

KSE

Kyiv
School of
Economics

Special types of regressions

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Agenda

- Dummy Variables
- Trend extraction
- Trend Extraction in EViews
- Smoothing methods
- Smoothing methods in EViews

Dummy Variables

Special functions

- @trend
- @seas(i)

$$y_t = \beta_0 + \beta_1 q_1 + \beta_2 q_2 + \beta_3 q_3 + \varepsilon_t$$

$$y_t = \beta_0 + \beta_1 q_1 + \beta_2 q_2 + \beta_3 q_3 + \beta_4 t + \varepsilon_t$$

Equation Estimation

Specification Options

Equation specification

Dependent variable followed by list of regressors including ARMA and PDL terms, OR an explicit equation like $Y=c(1)+c(2)*X$.

y c @seas(1) @seas(2) @seas(3)

Estimation settings

Method: LS - Least Squares (NLS and ARMA)

Sample: 1947q1 2005q4

OK Скасувати

Equation Estimation

Specification Options

Equation specification

Dependent variable followed by list of regressors including ARMA and PDL terms, OR an explicit equation like $Y=c(1)+c(2)*X$.

y c @seas(1) @seas(2) @seas(3) @trend

Estimation settings

Method: LS - Least Squares (NLS and ARMA)

Sample: 1947q1 2005q4

OK Скасувати

Dummy variables

$$q_1 = (1, 0, 0, 0, 1, 0, 0, 0, 1, 0, \dots)'$$

$$q_2 = (0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, \dots)'$$

$$q_3 = (0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, \dots)'$$

1 quarter

$$y_t = \beta_0 + \beta_1 + \varepsilon_t$$

2 quarter

$$y_t = \beta_0 + \beta_2 + \varepsilon_t$$

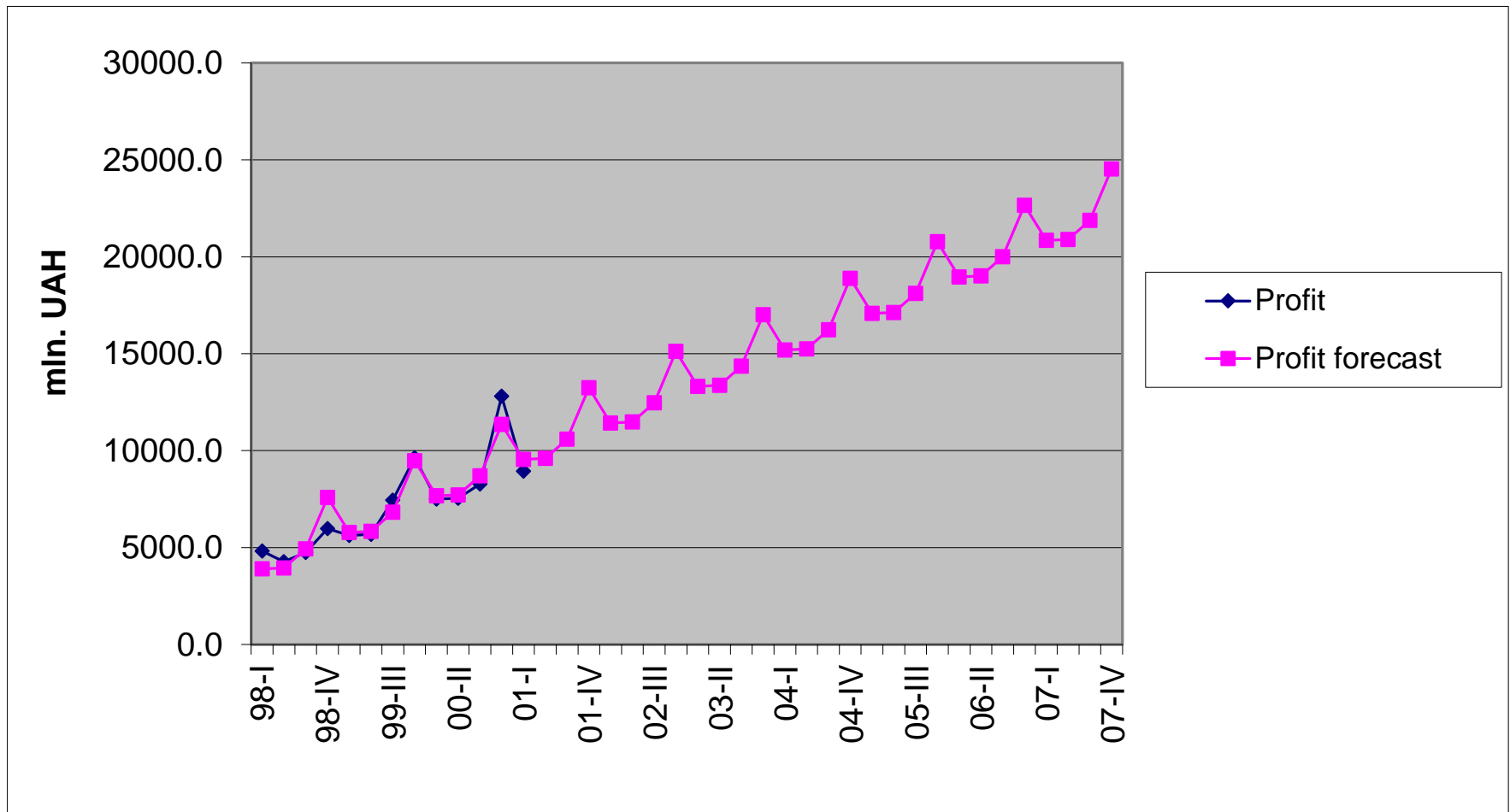
3 quarter

$$y_t = \beta_0 + \beta_3 + \varepsilon_t$$

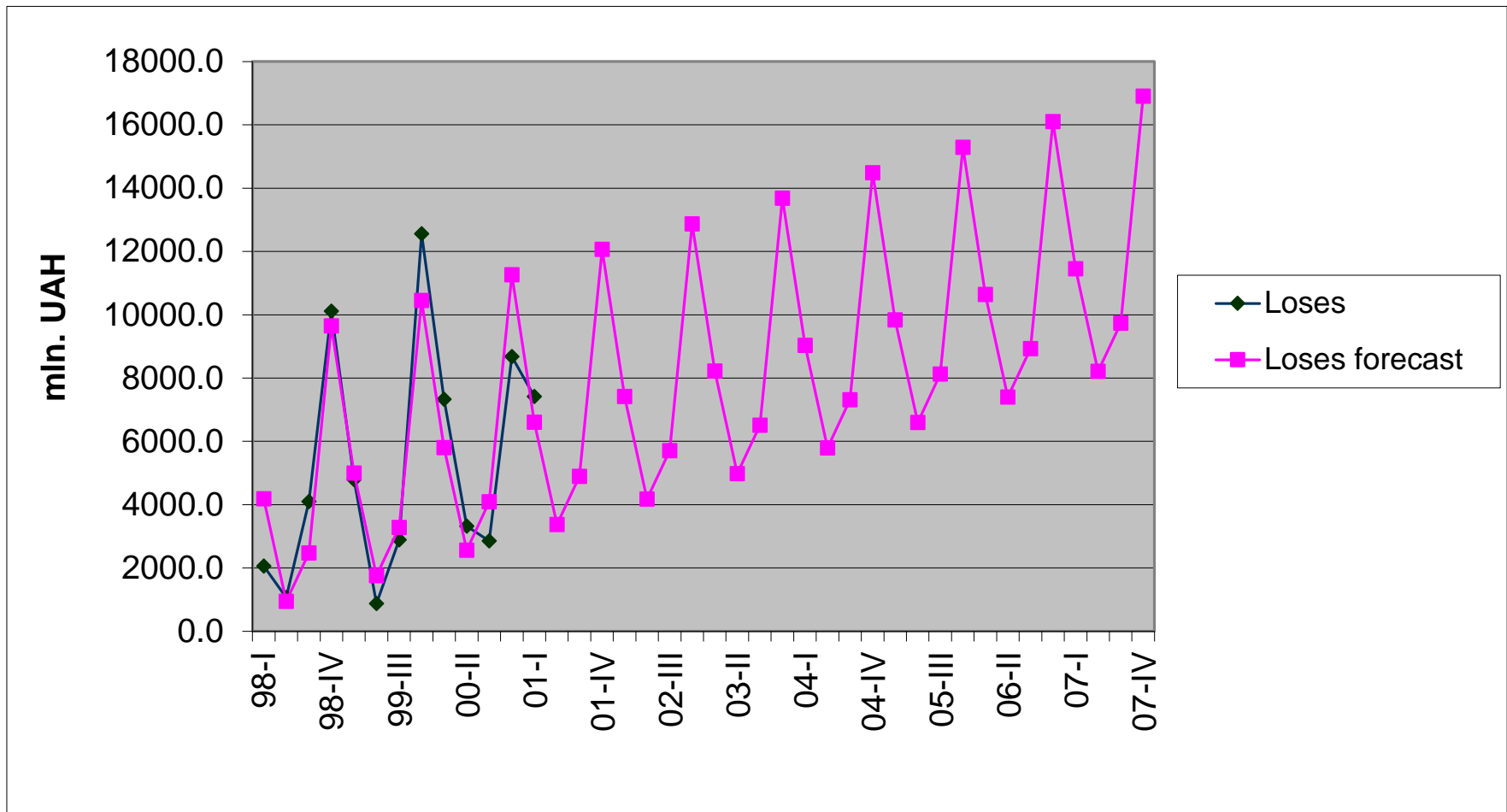
4 quarter

$$y_t = \beta_0 + \varepsilon_t$$

Example – 1



Example – 2



Dummy application

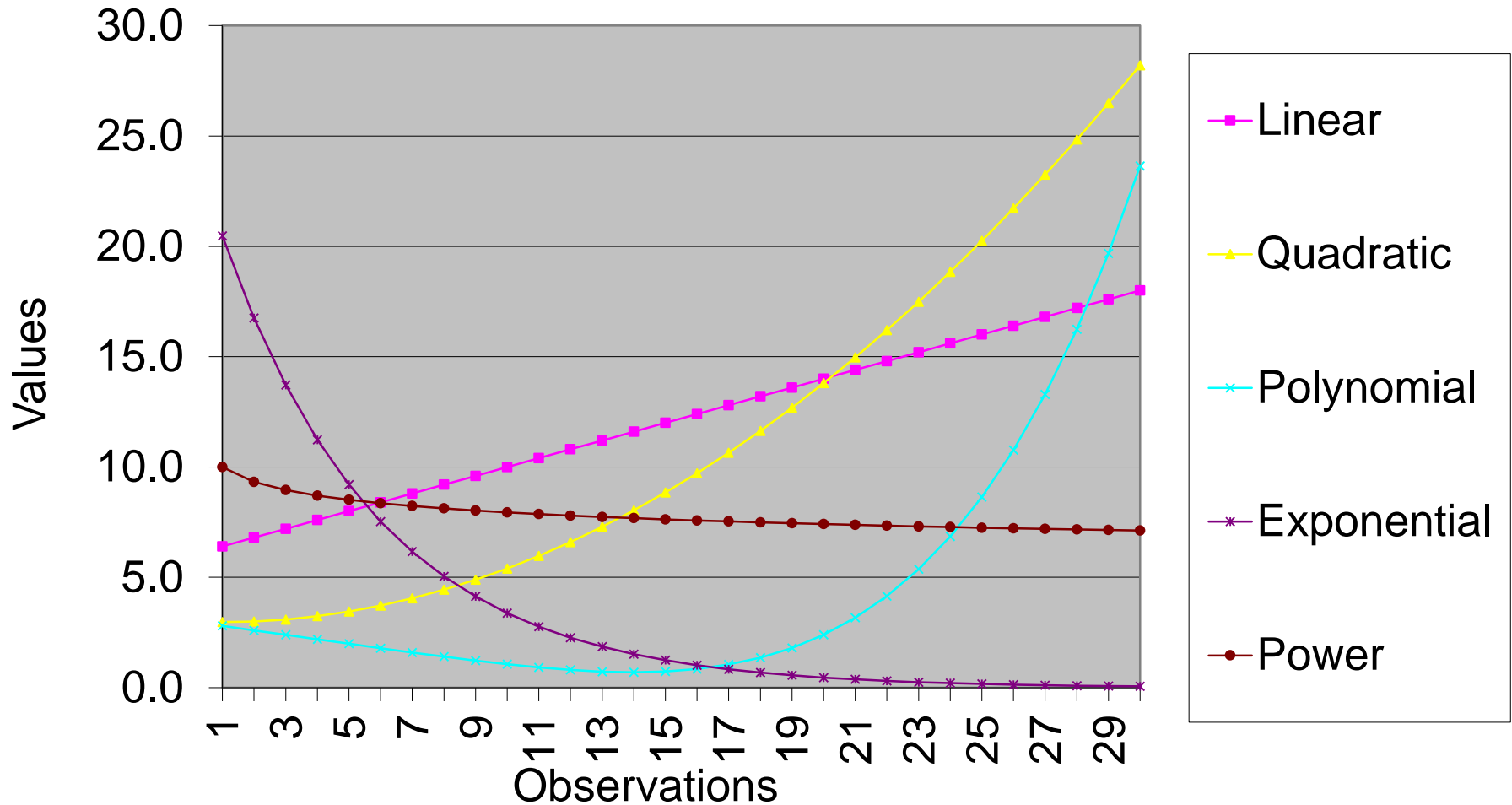
- Seasonality
- Crisis description
- Quality characteristics

Trend extraction

Trend functions

Linear	$f(t) = a_0 + a_1t$
Quadratic	$f(t) = a_0 + a_1t + a_2t^2$
Polynomial	$f(t) = a_0 + a_1t + a_2t^2 + \dots + a_nt^n$
Exponential	$f(t) = a_0 e^{a_1t}$
Power	$f(t) = a_0t^{a_1}$

Example



Trend Extraction in EViews

Quadratic trend example – 1

Equation Estimation ×

Specification Options

Equation specification
Dependent variable followed by list of regressors including ARMA and PDL terms, OR an explicit equation like $Y=c(1)+c(2)*X$.

g c @trend @trend^2

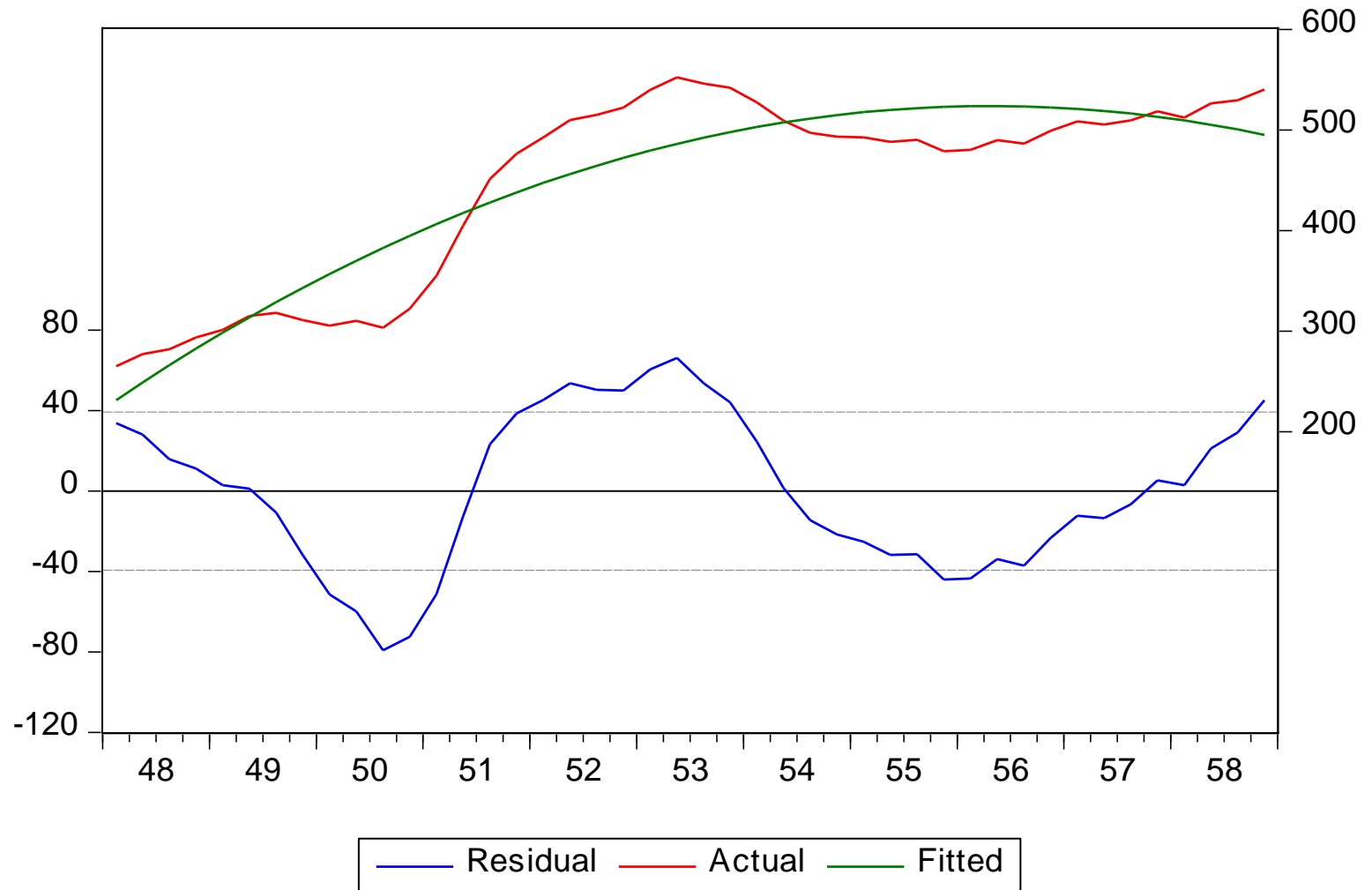
Estimation settings

Method: LS - Least Squares (NLS and ARMA) ▾

Sample: 1948q1 1958q4

OK Скасувати

Quadratic trend example – 2



Exponential trend example

$$f(t) = a_0 + a_1 e^t$$

Equation Estimation

Specification Options

Equation specification

Dependent variable followed by list of regressors including ARMA and PDL terms, OR an explicit equation like $Y=c(1)+c(2)^*X$.

y c exp(@trend)

Estimation settings

Method: LS - Least Squares (NLS and ARMA)

Sample: 1947q1 2005q4

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Smoothing Methods

Exponential smoothing

- Exponential smoothing is commonly applied to smoothen data, acting as low-pass filters to remove high frequency noise.
- New time series is build by the rule:

$$S_1 = y_1,$$
$$S_t = \alpha y_t + (1 - \alpha)S_{t-1}, \quad 0 < \alpha < 1.$$

- A constant can be found by different ways.

$$\alpha = \frac{2}{T + 1}.$$

Forecast

- The forecast is equal to the last value of smoothed series:

$$\hat{y}_{T+p} = S_T, \quad p = 1, 2, \dots$$

Example – 1



Double exponential smoothing

- This method uses exponential smoothing twice with the same coefficient.

$$S'_t = \alpha y_t + (1 - \alpha) S'_{t-1},$$

$$S''_t = \alpha S'_t + (1 - \alpha) S''_{t-1}, \quad 0 < \alpha < 1.$$

- Forecast

$$\hat{y}_{T+p} = S''_T, \quad p = 1, 2, \dots$$

Example



Triple exponential smoothing

- Uses exponential smoothing 3 times with the same constant.

$$S_t' = \alpha y_t + (1 - \alpha)S_{t-1}',$$

$$S_t'' = \alpha S_t' + (1 - \alpha)S_{t-1}'',$$

$$S_t''' = \alpha S_t'' + (1 - \alpha)S_{t-1}''', \quad 0 < \alpha < 1.$$

- Forecast

$$\hat{y}_{T+p} = S_T''', \quad p = 1, 2, \dots$$

Example



Non-seasonal Holt-Winters smoothing

- This model extract trend as a complimentary series:

$$S'_2 = y_2, \quad S''_2 = y_2 - y_1,$$

$$S'_t = \alpha y_t + (1 - \alpha)(S'_{t-1} + S''_{t-1}), \quad 0 < \alpha < 1,$$

$$S''_t = \beta(S'_t - S'_{t-1}) + (1 - \beta)S''_{t-1}, \quad 0 < \beta < 1.$$

- Forecast:

$$\hat{y}_{T+p} = S'_T + pS''_T, \quad p = 1, 2, \dots$$

Example



Multiplicative Winters smoothing

3 series are analysed:

- a_t - smoothed series,
- b_t - trend component,
- c_t - seasonality index

$$a_t = \alpha \left(\frac{y_t}{c_{t-s}} \right) + (1 - \alpha)(a_{t-1} + b_{t-1}),$$

$$b_t = \beta (a_t - a_{t-1}) + (1 - \beta) b_{t-1},$$

$$c_t = \gamma \left(\frac{y_t}{a_t} \right) + (1 - \gamma) c_{t-s}, \quad t = \overline{2s+1, T}.$$

S – number of season cycles

Forecast

$$\hat{y}_{T+p} = (a_T + pb_T) c_{T-s+p},$$

$$p = 1, 2, \dots, s,$$

$$\hat{y}_{T+p} = (a_T + pb_T) c_{T-2s+p},$$

$$p = s+1, s+2, \dots, 2s.$$

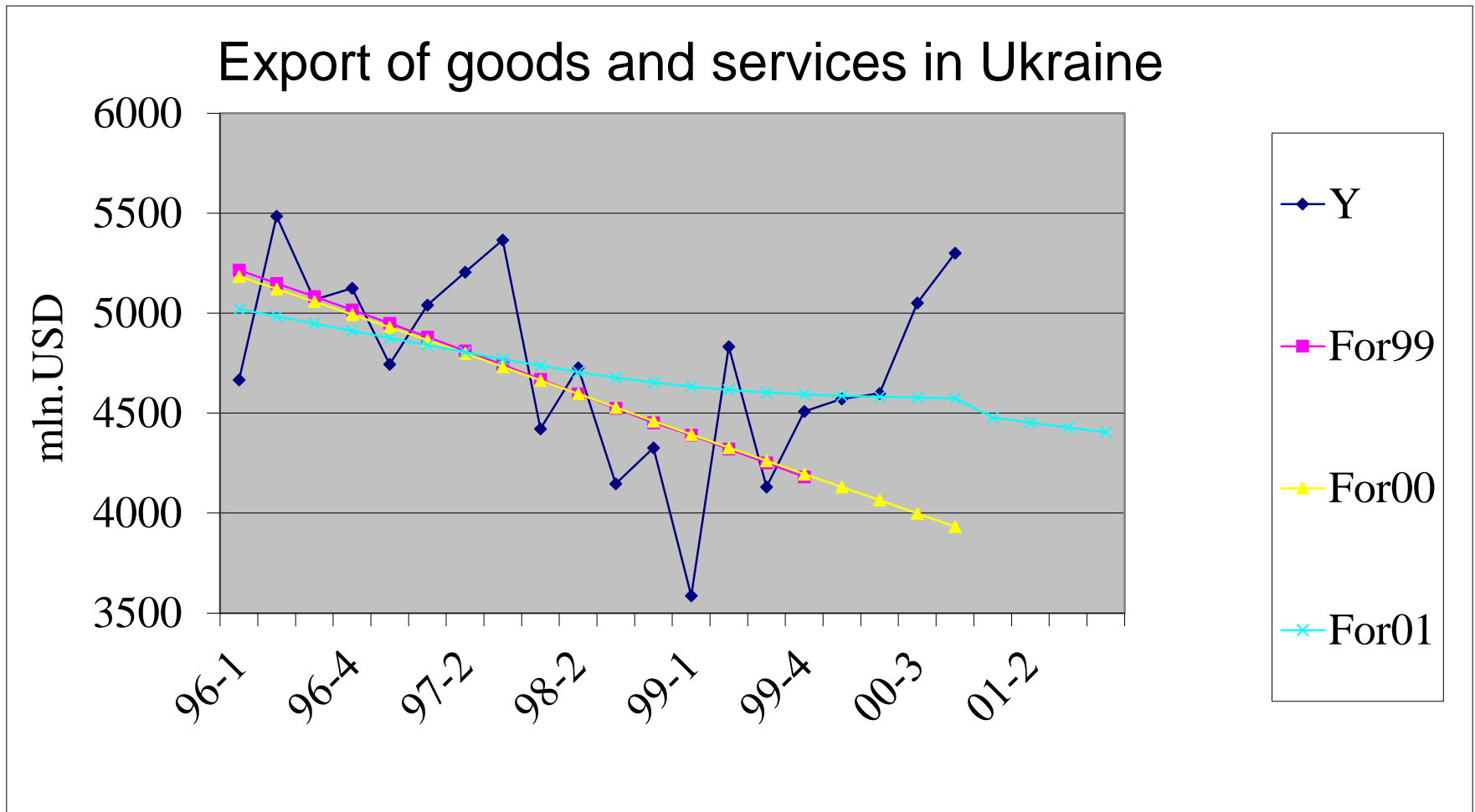
Hodrick-Prescott Filter

- Let $y_t = f(t) + \varepsilon_t$,

$$S = \sum_{t=1}^T (y_t - f(t))^2 + \lambda \sum_{t=2}^{T-2} \left((f(t+1) - f(t)) - (f(t) - f(t-1)) \right)^2 \rightarrow \min$$

- $\lambda=100$ for annual, $\lambda=1600$ for quarterly, $\lambda=14400$ for monthly data.

Example



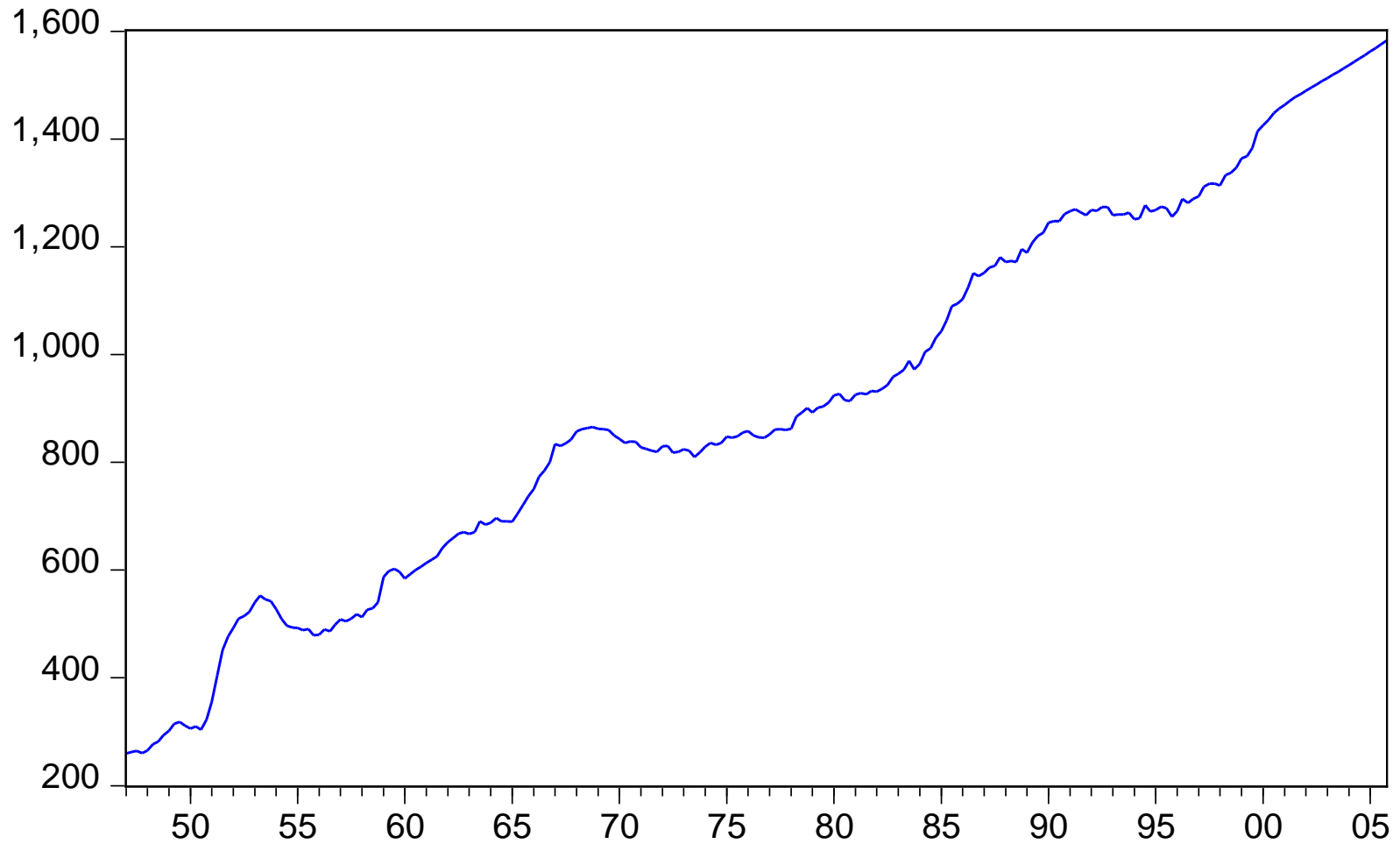
Smoothing methods in EIEWS

Example file

- Macromod.wf1
- Sample1: 1947q1 2005q4
- Sample2: 1947q1 1958q4

Government expenditures

G



Proc-Exponential smoothing- Simple Exponential smoothing

Exponential Smoothing ✕

Smoothing method	# of params
<input checked="" type="radio"/> Single	1
<input type="radio"/> Double	1
<input type="radio"/> Holt-Winters - No seasonal	2
<input type="radio"/> Holt-Winters - Additive	3
<input type="radio"/> Holt-Winters - Multiplicative	3

Smoothing parameters

Alpha: (mean) Enter number between 0 and 1, or E to estimate.

Beta: (trend)

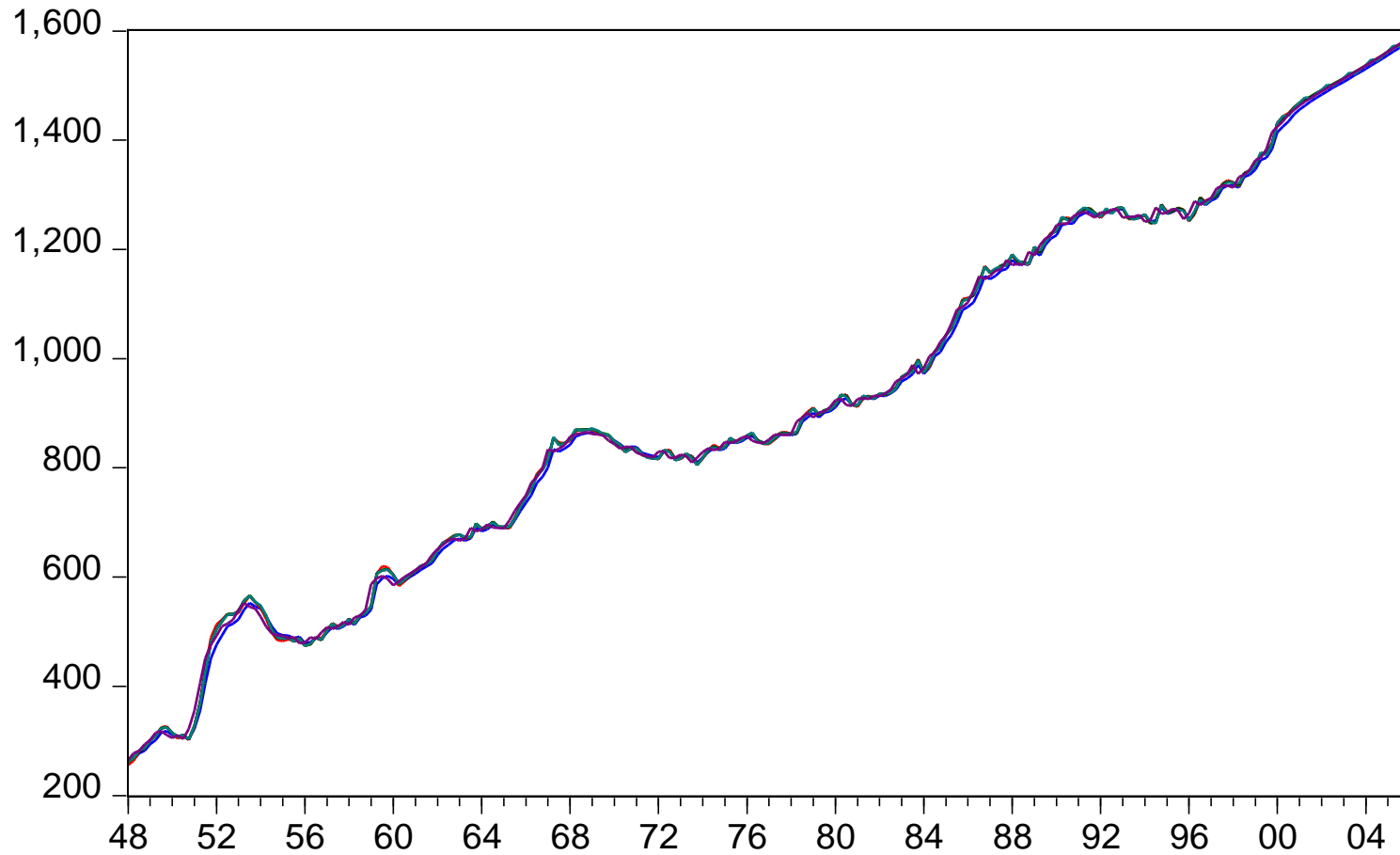
Gamma: (seasonal)

Smoothed series
Series name for smoothed and forecasted values.

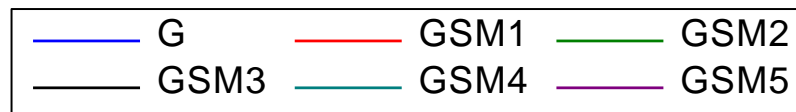
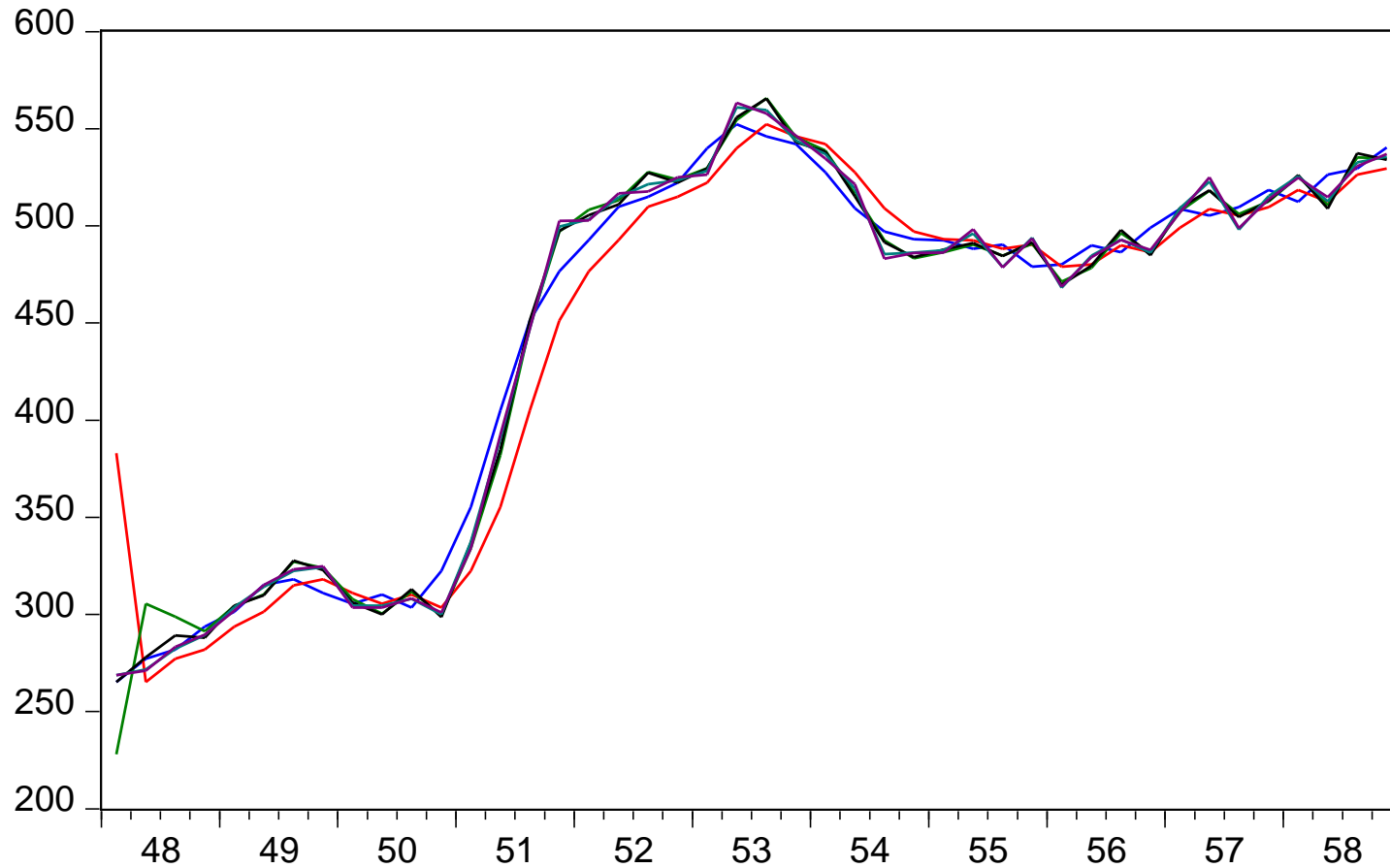
Estimation sample
Forecasts begin in period following estimation endpoint.

Cycle for seasonal

Outcome – 1



Outcome – 2



Proc-Hodrick-Prescott Filter...

Hodrick-Prescott Filter ✕

Output series

Smoothed series:

Cycle series:

Blank fields will not generate

Smoothing Parameter

Lambda:

Edit lambda directly

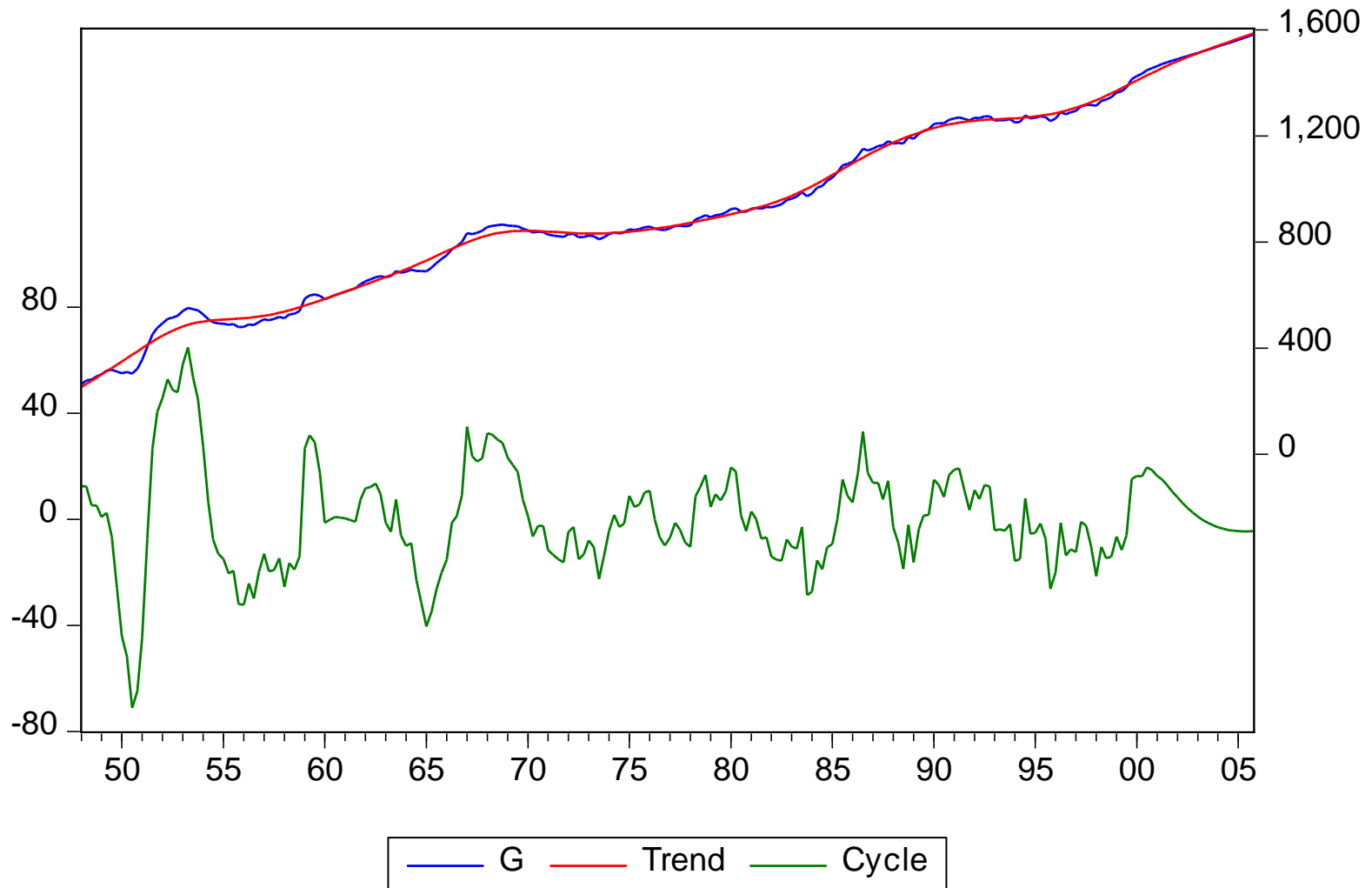
Set lambda by Ravn Uhlig frequency rule

Power:

Power does not matter for quarterly

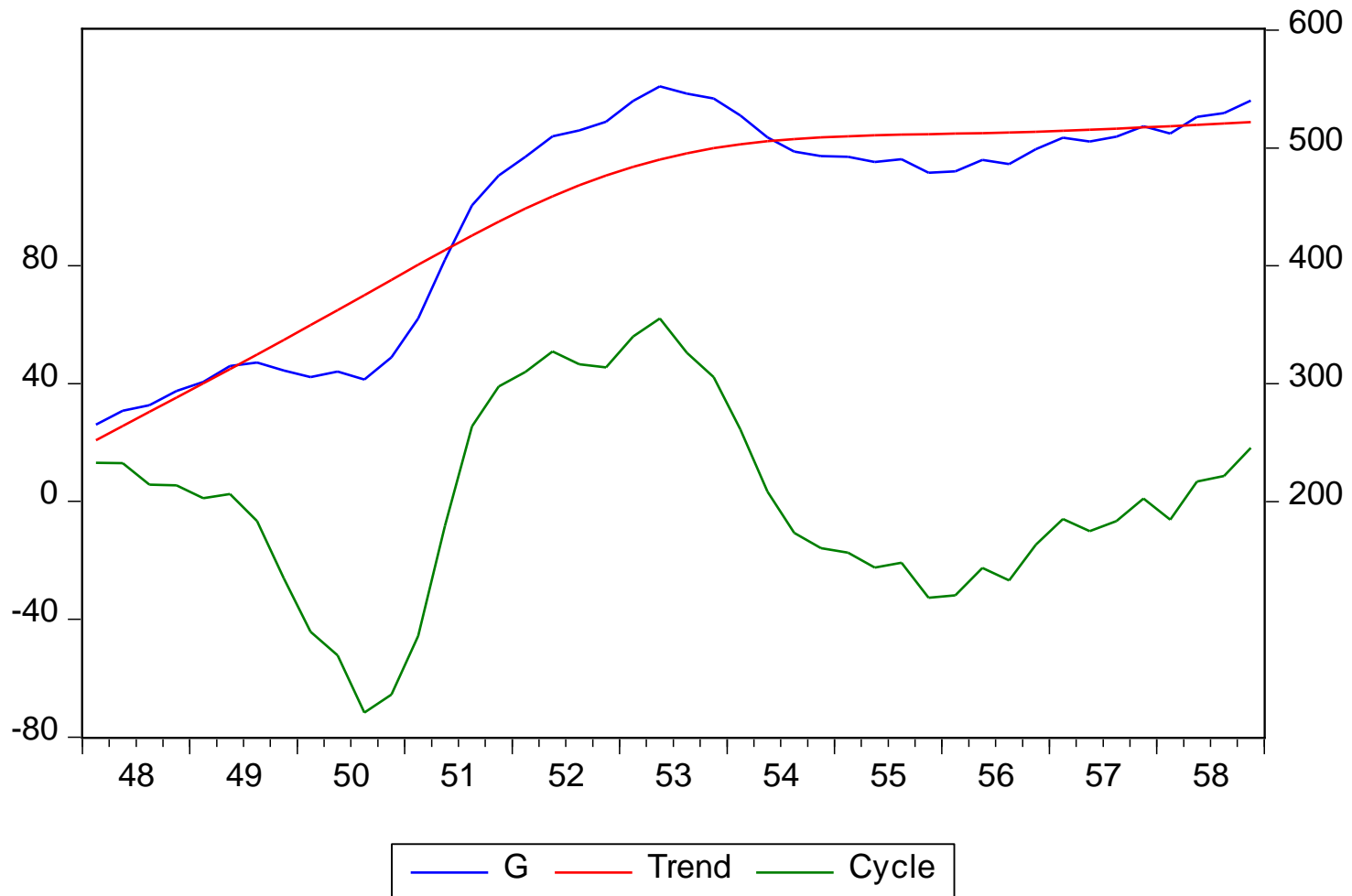
Outcome – 1

Hodrick-Prescott Filter (lambda=1600)



Outcome – 2

Hodrick-Prescott Filter (lambda=1600)



Review

Linear regression

$$y_t = \beta_0 + \beta_1 x_{1t} + \beta_2 x_{2t} + \dots + \beta_{k-1} x_{k-1t} + \varepsilon_t, t = \overline{1, n}$$

y_t - dependent variable;

$x_{1t}, x_{2t}, \dots, x_{k-1t}$ - independent variables;

ε_t - residuals.

Dummy application

- Seasonality
- Crisis description
- Quality characteristics

Trend functions

Linear	$f(t) = a_0 + a_1 t$
Quadratic	$f(t) = a_0 + a_1 t + a_2 t^2$
Polynomial	$f(t) = a_0 + a_1 t + a_2 t^2 + \dots + a_n t^n$
Exponential	$f(t) = a_0 e^{a_1 t}$
Power	$f(t) = a_0 t^{a_1}$

Smoothing

- Exponential smoothing
- Double exponential smoothing
- Triple exponential smoothing
- Non-seasonal Holt-Winters smoothing
- Multiplicative Winters smoothing

- Hodrick-Prescott Filter

Thank you for attention!