# Tutorial 1

**keywords**: EViews, variables, data set, sample, population, descriptive analytics, summary statistics, histogram, scatter plots, correlation, simple linear regression model, predictive analytics

December 1, 2022

### Exploring EViews with wine.wf1

EViews is a computer software designed specifically for econometric analysis. Using EViews to perform basic econometric analysis on the workfile, wine.wf1.

### Objects

The elements in the whitespace of the *wine.wf1* workfile are called EViews **objects**. There are many types of EViews objects but in this unit we mostly focus on **series**, **groups**, **graphs**, **equations**, **scalar** and **coef** objects. Each object type is identified by a unique icon,

Icon	$\mathbf{Object}$		
	series		
	group		
	$\operatorname{graph}$		
	equation		
	scalar		
	$\operatorname{coef}$		

#### Data set

- Range: 1 21 21 obs indicates that the data set in wine.wf1 contains 21 observations starting at observation number 1 to 21.
- Sample: 1 21 21 obs denotes the sample which EViews will use to produce graphs, run regressions, and obtain other EViews outputs. The current sample contains all observations in the data set.

#### Variables

The *wine.wf1* workfile contains data on 21 countries. This data set holds information about each country through the following variables:

- alcohol litres of wine consumed per capita per annum
- deaths number of deaths per 100,000 of population
- heart number of deaths from heart disease per 100,000 of population
- liver number of deaths from liver disease per 100,000 of population

A description of each variable appears by clicking **Details** +/- on the workfile windows.

To view the data in a single variable, double click the variable. To view the data in multiple variables, select and highlight the variables of interest then,

$$Right\ click \rightarrow Open \rightarrow as\ Group$$

Descriptive analytics with histograms and descriptive statistics

To obtain the histogram and descriptive statistics of the variable alcohol,

double-click on the variable alcohol,

then,

 $View \rightarrow Descriptive\ Statistics\ \&\ Tests \rightarrow Histogram\ and\ Stats$ 

Figure 1: Histogram and descriptive statistics of litres of wine consumed per capita per annum from a sample of 21 countries.

As we can see from the figure, wine consumption per capita per annum is positively skewed (right-tailed). On average, countries consume approximately 2.838 litres of wine per capita per annum. The mean wine consumption is greater than the median wine consumption,

which also indicates that wine consumption is positively skewed.

We can also obtain descriptive statistics for multiple variables in a single spreadsheet. One way to do this is by selecting Quick from the EViews workfile menu,

 $Quick \rightarrow Group \ Statistics \rightarrow Descriptive \ Statistics \rightarrow Common \ sample$ 

then type in the variables of interest into the Series List window,

Note: The *Freeze* button takes a screenshot, while the *Name* button names and saves the object into the workfile.

#### Scatter plots and correlations

Scatter plots are graphs used to visually explore the relationship between two variables. To obtain a scatterplot of *deaths* (y-axis) against *alcohol* (x-axis),

$$Quick \rightarrow Graph \dots$$

in the Series List window, type the x-variable followed by the y-variable,

Under Specific, select Scatter and press OK,

$$Specific: Scatter \rightarrow OK$$

To give this graph the name scatter\_alcohol\_deaths, select Name,

then,

 $Name\ to\ identify\ object: scatter\_alcohol\_deaths$ 

Figure 2: Scatter plot of deaths per 100,000 of population against litres of wine consumption per capita per annum.

From the scatter plot, we observe a moderate negative linear relationship between wine consumption per capita per annum and deaths per 100,000 of population.

To obtain a table of sample correlation coefficients,

$$Quick \rightarrow Group \ Statistics \rightarrow Correlations$$

then type in the variables of interest into the Series List windows,

The numbers in the table above represent the following sample correlation coefficients:

$$\widehat{corr}(alcohol, deaths) = -0.41420$$
 $\widehat{corr}(alcohol, heart) = -0.66134$ 
 $\widehat{corr}(alcohol, liver) = 0.73719$ 
 $\widehat{corr}(deaths, heart) = 0.74980$ 
 $\widehat{corr}(deaths, liver) = -0.20189$ 

The sample correlation coefficient measures strength and direction of the linear relationship between 2 variables and takes on a value between -1 and 1.

- If the sample correlation coefficient equal to -1, then the two variables have a perfect negative linear relationship.
- If the sample correlation coefficient equal to 1, then the two variables have a perfect positive linear relationship.
- If the sample correlation coefficient equal to 0, then the two variables have no linear relationship.

Therefore, the magnitude of the sample correlation coefficient measures the strength of the linear relationship between two variables and the size of the sample correlation coefficient measures the direction of the linear relationship.

#### Estimating a simple linear regression model

#### Background

Simple Linear Regression Model

Simple - one independent variable

Linear - linear in the parameters

Suppose we specify a simple linear regression model of y on a constant (intercept) and x,

$$y = \beta_0 + \beta_1 x + u \tag{1}$$

where u is the error term that captures unobserved factors that affect y (factors other than x that affect y). It is included in the model because no matter how we specify the model, we cannot guarantee that the x input will perfectly output y i.e. there will always be some error.  $\beta_0$  and  $\beta_1$  are the intercept and slope coefficients respectively. They are population parameters and are unknown in practice.

To quantify the relationship between y and x, we need to estimate these population parameters. With data on y and x, we can use an estimator (more on this in subsequent weeks) to estimate (??), giving us estimates of  $\beta_0$  and  $\beta_1$  and the following estimated regression model,

$$\hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x \tag{2}$$

Suppose that the true population relationship between deaths per 100,000 in the population and wine consumption per capita per annum is represented by the following simple regression model,

$$deaths = \beta_0 + \beta_1 alcohol + u \tag{3}$$

To estimate our model, we need data on *deaths* and *alcohol*. The EViews workfile wine.wf1 contains data on *deaths* and *alcohol* from 21 countries i = 1, 2, ..., 21 (n = 21). We can express our model in terms of each country i,

$$deaths_i = \beta_0 + \beta_1 alcohol_i + u_i$$

$$i = 1, 2, \dots, 21$$
(4)

To estimate in EViews,

 $Quick \rightarrow Estimate\ Equation \dots$ 

then in the Equation Estimation windows type in,

 $deaths\ c\ alcohol$ 

Dependent Variable: DEATHS

Method: Least Squares

Sample: 1 21

Included observations: 21

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C ALCOHOL	$876.2050 \\ -16.26352$	30.46816 8.198924	28.75805 -1.983616	0.0000 0.0619
R-squared Adjusted R-squared S.E. of regression Sum squared resid Log likelihood F-statistic Prob(F-statistic)	$\begin{array}{c} 0.171562 \\ 0.127960 \\ 90.13207 \\ 154352.0 \\ -123.2736 \\ 3.934731 \\ 0.061939 \end{array}$	Mean deper S.D. depend Akaike info Schwarz cri Hannan-Qu Durbin-Wat	lent var criterion terion inn criter.	830.0476 96.51864 11.93082 12.03030 11.95241 1.964148

Table 1: Regression output of deaths on a constant and alcohol

We report the estimated model by placing a 'hat' above the dependent variable and the standard error of the estimated coefficient in parenthesis underneath the estimated coefficient,

$$\widehat{deaths}_{i} = 876.2050 - 16.2635alcohol_{i}$$

$$(30.4682) \qquad (8.1989)$$

$$i = 1, 2, \dots, 21$$

To name/save this regression output into our workfile,

 $Name \rightarrow Name \ to \ indentify \ object : eq01 \rightarrow OK$ 

### Predictive analytics

So far, we have generated outputs in EViews by 'pointing-and-clicking'. We can also obtain these outputs by executing lines of code through the **Command** window,

The Command window can also functions as a calculator. For example, we can use our estimated regression model,

$$\widehat{deaths}_i = 876.2050 - 16.2635alcohol_i$$

$$i = 1, 2, \dots, 21$$

to calculate the number of death per 100,000 of the population for a country that consumes  $\underline{3}$  litres of wine per capita per annum,

$$\widehat{deaths} = 876.2050 - 16.2635 \times 3$$

This is an example of predictive analytics. We have estimated a regression model of *deaths* and used this model to predict the number of death per 100,000 when a country consumes 3 litres of wine per capita per annum.

From the Command window, executing the code

$$scalar\ prediction = c(1) + c(2)^*3$$

will store the calculation into a scalar object named prediction.

Note: c(1) and c(2) represents the  $1^{st}$  and  $2^{nd}$  estimated coefficient of the <u>most recent</u> estimated regression model i.e. the values 876.2050 and -16.2635