# CHAPTER 10 Regression with Panel Data



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#### **Definition of Panel Data**

- Panel data = consist of observations on the same n entities at two or more time periods T. If the data set contains observations on the variables X and Y, then the data are denoted:
- where the first subscript, *i*, refers to the entity being observed, and the second subscript, *t*, refers to the date at which it is observed.
- Those data are for ? entities (states), where each entity is observed in ? time periods, for a total of ? observations.



# **Example: Traffic Fatalities and Alcohol Taxes**

- There are approximately 40,000 highway traffic fatalities each year in the U.S.
- Approximately one-third of fatal crashes involve a driver who was drinking, and this fraction rises during peak drinking periods.
- One study estimates that as many as ? of drivers on the road between 1 A.M. and 3 A.M. have been drinking, and that a driver who is legally drunk is at least ? times as likely to cause a fatal crash as a driver who has not been drinking.



# **Example: Traffic Fatalities and Alcohol Taxes**

- The measure of traffic deaths: **fatality rate** = the number of annual traffic deaths per 10,000 people in the population in the state.
- The measure of alcohol taxes: "real" tax on a case of beer
  = the beer tax put into 1988 dollars by adjusting for inflation.



fatarate = 2.0104 +0.1485 beertax





FataRate = 2.01 + 0.15BeerTax(0.15) (0.13)

F = 0.62

 $R^2 = 0.01$ 

Standard Error of Regression = 0.67

 $\overline{FataRate} = 2.09$ 





beertax

FataRate = 1.86 + 0.44BeerTax(0.11) (0.13) F = 7.12

 $R^2 = 0.134$ 

Standard Error of Regression = 0.49

 $FataRate_{1988} = 2.07$ 

- Should we conclude that an increase in the tax on beer leads to more traffic deaths?
- These regressions could have substantial





### Many factors affect the fatality rate

- The quality of the automobiles driven in the state.
- Whether the state highways are in good repair.
- The density of cars on the road.
- Whether it is socially acceptable to drink and drive.
- Some of these variables might be very hard or even impossible to measure.
- If these factors remain constant over time in a given state, then another route is available. Because we have panel data, we can in effect hold these factors constant, even though we cannot measure them.

## 10.2 Panel Data with Two Time Periods: "Before and After" Comparisons

- Let Z<sub>i</sub> be a variable that determines the fatality rate in the *i*<sup>th</sup> state, but does not change over time (so the *t* subscript is omitted). Ohio specific factor.
- For example,  $Z_i$  might be the local cultural attitude toward drinking and driving, which changes slowly and thus could be considered to be constant between 1982 and 1988.

 $(FatalityRate_{i1988} - FatalityRate_{i1982})$ 

$$=\beta_1 \left(BeerTax_{i1988} - BeerTax_{i1982}\right) + \left(\varepsilon_{i1988} - \varepsilon_{i1982}\right)$$

Specifying the regression in changes eliminates the effect of the unobserved variables Z<sub>i</sub> that are constant over time. In other words, analyzing changes in Y and X has the effect of controlling for variables that are constant over time, thereby eliminating this source of ? bias.





diffBT

# **Estimation result**

 $(FatalityRate_{i1988} - FatalityRate_{i1982})$ = -0.072 - 1.04 (BeerTax\_{i1988} - BeerTax\_{i1982}) (0.065) (0.36) F = 6.22 R<sup>2</sup> = 0.12 Standard Error of Regression = 0.394  $\overline{\Delta FataRate} = -0.0195$ 

- An increase in the real beer tax by \$1 per case reduces the traffic fatality rate by ? deaths per 10,000 people.
- By examining changes in the fatality rate over time, the regression controls for fixed factors such as cultural attitudes toward drinking and driving.

## **10.3 Fixed Effects Regression**

Purpose: Fixed effects regression is a method for controlling for ? in panel data when the omitted variables vary across entities (states, individuals, groups) but do not change over time.

#### Model

$$FatalityRate_{it} = \beta_0 + \beta_1 BeerTax_{it} + \beta_2 Z_i + \varepsilon_{it}$$

 Z<sub>i</sub> is an unobserved variable that varies from one state to the next but does not change over time (for example, cultural attitudes toward drinking & driving).

#### $FatalityRate_{it} = \beta_0 + \beta_1 BeerTax_{it} + \beta_2 Z_i + \varepsilon_{it}$

- Because Z<sub>i</sub> varies from one state to the next but is constant over time, the model can be interpreted as having n intercepts, one for each state.
- Rewrite the model as:

- $\alpha_i =$
- The slope coefficient of the population regression line,  $\beta_1$ , is the same for all states, but the intercept of the population regression line varies from one state to the next.

### $FatalityRate_{it} = \beta_1 BeerTax_{it} + \alpha_i + \varepsilon_{it}$

- The state-specific intercepts  $\alpha_i$ 's can be expressed using binary variables to denote the individual states.
- We cannot include all *n* binary variables plus a common intercept (perfect multicollinearity called dummy variable trap). So we omit the dummy variable for the first group.
   Rewrite

#### **Estimation result**

*FatalityRate* =

 $3.48 - 0.66BeerTax - 0.57D_{AZ} + 0.52D_{GA} - 2.19D_{NY} + 0.56D_{SC} + \dots$   $(0.31) \quad (0.2) \qquad (0.27) \quad (0.18) \qquad (0.3) \qquad (0.11)$ 

F = 57

 $R^2 = 0.905$ 

Standard Error of Regression = 0.19

FataRate = 2.04

• An increase in the real beer tax by \$1 per case reduces the traffic fatality rate by ? deaths per 10,000 people.

FR = 3.4776 -0.6559 beertax -0.5677 AZ -0.655 AR -1.5095 CA -1.4843 CO -1.8623 CT -1.3076 DE -0.2681 FL +0.5246 GA -0.669 ID -1.9616 IL -1.4615 IN -1.5439 IA -1.2232 KS -1.2175 KY -0.8471 LA -1.1079 ME -1.7064 MD -2.1097 MA -1.4845 MI -1.8972 MN -0.0291 MS -1.2963 MD -0.3604 MT -1.5222 NE -0.6008 NV -1.2545 NH -2.1057 NJ +0.4264 NM -2.1867 NY -0.2905 NC -1.6234 ND -1.6744 OH -0.5451 OK -1.168 OR -1.7675 PA -2.2651 RI +0.5572 SC -1.0037 SD -0.8757 TN -0.9175 TX -1.164 UT -0.966 VT -1.2902 VA -1.6595 WA -0.8968 WV -1.7593 WI -0.2285 WY



beertax

$$FatalityRate_{it} = \beta_1 BeerTax_{it} + \alpha_i + \varepsilon_{it}$$

- What are other sources of omitted variable bias?
- Over this period cars were getting ? and occupants were increasingly wearing ?
- If the real tax on beer rose on average during the mid-1980s, then it could be picking up the effect of overall automobile safety improvements.
- If safety improvements evolved over time but were the same for all states, then we can eliminate their influence by including time fixed effects.

# **10.4 Regression with Time Fixed Effects**

- Time fixed effects = control for variables that are constant across entities but evolve over time.
- It is plausible to think of automobile safety as an omitted variable that changes over time but has the same value for all states.
- Entity & Time fixed effect regression model

 $FatalityRate_{it} = \beta_0 + \beta_1 BeerTax_{it} + \beta_2 Z_i + \beta_3 S_t + \varepsilon_{it}$ 









✓ Where  $\alpha_i$  is the ? and  $\lambda_t$  is the ? . ✓ This model can equivalently be represented using n - 1 ? variables and T - 1 ? variables, along with an intercept:

$$\begin{aligned} FatalityRate_{it} &= \beta_0 + \beta_1 BeerTax_{it} \\ &+ \gamma_{AZ} D_{AZ} ... + \gamma_{WY} D_{WY} + \delta_{83} D_{83} ... + \delta_{88} D_{88} + \varepsilon_{it} \end{aligned}$$

#### **Estimation result**

FatalityRate = 3.51 - 0.64BeerTax(0.25)

 $-0.55D_{AZ} + \dots - 0.20D_{WY} - 0.08D_{1983} \dots - 0.05D_{1988}$ 

 $AdjR^2 = 0.89$  F = 51.93 SER = 0.188

• This specification includes the beer tax, 47 state dummy variables, 6 year dummy variables, and an intercept, so that this regression actually has right hand variables.

• An increase in the real beer tax by \$1 per case reduces the traffic fatality rate by ? deaths per 10,000 people.