

Computer Exercise 0 – Outline Answers

Read the data set in

```
u cex0
```

To summarise the data

```
su unr inf
```

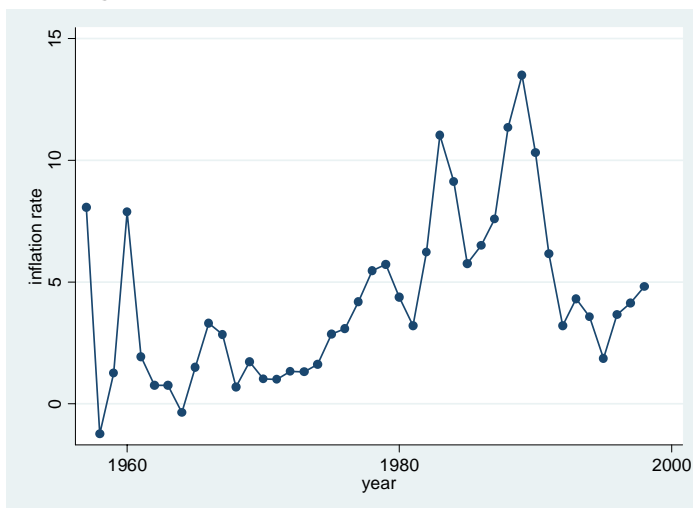
Variable	Obs	Mean	Std. Dev.	Min	Max
unr	42	5.652381	1.645873	2.9	9.7
inf	42	4.223176	3.412677	-1.244818	13.49863

so the mean unemployment rate (% of labour force unemployed) is 5.5% and the mean inflation rate (annual %age change in prices) is 4.2% with a minimum of -1.2% and maximum of 13.5%

To graph the data use the commands

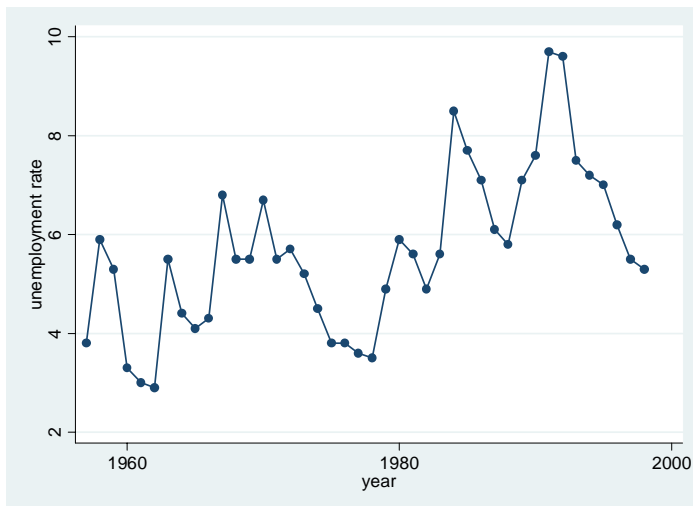
```
twoway connect inf year, xlabel(1960 1980 2000)
```

which gives



and

```
twoway connect unr year, xlabel(1960 1980 2000)
```



```
reg inf unr
```

Source	SS	df	MS	Number of obs =	42
Model	31.058555	1	31.058555	F(1, 40) =	2.78
Residual	446.442404	40	11.1610601	Prob > F =	0.1031
				R-squared =	0.0650
				Adj R-squared =	0.0417
Total	477.500959	41	11.6463649	Root MSE =	3.3408

inf	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
unr	.5288134	.3170039	1.67	0.103	-.1118753 1.169502
_cons	1.234121	1.864506	0.66	0.512	-2.534186 5.002428

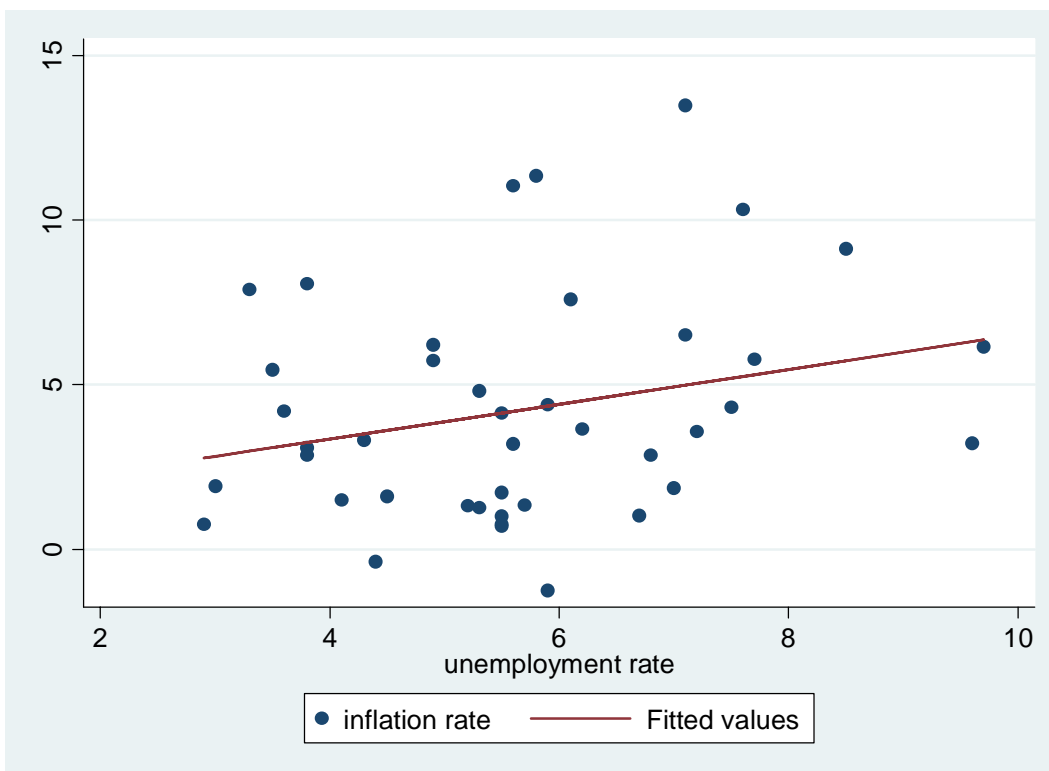
So estimates suggest a 1 **percentage point** rise in the unemployment rate leads to a 0.53 **percentage point** rise in inflation

To graph fitted and actual values, first save predicted values from the regression using the command

```
predict infhat
```

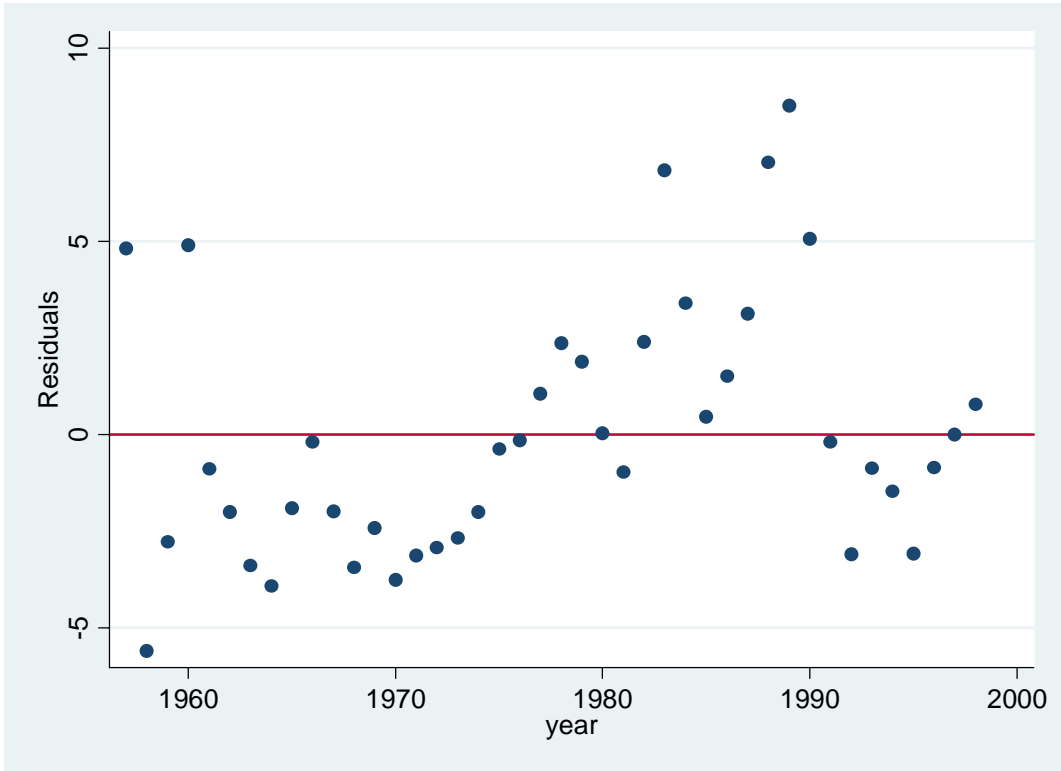
and then graph the results

```
graph twoway (scatter inf unr) (line yhat unr)
```



Can see even though this is the line of best fit (OLS) there are still large residual values for most observations. This is one way of finding out that the model does not summarise the variation in the dependent variable very well. The same impression is given by the graph of the residuals from the same regression.

```
predict reshat, resid  
twoway scatter uhat year, yline(0)
```



The model tends to overpredict inflation in the early years of inflation when inflation is low, ie $(y - \hat{y})$ is negative.

3. To rescale the data use the command

```
replace unr=unr/100
```

and then

```
reg inf unr
```

Source	SS	df	MS			
Model	31.0585592	1	31.0585592	Number of obs =	42	
Residual	446.4424	40	11.16106	F(1, 40) =	2.78	
Total	477.500959	41	11.6463649	Prob > F =	0.1031	
				R-squared =	0.0650	
				Adj R-squared =	0.0417	
				Root MSE =	3.3408	

inf	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
unr	52.88135	31.70039	1.67	0.103	-11.18753	116.9502
_cons	1.234121	1.864506	0.66	0.512	-2.534186	5.002428

Rescaling the independent variable multiplies the estimate of the **slope coefficient** by the inverse of the constant of multiplication (ie $1/1/100 = 100$). Note that the estimate of the constant is unchanged. Why?

```
replace inf=inf/100
```

```
reg inf unr
```

Source	SS	df	MS			
Model	.003105856	1	.003105856	Number of obs =	42	
				F(1, 40) =	2.78	
				Prob > F =	0.1031	

Residual		.044644239	40	.001116106		R-squared	=	0.0650

Total		.047750095	41	.001164636		Adj R-squared	=	0.0417

	inf		Coef.	Std. Err.	t	P> t		[95% Conf. Interval]

	unr		.5288135	.3170039	1.67	0.103		-.1118752 1.169502
	_cons		.0123412	.0186451	0.66	0.512		-.0253419 .0500243

Rescaling both explanatory and independent variable leaves both coefficients unchanged. Why?

4. Transforming independent variable into logs

```
g lu=log(unr)
```

```
. reg inf lu
```

Source		SS	df	MS		Number of obs	=	42

Model		.002699954	1	.002699954		F(1, 40)	=	2.40
Residual		.045050141	40	.001126254		Prob > F	=	0.1294

Total		.047750095	41	.001164636		R-squared	=	0.0565

	inf		Coef.	Std. Err.	t	P> t		[95% Conf. Interval]

	lu		.027337	.0176559	1.55	0.129		-.008347 .0630209
	_cons		.1219227	.0517292	2.36	0.023		.0173741 .2264713

sincec this is a semi-log model ($dinf/d\ln(unr) = dinf/(dunr/unr) = \text{unit change in inflation wrt \% change in unemployment}/100$).

So change in inflation = $(0.027/100) * \% \text{change in unemployment}$

In this case a 1% increase in the unemployment rate would raise inflation by 0.027/100 (0.027 percentage points)

```
replace unr=unr*100
```

```
g lu2=log(unr)
```

```
reg inf lu2
```

Source		SS	df	MS		Number of obs	=	42

Model		.002699954	1	.002699954		F(1, 40)	=	2.40
Residual		.045050141	40	.001126254		Prob > F	=	0.1294

Total		.047750095	41	.001164636		R-squared	=	0.0565

	inf		Coef.	Std. Err.	t	P> t		[95% Conf. Interval]

	lu2		.027337	.0176559	1.55	0.129		-.008347 .0630209
	_cons		-.0039687	.0302852	-0.13	0.896		-.0651773 .0572398

So variables measured in logs are invariant to the scale in which the original variable (unemployment) is measured. This is because the log specification implies a relative effect of the variable not an absolute one. (see question 6 on exercise 1)