

Instrument Variable EE12.2

我們直接從課本的 Empirical Exercise 12.2 來複習如何用 Stata 處理內生性問題，並實作 2SLS。

E12.2 Does viewing a violent movie lead to violent behavior? If so, the incidence of violent crimes, such as assaults, should rise following the release of a violent movie that attracts many viewers. Alternatively, movie viewing may substitute for other activities (such as alcohol consumption) that lead to violent behavior, so that assaults should fall when more viewers are attracted to the cinema. On the text website, <http://www.pearsonglobaleditions.com>, you will find the data file **Movies**, which contains data on the number of assaults and movie attendance for 516 weekends from 1995 through 2004.⁷ A detailed description is given in **Movies_Description**, available on the website. The data set includes weekend U.S. attendance for strongly violent movies (such as *Hannibal*), mildly violent movies (such as *Spider-Man*), and nonviolent movies (such as *Finding Nemo*). The data set also includes a count of the number of assaults for the same weekend in a subset of counties in the United States. Finally, the data set includes indicators for year, month, whether the weekend is a holiday, and various measures of the weather.

EE12.2 Questions

- a.
 - i. Regress the logarithm of the number of assaults [$\ln_assaults = \ln(assaults)$] on the year and month indicators. Is there evidence of seasonality in assaults? That is, do there tend to be more assaults in some months than others? Explain.
 - ii. Regress total movie attendance ($attend = attend_v + attend_m + attend_n$) on the year and month indicators. Is there evidence of seasonality in movie attendance? Explain.
- b. Regress $\ln_assaults$ on $attend_v$, $attend_m$, $attend_n$, the year and month indicators, and the weather and holiday control variables available in the data set.
 - i. Based on the regression, does viewing a strongly violent movie increase or decrease assaults? By how much? Is the estimated effect statistically significant?
 - ii. Does attendance at strongly violent movies affect assaults differently than attendance at moderately violent movies? Differently than attendance at nonviolent movies?

⁷These are aggregated versions of data provided by Gordon Dahl of University of California–San Diego and Stefano DellaVigna of University of California–Berkeley and were used in their paper “Does Movie

EE12.2 Questions

- iii. A strongly violent blockbuster movie is released, and the weekend's attendance at strongly violent movies increases by 6 million; meanwhile, attendance falls by 2 million for moderately violent movies and by 1 million for nonviolent movies. What is the predicted effect on assaults? Construct a 95% confidence interval for the change in assaults. [*Hint:* Review Section 7.3 and material surrounding Equations (8.7) and (8.8).]
- c. It is difficult to control for all the variables that affect assaults and that might be correlated with movie attendance. For example, the effect of the weather on assaults and movie attendance is only crudely approximated by the weather variables in the data set. However, the data set does include a set of instruments—*pr_attend_v*, *pr_attend_m*, and *pr_attend_n*—that are correlated with attendance but are (arguably) uncorrelated with weekend-specific factors (such as the weather) that affect both assaults and movie attendance. These instruments use historical attendance patterns, not information on a particular weekend, to predict a film's attendance in a given weekend. For example, if a film's attendance is high in the second week of its release, then this can be used to predict that its attendance was also high in the first week of its release. (The details of the construction of these instruments are available in the Dahl and DellaVigna paper referenced in footnote 5.) Run the regression from (b) (including year, month, holiday, and weather controls) but now using *pr_attend_v*, *pr_attend_m*, and *pr_attend_n* as instruments for *attend_v*, *attend_m*, and *attend_n*. Use this IV regression to answer (b)(i)–(b)(iii).

- d.** The intuition underlying the instruments in (c) is that attendance in a given week is correlated with attendance in surrounding weeks. For each movie category, the data set includes attendance in surrounding weeks. Run the regression using the instruments *attend_v_f*, *attend_m_f*, *attend_n_f*, *attend_v_b*, *attend_m_b*, and *attend_n_b* instead of the instruments used in (c). Use this IV regression to answer (b)(i)–(b)(iii).
- e.** There are nine instruments listed in (c) and (d), but only three are needed for identification. Carry out the test for overidentification summarized in Key Concept 12.6. What do you conclude about the validity of the instruments?
- f.** Based on your analysis, what do you conclude about the effect of violent movies on (short-run) violent behavior?

EE12.2 Data Descriptions

Movie Data

1. Observations: 516 weekends
2. Time Period : 1995-2004

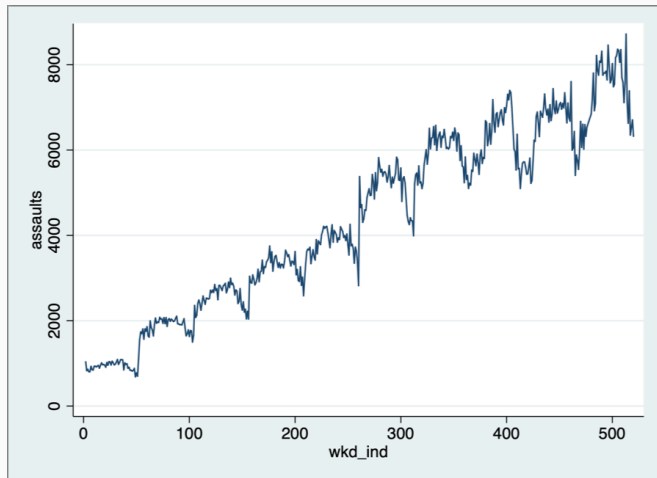
Variable Name	Description
<i>Assaults and Movie Attendance</i>	
assaults	number of assaults and intimidation in a subset of U.S. counties
attend v	attendance strongly violent movies (in millions)
attend m	attendance mildly violent movies (in millions)
attend n	attendance nonviolent movies (in millions)
<i>Weather, Holiday and Calendar Variables</i>	
year1 to year10	indicator variable for year of the sample (1995-2004)
month1 to month12	indicator variables for month of the year (January-December)
h chris	indicator variable for Christmas weekend
h newyr	indicator variable for New Years weekend
h easter	indicator variable for Easter weekend
h july4	indicator variable for July 4 (U.S. Independence Day) weekend
h mem	indicator variable for Memorial Day weekend
h labor	indicator variable for Labor Day weekend
w rain	fraction of locations with rain
w snow	fraction of locations with snow
w maxa	fraction of locations with maximum daily temperature between 80°F and 90°F
w maxb	fraction of locations with maximum daily temperature between 90°F and 100°F
w maxc	fraction of locations with maximum daily temperature greater than 100°F
w mina	fraction of locations with minimum daily temperature less than 10°F
w minb	fraction of locations with minimum daily temperature between 10°F and 20°F
w minc	fraction of locations with minimum daily temperature between 20°F and 32°F
<i>Instruments</i>	
pr attend v	predicted attendance violent movies
pr attend m	predicted attendance moderately violent movies
pr attend n	predicted attendance nonviolent movies
attend v f	attendance violent movies one week in the future
attend m f	attendance moderately violent movies one week in the future
attend n f	attendance nonviolent movies one week in the future
attend v b	attendance violent movies one week in the past
attend m b	attendance moderately violent movies one week in the past
attend n b	attendance nonviolent movies one week in the past

EE12.2 a(i.)

To detect whether there is time trend or not:

$$\log(\text{assaults}) = \beta_0 + \psi_1 \text{year} + \psi_2 \text{month} + u$$

Or just simply graph a twoway plot with the time indicator.

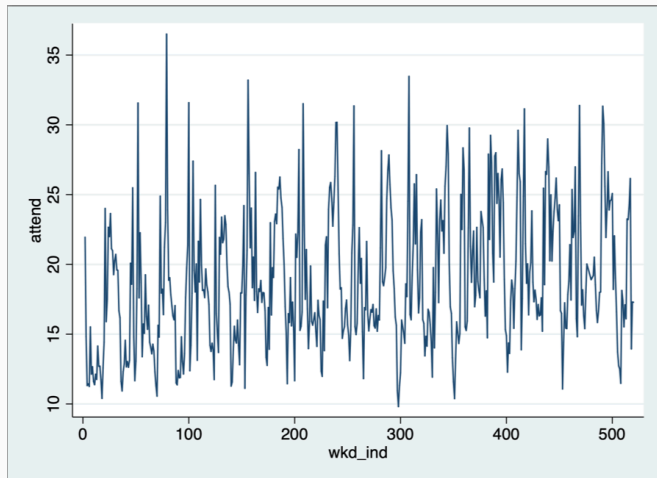


EE12.2 a(ii.)

To detect whether there is time trend or not:

$$attendance = \beta_0 + \phi_1 year + \phi_2 month + \tilde{u}$$

Or just simply graph a twoway plot with the time indicator.



The original model is:

$$\log(\text{assaults}) = \beta_0 + \beta_1 \text{attend_v} + \beta_2 \text{attend_m} + \beta_3 \text{attend_n} + \phi \text{Time} + \psi \text{Controls}$$

And the variables of interest are `attend_v`, `attend_m`, `attend_n`

Till now, we do not say the variables of interest are exog. or endog.

EE12.2 b(iii.)

The original model is:

$$\log(\text{assaults}) = \beta_0 + \beta_1 \text{attend_v} + \beta_2 \text{attend_m} + \beta_3 \text{attend_n} + \phi \text{Time} + \psi \text{Controls}$$

What we want to know is the change in the dependent variable:

$$\Delta \text{assaults}.$$

Let's start from:

$$\Delta \log(\text{assaults}) = \beta_1 \Delta \text{attend_v} + \beta_2 \Delta \text{attend_m} + \beta_3 \Delta \text{attend_n}$$

Then,

$$\Delta \log(\widehat{\text{assaults}}) = \hat{\beta}_1 \Delta \text{attend_v} + \hat{\beta}_2 \Delta \text{attend_m} + \hat{\beta}_3 \Delta \text{attend_n}$$

With the description in the questions:

$$\Delta \log(\widehat{\text{assaults}}) = 6\hat{\beta}_1 - 2\hat{\beta}_2 - \hat{\beta}_3$$

$$\Delta \log(\widehat{assaults}) = 6\hat{\beta}_1 - 2\hat{\beta}_2 - \hat{\beta}_3$$

And we can easily calculate the estimate: $\Delta \log(\widehat{assaults}) = -.01063206$

That is: $\Delta \widehat{assaults} \approx e^{-.01063206}$

Given $\Delta \log(\widehat{assaults}) = 6\hat{\beta}_1 - 2\hat{\beta}_2 - \hat{\beta}_3$

To obtain $se(\Delta \log(\widehat{assaults})) = se(6\hat{\beta}_1 - 2\hat{\beta}_2 - \hat{\beta}_3)$,

we need to apply approaches in section 7.3.

We may simplify the model by:

$$y = \beta_0 + \beta_1 v + \beta_2 m + \beta_3 n + u$$

Now let's focus on $\beta_1 v + \beta_2 m + \beta_3 n$ only.

Our goal is to have a explanatory variable which has a coefficient equals to the above number.

Simplified model:

$$y = \beta_0 + \beta_1 v + \beta_2 m + \beta_3 n + u$$

$$y = \beta_1 v + \beta_2 m + \beta_3 n - 6\beta_1 n + 2\beta_2 n + 6\beta_1 n - 2\beta_2 n + u$$

$$y = \beta_1(v + 6n) + \beta_2(m - 2n) + (\beta_3 - 6\beta_1 + 2\beta_2)n + u$$

Now we can simply use OLS to obtain the s.e.

Now we want to use IVs.

Notations:

Y : *dependent variable*

X : *possibly endog. explanatory variables*

W : *exog. explanatory variables*

Z : *instrument variables*

Recall the 2SLS:

- Regress X on Z, W
- Obtain \hat{X}
- Regress Y on \hat{X}, W
- Obtain the coef. of \hat{X}

Denote

$Y : \log(\text{assaults})$

$X : \text{attend_v}, \text{attend_m}, \text{attend_n}$

$W : \text{Time}, \text{Holiday}, \text{Weather}$

$Z : \text{pr_attend_v}, \text{pr_attend_m}, \text{pr_attend_n}$

which are predictions based on historical attendance patterns.

THINK: Why are these variable instruments?

We'll demonstrate the manual 2SLS approach first.

We may simply use `ivreg` or `ivregress`

The format would be:

```
ivreg Y (X=Z) W, r
```

or:

```
ivregress 2sls Y (X=Z) W, r
```

```
ivregress gmm Y (X=Z) W, r
```

Now change the IVs:

$Z : \text{attend_v_f}, \text{attend_m_f}, \text{attend_n_f},$

$\text{attend_v_b}, \text{attend_m_b}, \text{attend_n_b}$

which are the lagged terms and the future terms.

For Overidentifying Restriction Test, we first apply J-Test in textbook p.449, then we'll demonstrate a simple and more general command in Stata.

Note that the model now is:

$$Y = \beta_{2SLS} \hat{X} + \gamma W + u$$

We may obtain $\hat{u}_{2SLS} = Y - \hat{\beta}_{2SLS} \hat{X} + \hat{\gamma} W$

Then we regress \hat{u}_{2SLS} on Z, W , that is:

$$\hat{u}_{2SLS} = \delta Z + \tilde{\gamma} W + e$$

And we can obtain the J-statistic from $J = mF$ where

m = the number of instruments

F = the joint test statistic of $\delta = 0$

k = the number of exog. variables

Under homosk., $J \sim \chi^2(m - k)$

An easy command is:

```
ivregress Y (X=Z) W, r  
estat overid
```

will give you the test statistic under robust variance covariance estimator.