1.	What is Econometrics?
2	Steps in empirical economic analysis
3.	The structure of economic data
4.	Causality and the notion of ceteris
	paribus in econometric analysis

♦Econor	netrics is a kind of statistical	
method		
Estin	nating economic relationships,	
 Testi 	ng economic theories, and	
 Evalu 	ating & implementing policy	
etc		



I .	2 Steps in Empirical Analysis
۲	An empirical analysis uses data to test a theory or to estimate a relationship
1.	Constructing economic model wage = f(educ, exper, training) (1.2)
2.	Specifying econometric model wage = $\beta_0 + \beta_1 educ + \beta_2 exper + \beta_3 training + u$ (1.4) • <i>u</i> is error term, and β s are parameters.

•••	3 Structure of economic data
<u>C</u>	ross-sectional data
۲	Each observation is a new individual, firm
	etc. with information at a point in time.
	→e.g. table 1.1 & 1.2
۲	Cross-sectional data is a <i>random sample</i> .
	• If the data is not a random sample, we have a sample-selection problem.

A Cross-S	ectional Data	Set on Wage	s and Other I	ndividual Cha	racteristics
obsno	wage	educ	exper	female	married
1	3.10	11	2	1	0
2	3.24	12	22	1	1
3	3.00	11	2	0	0
4	6.00	8	44	0	1
5	5.30	12	7	0	1
÷	ŧ	:	ŧ	ŧ.	1
525	11.56	16	5	0	1
526	3.50	14	5	1	0

A Data	Set on Economic	Growth Rates ar	nd Country Chara	cteristics
obsno	country	gpcrgdp	govcons60	second60
1	Argentina	0.89	9	32
2	Austria	3.32	16	50
3	Belgium	2.56	13	69
4	Bolivia	1.24	18	12
E	1	4	1	E
61	Zimbabwe	2.30	17	6

Cor	nt. Structure of economic data
Т	ime series data
	Time series data has a separate observation for each time period $- e.g.$ table 1.3.
	Since <i>rarely a random sample</i> , we have to consider different problems about it.
۲	It has some category depending on time frequency.
	Special Lecture 8

Minimum Wage, Unemployment, and Related Data for Puerto Rico					
obsno	year	avgmin	avgcov	unemp	gnp
1	1950	0.20	20.1	15.4	878.7
2	1951	0.21	20.7	16.0	925.0
3	1952	0.23	22.6	14.8	1015.9
1	I.	1	1	1	1
37	1986	3.35	58.1	18.9	4281.6
38	1987	3.35	58.2	16.8	4496.7

D 1 1	
Pooled cros	s section
· · · · · · · · · · · · · · · · · · ·	random cross sections to account fferences –e.g. table 1.4.
Panel data	
♦ It consists	of the same random individual
observation	ns over time –e.g. table 1.5.

	Poole	Pooled Cross Sections: Two Years of Housing Prices					
obsno	year	hprice	proptax	sqrft	bdrms	bthrms	
1	1993	.85500	42	1600	3	2.0	
2	1993	67300	36	1440	3	2.5	
3	1993	134000	38	2000	4	2.5	
8	E	1	1	н	13	- E3	
250	1993	243600	41	2600	4	3.0	
251	1995	65000	16	1250	2	1.0	
252	1995	182400	20	2200	4	2.0	
253	1995	97500	15	1540	3	2.0	
ŧ	E.	÷	1	1	ŧ	E.	
520	1995	57200	16	1100	2	1.5	

	A T	Two-Year Panel Data Set on City Crime Statistics						
obsno	city	year	murders	population	unem	police		
1	1	1986	5	350000	8.7	440		
2	1	1990	8	359200	7.2	471		
3	2	1986	2	64300	5.4	75		
4	2	1990	1	65100	5.5	75		
		÷		Ŧ		1		
297	149	1986	10	260700	9.6	286		
298	149	1990	6	245000	9.8	334		
299	150	1986	25	543000	4.3	520		
300	150	1990	32	546200	5.2	493		

	goal is to infer that one
variable has	a <i>causal effect</i> on another
variable, for	testing economic theory or for
evaluating po	plicy.
	ct: A <u>ceteris paribus</u> change in one an effect on another variable.
 Ceteris par 	ibus: All other relevant factors are



The error term, u , ir	cludes other factors
이렇게 다운 가지 다 다 한 것이라. 이렇게 다 한 것이 있는 것이 하는 것 같이 많이 가지?	like experience or job
training.	1 5
• The estimate of β_1	is the return to education
∂Earnings	∂ <i>u</i>
$\beta_1 = \frac{\beta_1}{\partial aducation}$	$s.t.\frac{\partial u}{\partial education} = 0$

	ty & Ceteris Paribus cont.
엄마 이상 이 이 이 이 한 것이라.	establishing a relationship between es is rarely sufficient.
variabl	ve truly controlled for enough other es, then the estimated ceteris paribus can often be considered to be causal.
	metric methods can simulate a paribus experiment.