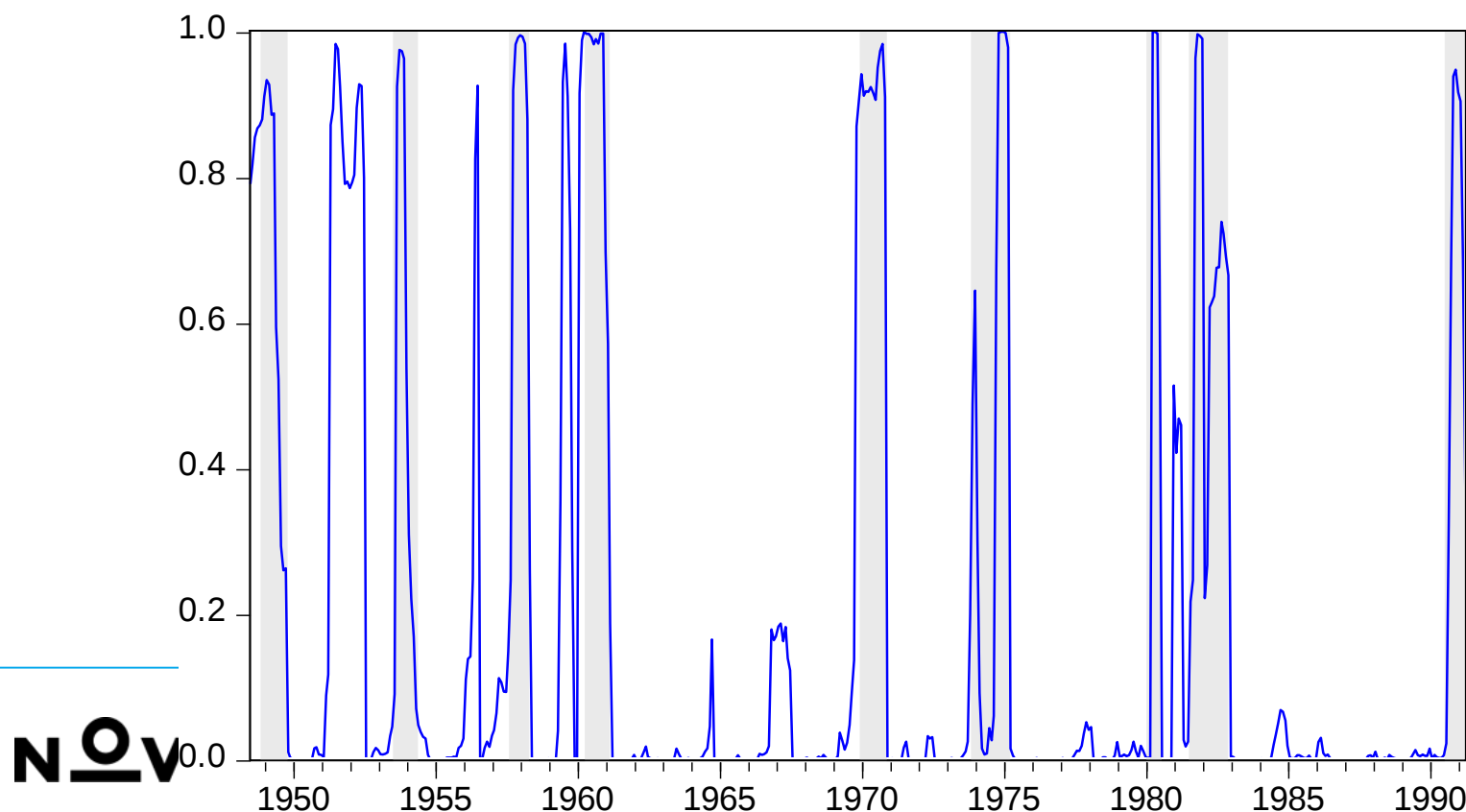


Markov Regime Switching Models

Time-Varying Transition Probabilities

Smoothed Regime Probabilities

$P(S(t) = 2)$



Markov Regime Switching Models

Regime Heteroskedasticity

Kim and Nelson (1999) - ew_excs.WF1

Y_t = monthly CRSP equal-weighted excess returns

Assumptions:

- 3 regimes
- Mean excess returns = 0
- Variance excess returns vary across the 3 regimes