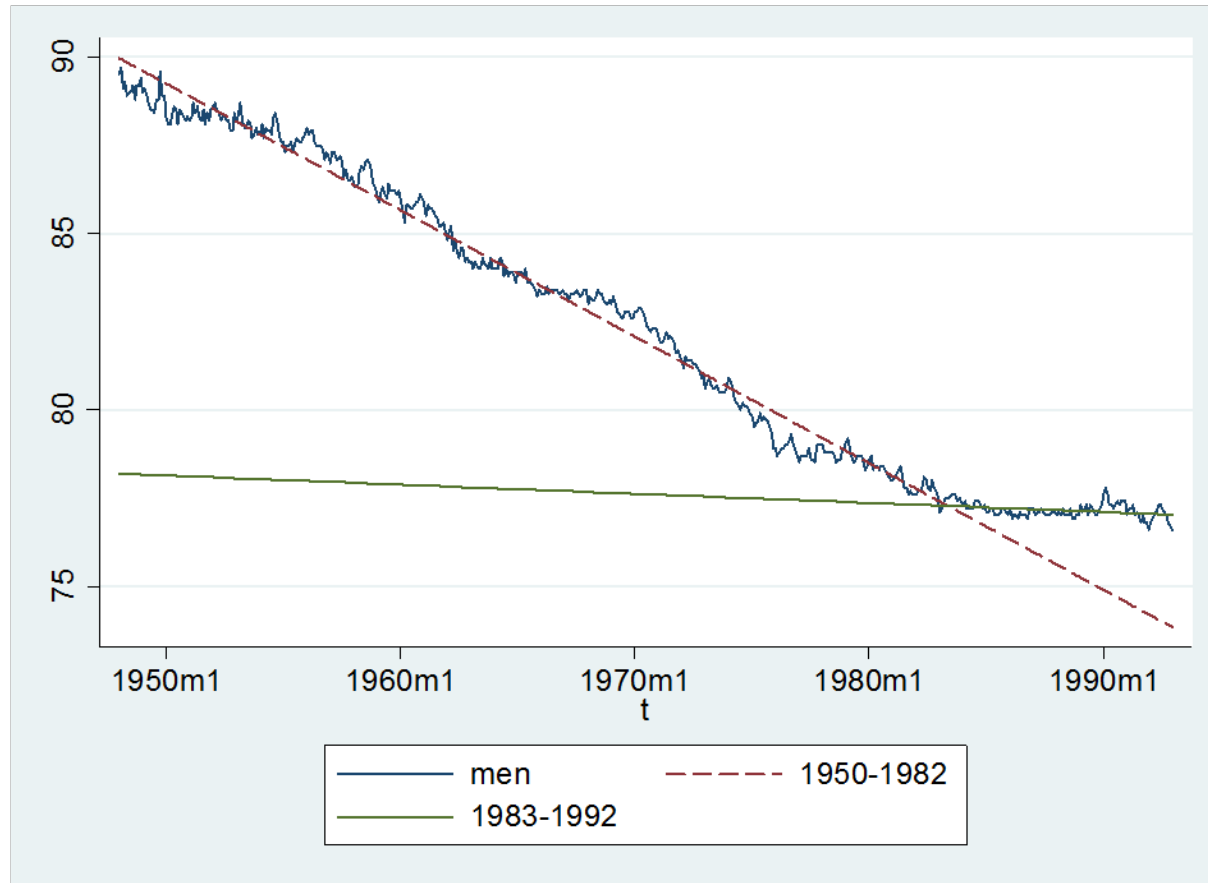


# Changing Trends

- We have seen in some cases that it appears that the trend slope has changed at some point.
- This is a type of structural change, sometimes called a **changing trend** or **breaking trend**.
- We can model this using the interaction of dummy variables with the trend.

# Labor Force Participation - Men



- Separate trends fit to 1950-1982 and 1983-1992

# Sub-Sample Trend Lines

- If you fit a trend for observations before and after a breakdate  $\tau$ , then for  $t \leq \tau$

$$T_t = \beta_0 + \beta_1 Time_t$$

and for  $t > \tau$

$$T_t = \alpha_0 + \alpha_1 Time_t$$

- Notice that both the intercept and slope change

# Estimation

- You can simply estimate on each sub-sample separately, and then forecast using the second set of estimates.
- Or, you can use dummy variable interactions.
- Define the dummy variable for observations after time  $\tau$

$$d_t = 1(t \geq \tau)$$

# Dummy Equation

$$\begin{aligned}T_t &= (\beta_0 + \beta_1 Time_t) \mathbb{1}(t < \tau) + (\alpha_0 + \alpha_1 Time_t) \mathbb{1}(t \geq \tau) \\ &= (\beta_0 + \beta_1 Time_t) + ((\alpha_0 - \beta_0) + (\alpha_1 - \beta_1) Time_t) \mathbb{1}(t \geq \tau) \\ &= \beta_0 + \beta_1 Time_t + \beta_2 d_t + \beta_3 Time_t d_t\end{aligned}$$

where

$$\beta_2 = \alpha_0 - \beta_0$$

$$\beta_3 = \alpha_1 - \beta_1$$

- This is a linear regression, with regressors  $Time_t$ ,  $d_t$  and  $Time_t d_t$

# Estimation

```
. generate d=(t>tm(1982m12))
. generate dt=d*t
. regress t d dt if (t<=tm(1992m12))
```

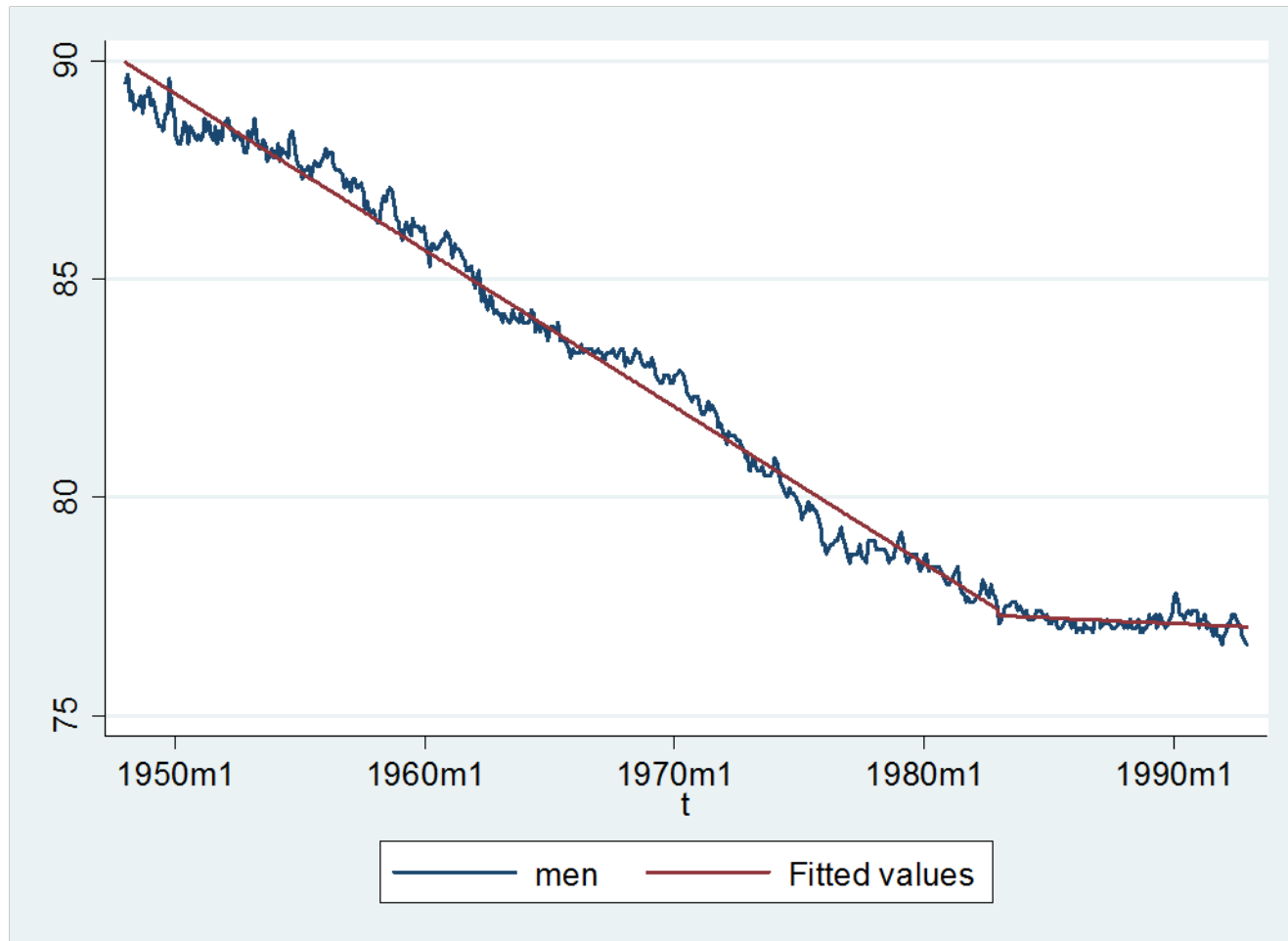
Source	SS	df	MS
Model	6947990	2	3473995
Residual	6173965	537	11497.1415
Total	13121955	539	24345

Number of obs = 540  
 F( 2, 537) = 302.16  
 Prob > F = 0.0000  
 R-squared = 0.5295  
 Adj R-squared = 0.5277  
 Root MSE = 107.22

t	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
d	-65.5	95.4503	-0.69	0.493	-253.0018	122.0018
dt	1	.2825718	3.54	0.000	.4449184	1.555082
_cons	65.5	5.232031	12.52	0.000	55.22224	75.77776

```
. predict yp
(option xb assumed; fitted values)
```

# Fitted



# Discontinuity

- One problem with this method is that the estimated trend function can be discontinuous
  - At the breakdate  $\tau$  there might be a jump in the trend function
  - This might not be sensible
  - We may wish to impose continuity

- In the model, this requires

$$\beta_0 + \beta_1\tau = \alpha_0 + \alpha_1\tau$$

or

$$\beta_2 + \beta_3\tau = 0$$



# Continuous Break

- You can impose a continuous trend by using a technique known as a **spline**

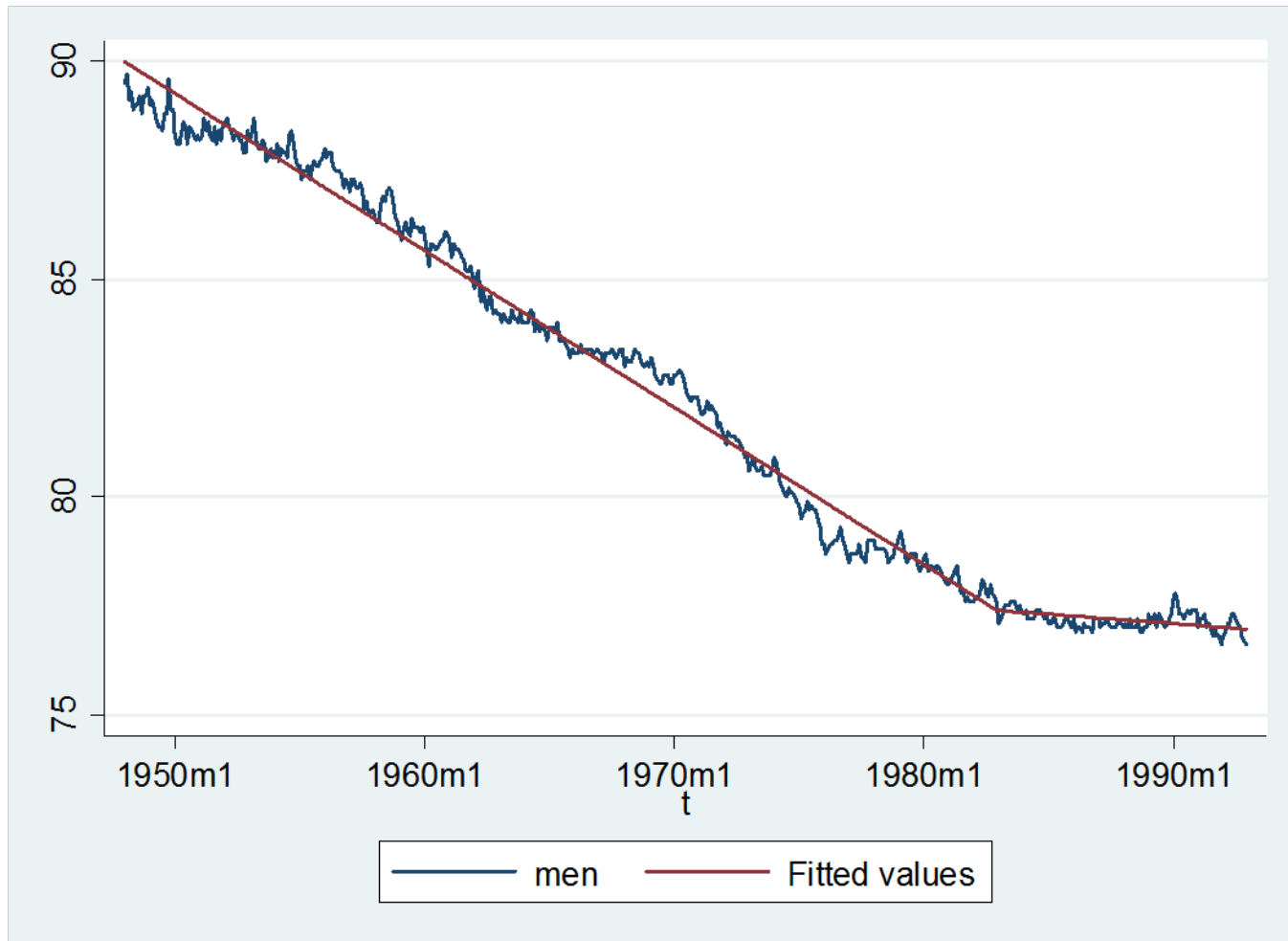
$$\begin{aligned} T_t &= \beta_0 + \beta_1 Time_t + \beta_2 (Time_t - \tau) 1(t \geq \tau) \\ &= \beta_0 + \beta_1 Time_t + \beta_2 Time_t^* \end{aligned}$$

where

$$Time_t^* = (Time_t - \tau) 1(t \geq \tau)$$

- The variable  $Time_t^*$  is 0 before the breakdate, and is a smoothly increasing trend afterwards.

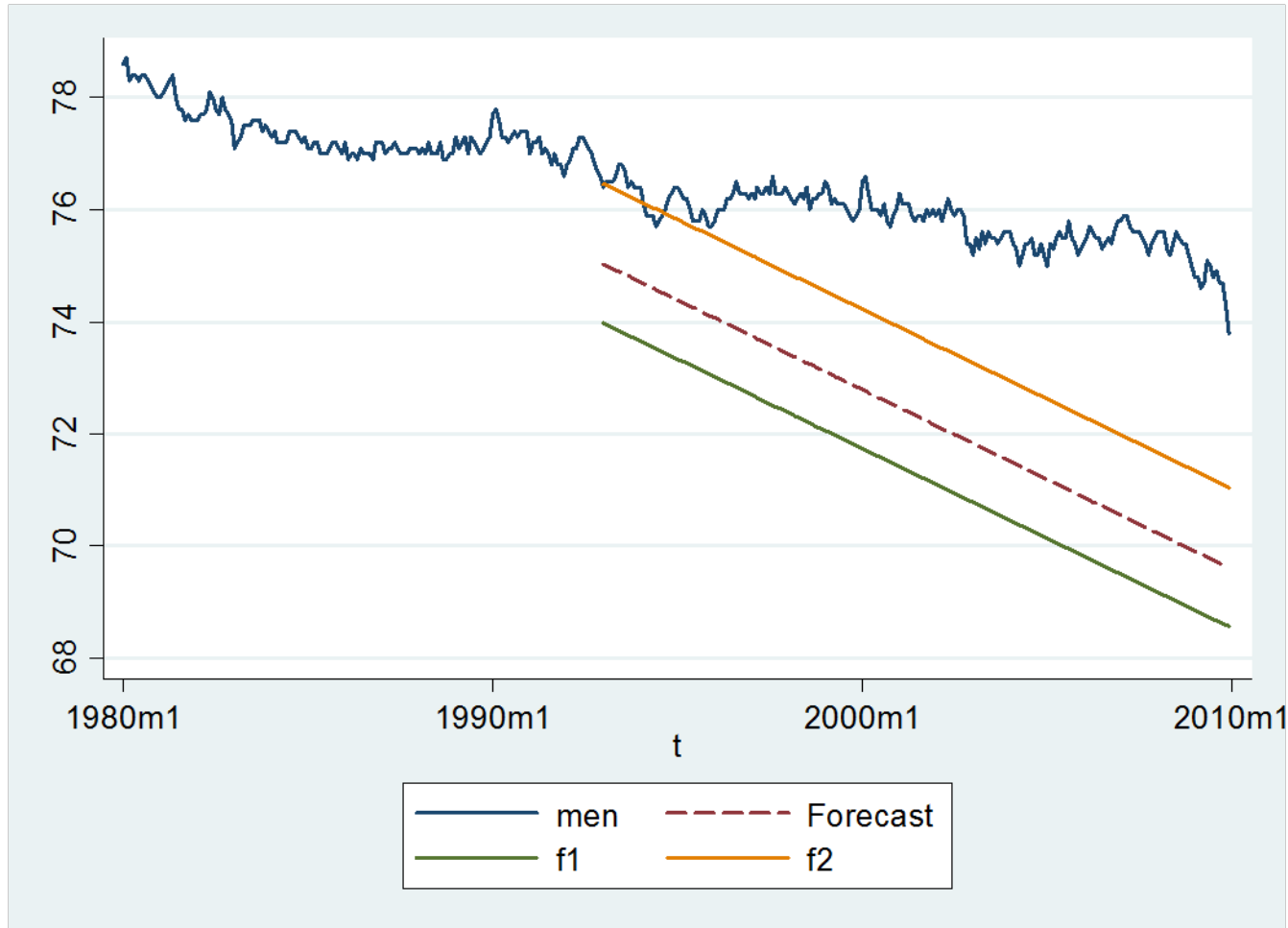
# Fitted Continuous Trend



# Continuous Trend Forecast



# Contrast with Linear Trend Forecast



# Real GDP

```
. generate tstar=(t-tq(1974q1))*(t>=tq(1974q1))
. regress y t tstar if t<=tq(1990q4)
```

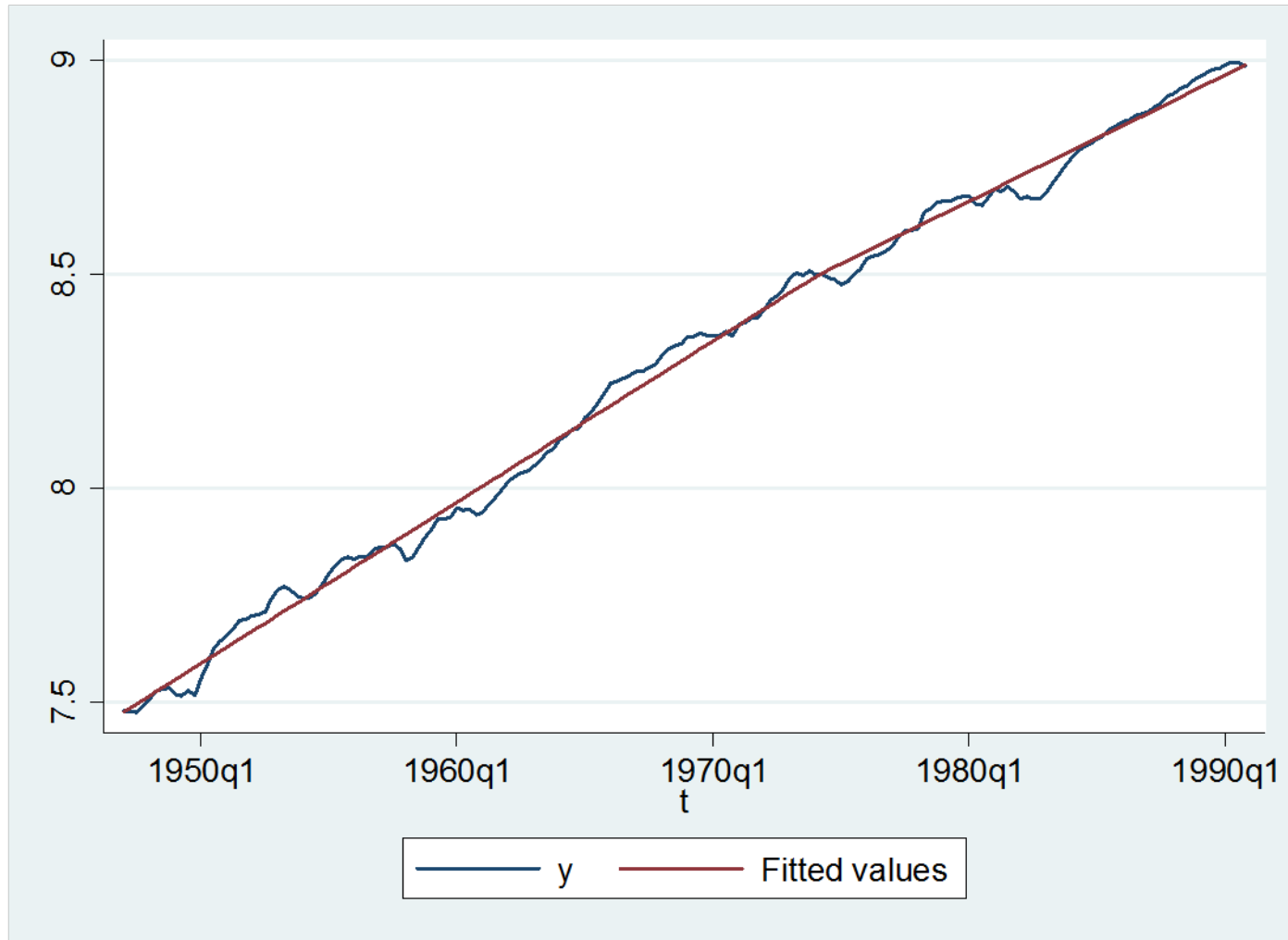
Source	SS	df	MS			
Model	34.7825634	2	17.3912817	Number of obs =	176	
Residual	.155349843	173	.000897976	F( 2, 173) =	19367.20	
Total	34.9379132	175	.199645218	Prob > F =	0.0000	
				R-squared =	0.9956	
				Adj R-squared =	0.9955	
				Root MSE =	.02997	

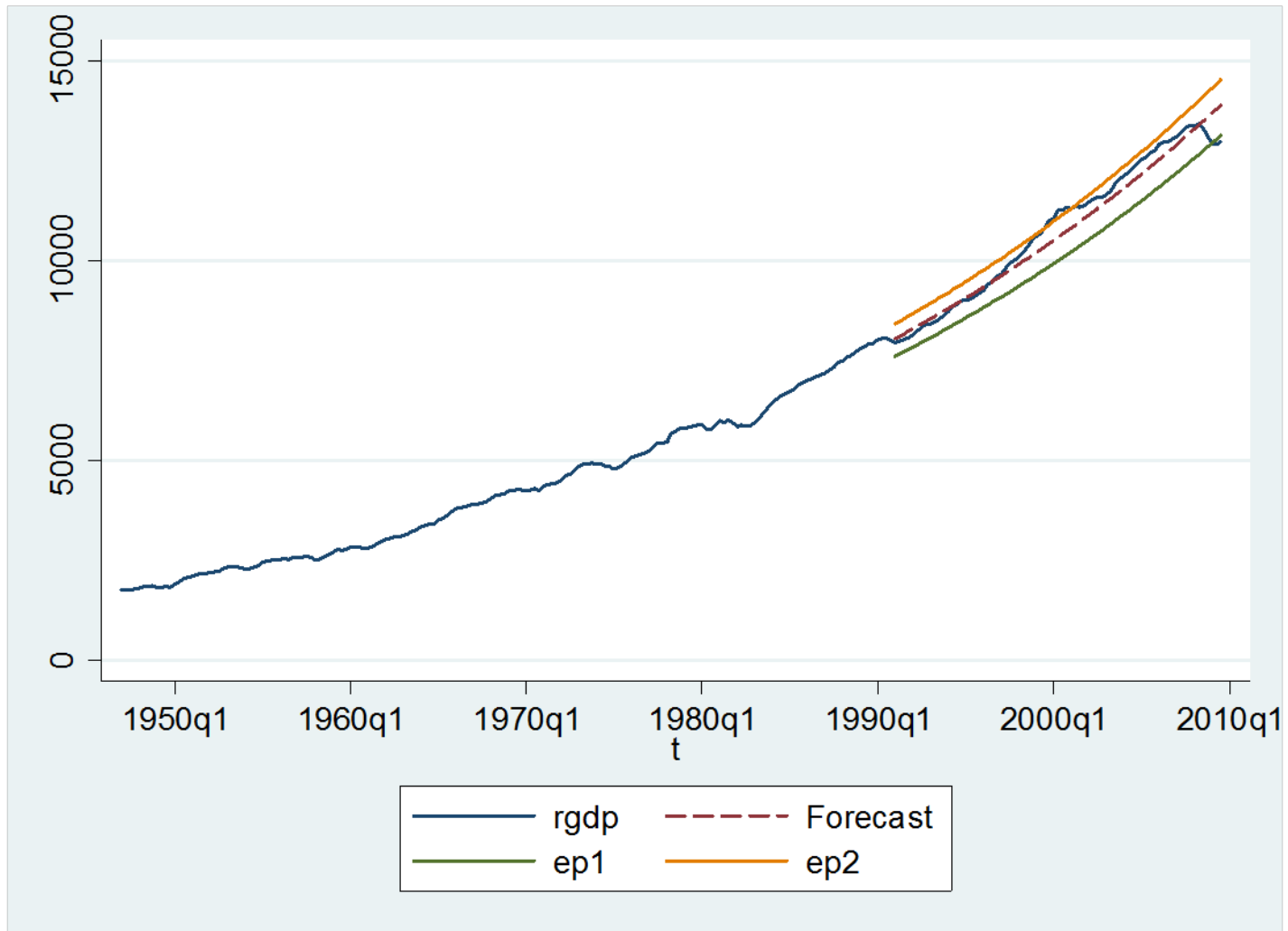
y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
t	.0094132	.0000775	121.43	0.000	.0092602	.0095662
tstar	-.0020582	.0001934	-10.64	0.000	-.0024398	-.0016766
_cons	7.96762	.0027667	2879.80	0.000	7.962159	7.973081

- Break in 1974q1

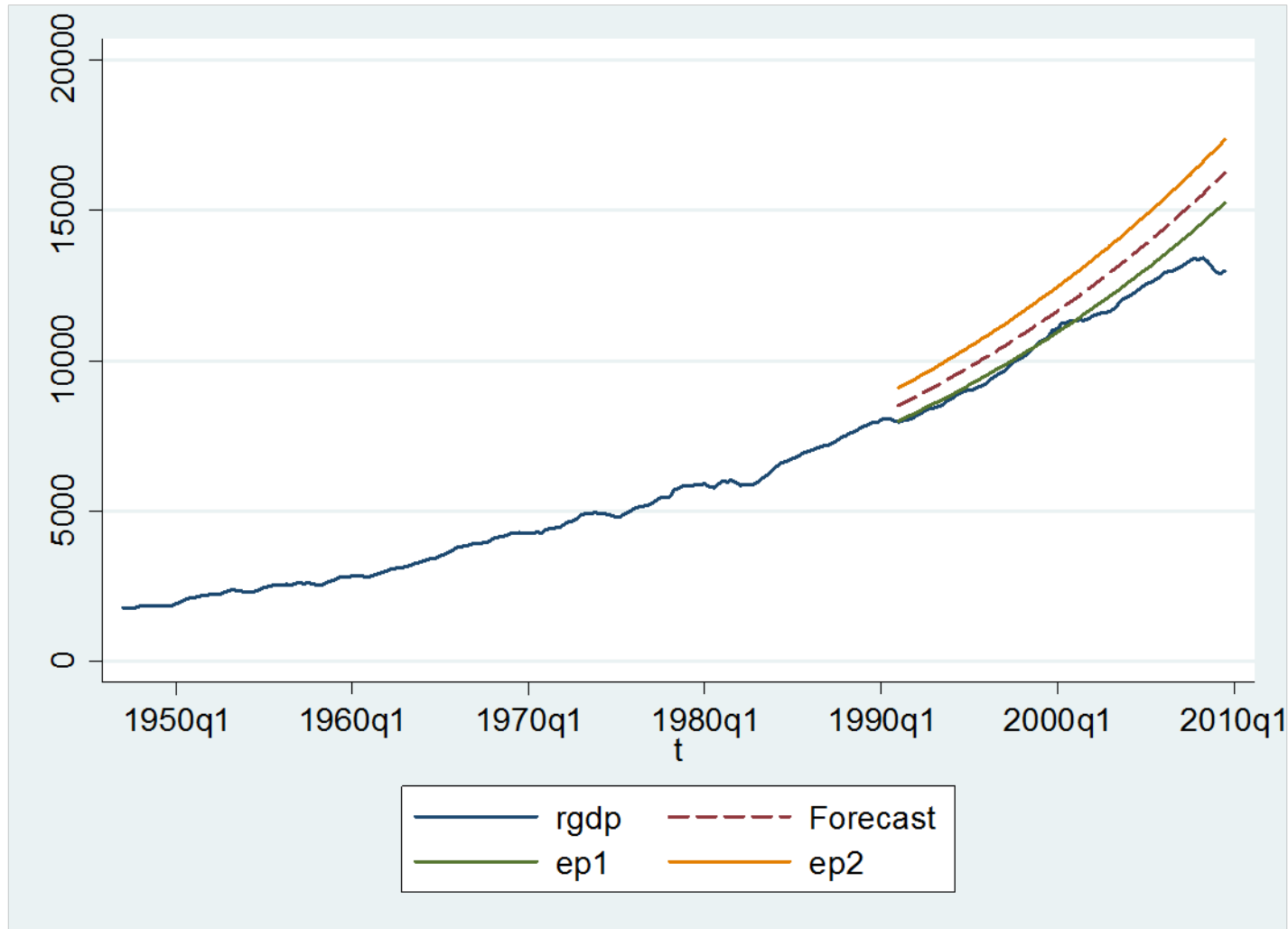
# Real GDP - fitted



# Forecast – Breaking Trend Model



# Contrast – Forecast from Linear Trend





# How to pick Breaks/Breakdates

- With caution, and skeptically
- Always have plenty of data (at least 10 years) after the breakdate
- Look for economic explanations
- Formally, the breakdate can be selected by minimizing the sum of squared errors